BudgetLongformer: Can we Cheaply Pretrain a SotA Legal Language Model From Scratch?

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Abstract

Pretrained transformer models have achieved state-of-the-art results in many tasks and benchmarks recently. Many state-of-the-art Language Models (LMs), however, do not scale well above the threshold of 512 input tokens. In specialized domains though (such as legal, scientific or biomedical), models often need to process very long text (sometimes well above 10000 tokens). Even though many efficient transformers have been proposed (such as Longformer, BigBird or FNet), so far, only very few such efficient models are available for specialized domains. Additionally, since the pretraining process is extremely costly in general - but even more so as the sequence length increases – it is often only in reach of large research labs. One way of making pretraining cheaper is the Replaced Token Detection (RTD) task, by providing more signal during training, since the loss can be computed over all tokens. In this work, we train Longformer models with the efficient RTD task on legal data to showcase that pretraining efficient LMs is possible using much less compute. We evaluate the trained models on challenging summarization tasks requiring the model to summarize long texts to show to what extent the models can achieve good performance on downstream tasks. We find that both the small and base models outperform their baselines on the in-domain BillSum and out-of-domain PubMed tasks in their respective parameter range. We publish our code and models for research purposes.

1 Introduction

Pretrained transformer models have achieved excellent performance across various Natural Language Processing (NLP) tasks such as Text Classification (TC), Named Entity Recognition (NER), Question Answering (QA) and summarization Devlin et al. (2019); Yang et al. (2020); He et al. (2021); Zhang et al. (2020a). Pretraining is very resource intensive (especially for large models), thus making it costly and only available for large organizations Sharir et al. (2020). The Masked Language Modeling (MLM) task has been very successful, with many models adopting the task in their pretraining Devlin et al. (2019); Liu et al. (2019); Belt-



Figure 1: Results on the BillSum dataset. Note that the x-axis is in log-scale.

agy et al. (2020); Zaheer et al. (2021). Since typically only 15% of the tokens are masked, the loss can be computed for those tokens only. Clark et al. (2020) introduced the Replaced Token

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Detection (RTD) task, which enables the loss to be computed on all tokens, making training more efficient. On the GLUE benchmark Wang et al. (2018), their ELECTRA model matches RoBERTa Liu et al. (2019) and XLNet Yang et al. (2020) using 1/4 their compute. Although ELECTRA's training strategy seems very promising, to the best of our knowledge, only few works have adopted the RTD task so far He et al. (2021); Kanakarajan et al. (2021).

On another note, domain-specific pretraining has been shown to improve downstream performance in many domains such as law Chalkidis et al. (2020); Xiao et al. (2021), biology Lee et al. (2019), scientific articles Beltagy et al. (2019), clinical documents Li et al. (2022), or even code Chen et al. (2021). Domain-specific pretraining coupled with the RTD task, however, has not been studied in the legal domain so far. Depending on the domain, documents might be extremely long. Texts from the legal domain, for example, tend to span multiple pages, ranging from 10s to 100s of pages, which translates to tens of thousands tokens. The quadratic time and memory requirement of the attention typically used in the transformer architecture Vaswani et al. (2017) prohibits efficient processing of sequences longer than 512 tokens on current hardware. A rich body of research investigates how transformers can be adapted to efficiently process longer input Tay et al. (2020b); Child et al. (2019); Beltagy et al. (2020); Zaheer et al. (2021); Roy et al. (2021); Kitaev et al. (2020); Tay et al. (2021); Lee-Thorp et al. (2021). Longformer Beltagy et al. (2020) is one of these efficient transformer architectures for long sequences, leveraging windowed and global attention. So far, to the best of our knowledge, there does not yet exist a public Longformer model pretrained on English legal data², although Xiao et al. (2021) have proven the effectiveness of the Longformer in dealing with long legal text in many Chinese-related tasks. This work aims to fill this gap.

To test the ability to grasp long-distance dependencies in the text, we mainly evaluated our Language Models (LMs) on the task of automatic (abstractive) summarization. It consists of capturing the most important concepts/ideas from the (long) document and then rewriting it in a shorter passage in a grammatical and logically coherent way Chen et al. (2019). In particular, we used the BillSum benchmark, as a domain-specific summarization task, obtaining a new state-of-the-art (SOTA) (see Figure 1); and the PubMed benchmark, to evaluate the model's ability outside the legal context (i.e., in the biomedical context), obtaining comparable metrics even though the LM has only been pretrained on legal data and the tokenizer is also optimized for legal data (see Figure 2).

We emphasize that this performance was achieved with a minimal pretraining phase due to the combination of the RTD task and the Longformer infrastructure, making our LM very attractive from the point of view of building costs. For instance, our model saw only 3.2M examples during pretraining, whereas RoBERTa Liu et al. (2019) or PEGASUS-large Zhang et al. (2020a) saw 4.1B examples. RoBERTa was trained for 1024 GPU days, whereas our small and base models only used 12 and 24 GPU days respectively (16GB NVIDIA V100 GPUs for both models). Since many tasks in legal NLP are formulated as TC problems, a hierarchical architecture has been used frequently to process long documents Chalkidis et al. (2019); Niklaus et al. (2021). However, this simple hierarchical architecture cannot be easily adapted to solve the more complex sequence-to-sequence tasks like token classification or summarization. For this reason, in this work we pretrain a more versatile Longformer model. We discuss related work in more detail in Appendix C.

Contributions The contributions of this paper are three-fold:

- We train and release a new model pretrained on recently published curated English legal text Henderson et al. (2022), capable of handling input spans longer than 512 tokens out of the box. We train our models by applying the promising, but seldom used RTD task Clark et al. (2020) on a Longformer model Beltagy et al. (2020), for the first time, calling it BudgetLongformer.
- On the BillSum benchmark Kornilova and Eidelman (2019), our models are a new SOTA compared to models of the same size. Especially, our small model outperforms all baseline approaches, and a transformer base model Vaswani et al. (2017) containing almost 4 times more encoder parameters (110M vs. 29M). It even outperforms the PEGASUS base model Zhang et al. (2020a) whose encoder is also almost 4 times larger and was pretrained specifically for the abstractive summarization task.
- We verified that pretraining with the RTD task is suitable for down-stream summarization tasks by evaluating our model on an out-of-domain benchmark (PubMed), obtaining comparable results with summarization-specific architectures.

²On the web there is a model based on Longformer in a legal domain but no link how it was obtained and on its actual performance (https://huggingface.co/saibo/legal-longformer-base-4096).

Main Research Questions In this work, we pose and examine four main research questions:

RQ1: Is it possible to generate an ad-hoc LM with domain (e.g. legal) expertise from scratch, reducing costs and CO_2 emissions?

RQ2: Is it possible to pretrain a Longformer model with the RTD task (aka BudgetLongformer)?

RQ3: How does our BudgetLongformer compare with other models on the challenging summarization task? Particularly in the case of a legal domain-specific benchmark such as BillSum?

RQ4: How well does our BudgetLongformer generalize to other domains, for example in the biomedical domain, as evaluated by the PubMed summarization benchmark?

2 Datasets

Pile of Law Henderson et al. (2022) recently released a large-scale English corpus suitable for pretraining LMs. It contains 256 GB of diverse legal text in English from various jurisdictions and judicial bodies including for example bills, court decisions and contracts from the US, Canada, and Europe even though the focus clearly lies on US data. While there are 28 US datasets available (253.25 GB or 99%), there is only 1 Canadian dataset³ (243 MB or 0.09%), 3 European datasets⁴ (2.3 GB or 0.9%), and 2 international datasets⁵ (212 MB or 0.08%). The non-US datasets only cover the categories "Legal Case Opinions and Filings", "Laws" and "Conversations", but do not cover categories "Legal Analyses", "Contracts / Business Documents" and "Study Materials", whereas the US data is much more diverse and covers all categories.

BillSum Kornilova and Eidelman (2019) introduced a legislative summarization dataset from 21K US bills from 1993 to 2018. It is challenging due to the technical nature and complex structure of the bills. Additionally, the bills are rather long, ranging from 5K to 20K characters (\sim 1K to 4K tokens⁶) with their summaries being up to 5K characters (\sim 1K tokens) long (more details in Appendix L).

PubMed Cohan et al. (2018) introduced another challenging summarization dataset in a specialized domain (scientific articles from the biomedical domain). It includes 133K scientific papers together with their abstracts in English. The papers are 3K words long on average and the summaries (abstracts) 200 words. Thus, similar to the BillSum dataset, this dataset is well suited as a test bed for methods capable of long document summarization. Note, that in this dataset the domain is vastly different from the legal domain (see Appendix L for more details).

3 Results

We describe the detailed experimental setup in Appendix D respectively. Table 2 in Appendix F compare the models evaluated on the summarization benchmarks.

BillSum Our results on the BillSum dataset are presented in Figure 1 and Table 4 in Appendix G. We observe that even our small diverse model clearly exceeds the baseline of the original article (DOC + SUM), even though their model is based on BERT-large which contains almost 12 times more encoder parameters and has been pretrained for 10 times more steps. Even more surprisingly, our small diverse model is on par with the PEGASUS-base model Zhang et al. (2020a) (37.58 vs. 37.78 Rouge-L), pretrained using the Gap-Sentences task specifically designed for abstractive summarization. Furthermore, their model contains almost 4 times more



Figure 2: Results on the PubMed dataset. Note that the x-axis is in log-scale.

encoder parameters and has seen 40 times more training examples during pretraining (128M vs. 3.2M; see Table 2 in Appendix F). By scaling up our model to the base size, we even approach the performance of PEGASUS-large (40.50 vs. 45.8 Rouge-L). PEGASUS-large has seen three orders of

³Canadian Court Opinions (ON, BC)

⁴European Court to Human Rights (ECtHR) Opinions, EUR-LEX and European Parliament Proceedings Parallel Corpus

⁵World Constitutions and U.N. General Debate Corpus

⁶Our experiments show that using our tokenizer one token corresponds to 5.33 characters on average.

magnitude more training examples during its pretraining in comparison to our model (4.1B vs. 3.2M) and contains more than twice as many encoder parameters (340M vs. 159M). We conclude that pretraining with the RTD task is highly effective, with minimal compute for long-input summarization in-domain.

PubMed Our results on the PubMed dataset are presented in Figure 2 and Table 5 in Appendix G. Similar to the results on BillSum, our small model outperforms the Transformer-base model by a large margin (23.24 vs. 19.02 Rouge-L) and approaches the PEGASUS-base model (23.24 vs. 25.2 Rouge-L) even though we did not specifically pretrain our model for summarization and our model has seen 40 times fewer examples during pretraining (3.2M vs. 128M). Similar again, we almost reach the performance of PEGASUS-large (26.53 vs. 27.69 Rouge-L) while having seen 1280 times fewer examples during pretraining (3.2M vs. 4.1B). Note, that we pretrain on a much narrower domain than PEGASUS (legal text vs. C4). Our tokenizer and model has never seen medical data during its pretraining phase. Finally, our tokenizer has 1/3 fewer tokens than the PEGASUS tokenizer (64K vs. 96K). In conclusion, pretraining with the RTD task is even effective on an out-of-domain downstream summarization task.

4 Conclusions and Future Work

In this section, we answer the main research questions, give a general conclusion and directions for future work. We discuss limitations and ethical concerns in Appendices A and B respectively.

Answers to Main Research Questions

RQ1: Is it possible to generate an ad-hoc LM with domain (e.g. legal) expertise from scratch, reducing costs and CO_2 emissions? Yes, we showcase in this work that it is possible to pretrain a domain-expertise LM from scratch with minimal compute, achieving comparable performance with methods that have seen more than three orders of magnitude more pretraining examples. Especially when there is no well-performing large teacher model available, our method is advisable.

RQ2: *Is it possible to pretrain a Longformer model with the RTD task (aka BudgetLongformer)?* Yes, in this work, we show that it is possible to pretrain a Longformer model with the RTD task.

RQ3: *How does our BudgetLongformer compare with other models on the challenging summarization task? Particularly in the case of a legal domain-specific benchmark such as BillSum?* Our LMs compare favorably to baselines on the challenging domain-specific summarization benchmark BillSum requiring the models to process long inputs. Our small model outperforms the larger PEGASUS-base model and our base model almost reaches the performance of the larger PEGASUS-large model. Both baselines have been pretrained with much more compute and data and additionally with a pretraining task crafted specifically for summarization.

RQ4: How well does our BudgetLongformer generalize to other domains, for example in the biomedical domain, as evaluated by the PubMed summarization benchmark? Yes, our results on the out-of-domain PubMed summarization benchmark show that our models compare favorably to baselines. Again, our small model outperforms PEGASUS-base and our base model approaches the performance of PEGASUS large.

Conclusion In this work, we show that we can successfully pretrain Longformer models with the RTD task. Using very little pretraining we can achieve SOTA performance on the challenging legal summarization task BillSum, outperforming PEGASUS, that has been pretrained specifically for summarization. Our model even outperforms PEGASUS on the out-of-domain PubMed dataset involving biomedical research articles. To sum up, we present a simple and extremely cheap way of pretraining a long-context LM in cases without the availability of a large teacher model.

Future Work Future work could test these models on further legal downstream tasks such as CUAD Hendrycks et al. (2021) or the recently released MultiLexSum dataset Shen et al. (2022). Additionally, one can test whether the out-of-domain results hold on other out-of-domain summarization datasets, such as BigPatent Sharma et al. (2019) or ArXiv Cohan et al. (2018). Future work could further scale up the models in terms of batch size, number of pretraining steps, number of parameters and amount of data to test what further gains can be achieved. Additionally, to save even more compute and to produce better models, one could investigate how to warm-start an ELECTRA pretraining from existing checkpoints. The difficulty, of course, lies in getting a suitable generator and discriminator trained with the same tokenizer. One possible setup might be Longformer-base as the generator and Longformer-large as the discriminator. Finally, one can investigate the use of other efficient transformers with the RTD task.

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A Limitations

ELECTRA-style training has the disadvantage of the setup being slightly more complicated, requiring a generator and a discriminator. Additionally, the generator should be smaller than the discriminator to ensure stable training. This makes it difficult to warm start from available checkpoints, since two models of different sizes are required. Often, small models are not released, which makes it difficult to warm-start base models using the RTD task. We leave the direction of warm starting a large discriminator with a base generator to future work.

Except for EUR-LEX (1.31 GB or 1.8% of our diverse dataset), our models have only seen US data during the pretraining phase. So, while these models are expected to work well on US data or datasets with similar content such as heavily influenced by the US or mainly common-law based, legal data from Europe for example is expected to look very different (mainly civil-law based except for the UK) and often translated from the original European languages. Thus, our models are not expected to transfer well to such kind of data.

Because of insufficient compute, we were not able to scale up our models in terms of parameter size, batch size and number of pretraining steps. So while we can show that our approach scales well from the small to the base model, it is unknown if this continues to even larger model sizes. Although it is expected to produce better results, we do not know if using a higher batch size and more pretraining steps boosts performance significantly. Additionally, the lacking compute budget made evaluating on more and especially large datasets like BigPatent impossible. Therefore, we cannot give any conclusions at this point to whether our results are robust across a wide range of datasets.

So far, we did not evaluate our summarization models using newer metrics such as BERTScore Zhang et al. (2020b) or BARTScore Yuan et al. (2021). However, our baselines only evaluated using ROUGE, so we would have needed to rerun the baseline experiments to be able to compare our results to on these newer scores.

So far, we did not have the resources to conduct a thorough human expert evaluation of the quality of our summarization outputs. Such an evaluation would be needed for production systems and for better comparison of models. However, it also requires highly educated lawyers and thus a high amount of resources.

For comparing the efficency of pretraining, the number of FLOPs would probably be best. We compared the models' efficiency based on the number of seen examples during pretraining, due to ready availability (most papers report the batch size and the number of steps, but few papers report

the FLOPs). Liu et al. (2019) for example, also report the number of GPU days used which we can also compare to. Devlin et al. (2019), however, trained using TPUs which makes the comparison difficult again.

B Ethics Statement

Pretraining language models is a very compute-heavy process and thus leaves a large carbon footprint Strubell et al. (2019); Patterson et al. (2021). Our method makes significantly reduces the compute requirements and thus the carbon footprint. As with any large LM there is the risk of it producing biased or unfair output. Researchers using the model should put into place respective safeguards to identify biased and/or toxic language.

C Related Work

Domain-Specific Language Models

Previous work showed that domain-specific pretraining shows promising results on datasets of specialized domains such as law Chalkidis et al. (2020); Xiao et al. (2021), biology Lee et al. (2019), scientific articles Beltagy et al. (2019), clinical documents Li et al. (2022), or even code Chen et al. (2021).

Gururangan et al. (2020) show that continued pretraining on a RoBERTa checkpoint on biomedical data, scientific articles in computer science, and reviews, clearly improves downstream performance in the respective domain-specific datasets. The effect was less pronounced on datasets from the news domain, presumably because RoBERTa has seen many news articles in its pretraining already.

Long Document Processing

In the past few years, a vast amount of research has been devoted to addressing the problem of quadratic time and memory complexity associated with the dense attention mechanism Vaswani et al. (2017), practically limiting the maximum sequence length severely (often to 512 tokens) Tay et al. (2020b); Child et al. (2019); Beltagy et al. (2020); Zaheer et al. (2021); Roy et al. (2021); Kitaev et al. (2020); Tay et al. (2021); Lee-Thorp et al. (2021). These research works have given rise to a new class of transformers, referred to as sparse transformers or efficient transformers Tay et al. (2020b). Reducing the cost associated with the computation of the dense attention matrix while maintaining the same performance is the core idea behind efficient transformers. This is often achieved by introducing sparsity in the attention matrix in a variety of ways that may be fixed pattern such as local (windowed) attention Child et al. (2019); Beltagy et al. (2020), global attention Zaheer et al. (2021) or learnable patterns such as routing attention Roy et al. (2021) and LSH attention Kitaev et al. (2020) or a random pattern Zaheer et al. (2021); Tay et al. (2021). Recently, Lee-Thorp et al. (2021) proposed to use Fourier transforms instead of the attention layer. A comprehensive list of efficient transformers and the detailed description of their attention mechanism can be found in the survey by Tay et al. (2020b). Tay et al. (2020a) proposed a series of tasks designed for testing the capabilities of these different models suitable for longer inputs. However, this so-called "Long Range Arena" considers mostly artificial tasks, with the goal of evaluating the models independently of any pretraining.

Efficient Pretraining

ELECTRA-style pretraining Clark et al. (2020) has been shown to reduce training cost substantially, while matching the performance of SOTA LMs. ELECTRA leverages a smaller generator model (discarded after pretraining), that changes some tokens. The larger discriminator model (used for down-stream tasks) must predict for each token if it was changed by the generator or not, similar to how Generative Adversarial Networks (GANs) are trained Goodfellow et al. (2014). This enables the loss to be relevant for every token, leading to much faster and thus more efficient training.

PileOfLaw Subset	Dataset Size	# Words	# Documents
caselaw			
CL Opinions	59.29GB	7.65B	3.39M
diverse			
Total	73.04GB	8.91B	2.1M
CL Opinions	8.74GB	1.13B	500K
CL Docket Entries and Court Filings	17.49GB	1.80B	500K
U.S. State Codes	6.77GB	829.62M	157
U.S. Code	0.27GB	30.54M	43
EUR-Lex	1.31GB	191.65M	106K
Edgar Contracts	7.26GB	0.97B	500K
Atticus Contracts	31.2GB	3.96B	488K

Table 1: The datasets used for pretraining our models. CL is short for Court Listener

D Experimental Setup

In this section, we describe how we set up the experiments. In all our experiments, we made use of AMP mixed precision training and evaluation to reduce costs and GPU memory. For all our experiments, we used the huggingface transformers library Wolf et al. (2020) available under an Apache 2.0 license.

D.1 BudgetLongformer

In the legal domain it is especially important that models can handle long input. So far, there does not exist an English legal model capable of handling more than 512 tokens. To make pretraining more affordable, we combined the well-proven Longformer model Beltagy et al. (2020) with the RTD task proposed by Clark et al. (2020).

D.2 Tokenizer

We trained a byte-level BPE tokenizer Wang et al. (2019) similar to Beltagy et al. (2020). To encode the complicated legal language well, we chose a relatively large vocabulary of 64K tokens (additionally, we did not apply any preprocessing/cleaning of the input texts). We trained the tokenizer using the huggingface tokenizers library⁷ on the entire PileOfLaw training split (\sim 192GB, \sim 22.5B tokens, \sim 7.5M documents), covering a wide array of English legal texts, mostly from the US.

D.3 Pretraining

We trained the *caselaw* models on the training subset "Court Listener Opinions" from the PileOfLaw (59.3 GB, 7.65B words, 3.39M documents). The *diverse* models were trained on caselaw ("Court Listener Opinions" & "Court Listener Docket Entry Documents"), legislation ("US Code", "State Codes" & "EURLEX") and contracts ("Atticus Contracts" & "EDGAR Contracts"). To balance the training data, we limited the number of documents to 500K (this affects Court Listener Opinions, Court Listener Docket Entry Documents and EDGAR Contracts. Please see Table 1 for more details. Our validation set consisted of 1000 randomly selected examples from the respective training set.⁸

To maximally use the available data, we concatenated all the examples and then cut them off in slices of the model's maximum sequence length (4096). We did this in batches of 1000 examples with multiprocessing to speed up data preparation. The last slice in each batch will not contain 4096 tokens, so we dropped it.

⁷https://github.com/huggingface/tokenizers

⁸We used such a relatively small validation set to save compute.

We trained both a small (29M parameters) and a base (159M parameters) model for each configuration. To reach 100K steps it took a bit less than 3 days for the small model and a bit less than 6 days for the base model on 4 16GB NVIDIA V100 GPUs. The achieved training and evaluation losses are shown in Table 7 in Appendix H. Interestingly, we find that the diverse models achieve lower training and evaluation losses. Please find more details in Appendix I.

Henderson et al. (2022) have experienced difficulties when the language model was trained on the entire Pile-of-Law. We believe that the highly imbalanced dataset concerning text types (contracts, court decisions, legislation, etc.) is the main reason for the training instability. This is one of the reasons why we adopted the procedure described above. As shown later in the results (see Section 3), our pretraining was stable. On the contrary, the diverse model – includes more lexical and layout diversity of documents – turns out to perform better and train more robustly on the summarization tasks.

D.4 Downstream Benchmarks

BillSum

When finetuning on the BillSum dataset Kornilova and Eidelman (2019) we used the following hyperparameters. We trained using early stopping with patience of 3 epochs. We paired our pretrained encoder model with a randomly initialized bart-base decoder model Lewis et al. (2020).⁹ We used a batch size of 32 and learning rate of 7e-5 after tuning in {5e-4, 9e-5, 7e-5, 5e-5, 3e-5, 1e-5}. We used the bart-base default config for num_beams (4) and no_repeat_ngram_size (3). We set the maximum input length to 1024 and the maximum target length to 256 to save compute. However, many summaries get cut off at 256 tokens. This is why we took our best model and trained it with maximum input length 4096 and maximum target length 1024 (see results in Table 4 and examples in Table 10). Due to high training costs, we only trained it with one random seed (42). Our models contain 29M (small) and 159M (base) parameters in the encoder and 96M parameters in the decoder resulting in a total of 125M (small) and 255M (base) parameters.

PubMed

Additionally, we evaluated on the PubMed summarization task Cohan et al. (2018) using the same settings as for the BillSum task. We set the maximum input length to 4096 and the maximum generation length to 512.

E Additional Experiments on LexGLUE

E.1 Dataset

Chalkidis et al. (2021) recently introduced a benchmark for the English legal domain called LexGLUE. LexGLUE contains six TC tasks and one QA task comprising diverse legal data such as US court decisions and contracts, terms of service documents, EU legislation and cases from the ECtHR. There exists a public leaderboard of diverse models on GitHub¹⁰, with Legal-BERT Chalkidis et al. (2020) performing best.

The LexGLUE benchmark focuses on evaluating LMs in legal TC and QA tasks. In LexGLUE, 4 out of 7 tasks involve documents with input lengths lower than 512 tokens on average. From the remaining 3 tasks, the ECtHR A and B tasks and the SCOTUS tasks involve documents with long span, and the median of the first two is also less than 1000 tokens. Usually, legal documents are much longer than 512 tokens and thus this distribution might not be representative of real-world tasks. Shorter input length tasks may be better handled by short-input models (e.g., BERT, RoBERTa, Legal-BERT, etc.).

⁹Interestingly, the randomly initialized decoder yielded better results than when we used the weights from the pretrained huggingface checkpoint at https://huggingface.co/facebook/bart-base.

¹⁰https://github.com/coastalcph/lex-glue

Results on LexGLUE (small models) The size of the dots indicates the maximum sequence length





E.2 Experimental Setup

We evaluated on LexGLUE Chalkidis et al. (2021) using the publicly available scripts without modification to ensure consistent and comparable results. Because of compute limitations, we ran each experiment with only one random seed (1) and with the default set of hyperparameters. We speculate that hyperparameter tuning could further improve the performance of the proposed model.

E.3 Results

Table 3 in Appendix F compares the models evaluated on the LexGLUE benchmark. Note, that these models differ strongly on many dimensions such as the number and types of training steps, the architecture, and the number of parameters.

Our results on the LexGLUE benchmark are presented in Table 6 in Appendix G and in Figures 3 and 4 for the small and base models respectively. Figure 5 in Appendix G shows all the models evaluated on LexGLUE combined.

From the results shown in Table 6, we can observe that our models do not improve on the SOTA for short input length tasks. This suggests that for such tasks a more accurate description of the first 512 tokens, obtained through a pretraining dataset with a comparable distribution of token inputs, is more appropriate. This could be an explanation for why our base model is not able to beat the trained models in the short input length, but ranks slightly behind.¹¹

Despite the previous statement, we can also note that there is quite a clear correlation between the Micro-F1 and the number of parameters of the model in the case of small-size models. LegalBERT-small is an exception, outperforming DistilBERT but having fewer parameters. But LegalBERT-small has been pretrained on the same data as is contained in 6 out of 7 LexGLUE tasks. It is also likely, that the test sets have been contained in the pretraining data. Our small model is still in this trend of performance to model size, despite having seen much fewer examples during pretraining (almost 200 times fewer than BERT-Tiny). While in the case of the base model, this trend is still true for

¹¹Note that Longformer and BigBird have been warm started from the RoBERTa checkpoint. Thus, they have been trained on short documents extensively during the first pretraining phase. Only in the second stage, these two models were fed long documents.

Results on LexGLUE (base models)

The size of the dots indicates the maximum sequence length



Figure 4: Results on the LexGLUE benchmark (base models). Note that the x-axis is in log-scale.

Model Name	Source	P. Steps (K)	P. BS	# P. Examples (M)	# Enc. Params (M)	Max Seq Len	Vocab Size (K)	PubMed Rouge-L	BillSum Rouge-L
DOC + SUM	Kornilova and Eidelman (2019)	1000	256	256	340	512	30		33.73
Transformer-base	Zhang et al. (2020a)				110	1024	96	19.02	30.98
PEGASUS-base	Zhang et al. (2020a)	500	256	128	110	1024	96	25.23	37.78
PEGASUS-large-C4	Zhang et al. (2020a)	500	8192	4096	340	1024	96	27.69	45.8
BudgetLongformer small diverse	ours	100	32	3.2	29	4096	64	23.24	37.58
BudgetLongformer base diverse	ours	100	32	3.2	159	4096	64	26.53	40.50

Table 2: Abbreviations: P.: Pretraining, BS: Batch Size, Enc.: Encoder, Params: Parameters. Comparison of the models evaluated on the summarization tasks BillSum and PubMed.

the same samples seen, if we leave out Legal-BERT and CaseLaw-BERT for the reasons already expressed. This suggests that potentially extending the pretraining dataset with also short documents might improve the performance of our model in this regime as well. In our case, we avoided focusing too much on this point since the purpose of the paper is to solve the legal long documents as input.

Finally, we did not tune the hyperparameters at all. It is well known that proper hyperparameter tuning and already selecting the right random seeds can significantly influence the downstream performance Liu and Wang (2021); Dodge et al. (2020). Note that especially our small models, like BERT-Tiny and miniLM, lag behind in the UnfairToS task (Macro-F1 score below 15). This could be due to an unlucky random seed (Mosbach et al. (2021) and Dodge et al. (2020) reported training performance strongly dependent on the random seed).

F Overview of Compared Models

In this section, we show detailed overviews of the model specifics (Tables 2 and 3).

G Detailed Results

In this section, we show detailed and comprehensive results of the compared models (Tables 4, 5 and 6 and Figure 5).

Model Name	Source	P. Steps (K)	P. BS	D. Steps (K)	D. BS	WS Steps (K)	WS BS	# P. Examples (M) \downarrow	$\# Params (M) \downarrow$	Max Seq Len↑	Vocab Size (K)	LexGLUE Micro-F1 ↑
small models												
BERT-Tiny	Turc et al. (2019)	1000	256	1400	256			614.4	4.4	512	31	70.1
miniLM	Wang et al. (2021)	1000	256	400	256			358.4	21	512	30	72.8
DistilBERT	Sanh et al. (2020)	1000	256	500	256			384	66	512	30	75.2
LegalBERT-small	Chalkidis et al. (2020)	1000	256					256	35	512	31	76.7
BudgetLongformer small caselaw	ours	100	32					3.2	29	4096	64	73.9
BudgetLongformer small diverse	ours	100	32					3.2	29	4096	64	73.4
base models												
BERT	Devlin et al. (2019)	1000	256					256	110	512	30	77.8
RoBERTa	Liu et al. (2019)	500	8192					4096	125	512	31	77.8
DeBERTa	He et al. (2021)	1000	256					256	139	512	128	78.3
BigBird	Zaheer et al. (2021)	500	8192			500	256	4224	127	4096	50	78.2
Longformer	Beltagy et al. (2020)	500	8192			65	64	4100.16	149	4096	31	78.5
Legal-BERT-base	Chalkidis et al. (2020)	1000	256					256	110	512	31	79.8
CaseLaw-BERT	Zheng et al. (2021)	2000	256					512	110	512	30	79.4
BudgetLongformer base caselaw	ours	100	32					3.2	159	4096	64	76.0
BudgetLongformer base diverse	ours	100	32					3.2	159	4096	64	76.9

Table 3: Abbreviations: P.: Pretraining, D.: Distillation, WS: Warm Start, BS: Batch Size, Params: Parameters. Comparison of the models evaluated on LexGLUE. In cases where we were not able to find the batch size in the papers, we assumed it to be 256, since this is the most widely used batch size in pretraining and the default for BERT. For DistilBERT we were not able to find the number of distillation steps, so we assumed 500K steps.

Model (max-in-len->max-gen-len)	# Enc. Params \downarrow	Rouge-1 ↑	Rouge-2 \uparrow	Rouge-L \uparrow
DOC + SUM (BERT large)	340M	40.80	23.83	33.73
Transformer base	110M	44.05	21.30	30.98
PEGASUS base	110M	51.42	29.68	37.78
PEGASUS large (C4)	468M	57.20	39.56	45.80
PEGASUS large (HugeNews)	468M	57.31	40.19	45.82
BudgetLongformer small diverse (1024->128)	29M	53.61	33.54	42.50
BudgetLongformer small diverse (1024->256)	29M	49.85	29.63	37.58
BudgetLongformer base diverse (1024->256)	159M	52.70	32.97	40.50
BudgetLongformer base diverse (1024->128)	159M	54.87	35.63	44.21
BudgetLongformer base diverse (4096->1024)	159M	55.45	36.68	43.23

Table 4: Results on the BillSum dataset. Enc. Params is short for Encoder Parameters.

H Pretraining Details

In this section, we show additional details regarding the pretraining process (Table 7).

I Hyperparameters and Training Details

In this section, we present additional details regarding the chosen hyperparameters.

I.1 Pretraining

We pretrained our models with batch size 32 and learning rate 5e-4 and 3e-4 for the small and base models respectively. We used a Longformer attention window of 256. As described in by Clark et al. (2020), we used 10000 warm up steps and a 4 and 3 times smaller generator than the discriminator in the small and base version respectively. In contrast to Clark et al. (2020) we reduced the generator's depth (number of hidden layers) instead of its width (embedding size, hidden size and intermediate size). We used a MLM probability of 25% for the generators.

For running the pretraining, we used an AWS p3.8xlarge instance with 4 16GB NVIDIA V100 GPUs. Training the four models to 100K steps each, took approx. 18 days or 72 GPU days in total. Previous debug runs additionally consumed approx. 3 days or 12 GPU days.

I.2 Downstream Benchmarks

Overall, we found the diverse models to be more robust in finetuning with less failed runs and typically higher performance.

For running the finetuning experiments, we used an AWS p3.16xlarge instance with 8 16GB NVIDIA V100 GPUs. Running the BillSum, PubMed, and LexGLUE experiments including hyperparameter tuning took approximately 25, 7, and 11 GPU days in total respectively.

Model (max-in-len->max-gen-len)	# Enc. Params \downarrow	Rouge-1 ↑	Rouge-2 \uparrow	Rouge-L \uparrow
Transformer base	110M	33.94	7.43	19.02
PEGASUS base	110M	39.98	15.15	25.23
PEGASUS large (C4)	468M	45.49	19.90	27.69
PEGASUS large (HugeNews)	468M	45.09	19.56	27.42
BudgetLongformer small diverse (1024->128)	29M	37.64	15.72	26.03
BudgetLongformer small diverse (1024->512)	29M	34.98	13.56	23.24
BudgetLongformer base diverse (1024->128)	159M	39.19	16.93	27.21
BudgetLongformer base diverse (1024->512)	159M	41.16	18.15	26.53

Table 5: Results on the PubMed dataset. Enc. Params is short for Encoder Parameters.

model	ECtHR A	ECtHR B	SCOTUS	EUR-LEX	LEDGAR	UNFAIR-ToS	CaseHOLD	Average
small models								
BERT-Tiny	63.7 / 44.0	63.9 / 50.4	61.1/35.7	57.9/25.0	83.8 / 73.3	93.9 / 11.1	66.2	70.1 / 43.7
miniLM	67.9 / 55.1	66.6 / 61.0	60.8 / 45.5	62.2/35.6	86.7 / 79.6	93.9 / 13.2	71.3	72.8 / 51.6
DistilBERT	69.9/61.1	70.5 / 69.1	67.0/55.9	66.0/51.5	87.5 / 81.5	97.1 / 79.4	68.6	75.2 / 66.7
LegalBERT-small	70.4 / 62.6*	71.3 / 69.4*	71.3 / 59.7*	66.1 / 48.2*	87.8 / 82.0*	97.4/81.7	72.9*	76.7 / 68.1
BudgetLongformer small caselaw	65.0 / 46.4	75.3 / 58.2	70.6 / 50.8*	58.1/24.2	85.5 / 76.7	89.5 / 10.5	71.9*	73.7 / 48.4
BudgetLongformer small diverse	64.3 / 47.1	74.4 / 49.4	68.3 / 45.6*	61.5 / 30.8*	85.5 / 76.7*	88.9 / 10.5	70.8*	73.4 / 47.3
base models								
BERT	71.2 / 63.6	79.7 / 73.4	68.3 / 58.3	71.4 / 57.2	87.6/81.8	95.6/81.3	70.8	77.8 / 69.5
RoBERTa	69.2 / 59.0	77.3 / 68.9	71.6/62.0	71.9 / 57.9	87.9 / 82.3	95.2 / 79.2	71.4	77.8 / 68.7
DeBERTa	70.0 / 60.8	78.8 / 71.0	71.1 / 62.7	72.1 / 57.4	88.2 / 83.1	95.5 / 80.3	72.6	78.3 / 69.7
BigBird	70.0 / 62.9	78.8 / 70.9	72.8 / 62.0	71.5 / 56.8	87.8 / 82.6	95.7 / 81.3	70.8	78.2 / 69.6
Longformer	69.9 / 64.7	79.4 / 71.7	72.9 / 64.0	71.6/57.7	88.2 / 83.0	95.5 / 80.9	71.9	78.5 / 70.5
CaseLawBERT	69.8 / 62.9	78.8 / 70.3	76.6 / 65.9*	70.7 / 56.6	88.3 / 83.0	96.0 / 82.3	75.4*	79.4 / 70.9
LegalBERT-base	70.0 / 64.0*	80.4 / 74.7*	76.4 / 66.5*	72.1 / 57.4*	88.2 / 83.0*	96.0 / 83.0	75.3*	79.8 / 72.0
BudgetLongformer base caselaw	67.2 / 55.9	76.6/61.1	74.9 / 62.3*	64.7 / 42.9	86.9 / 80.4	89.5 / 10.5	72.1*	76.0 / 55.0
BudgetLongformer base diverse	66.3 / 52.6	77.9 / 72.3	75.4 / 62.9*	65.6 / 44.4*	87.0 / 81.0*	95.1 / 76.7	71.3*	76.9 / 65.9

Table 6: Results on LexGLUE. Because of limited compute, we only ran 1 random seed for our models. The other results are reported on GitHub¹². The asterix denotes datasets which are (partly) covered in the pretraining dataset. For each column we report the results in the format micro-averaged F1 score / macro-average F1 score. For the CaseHOLD task, both scores are the same.

J Library Versions

We used the following versions to the libraries in a pip requirements.txt format: datasets==2.4.0 huggingface-hub==0.9.0 nltk==3.7 pandas==1.3.5 rouge-score==0.1.2 scikit-learn==1.0.2 scipy==1.7.3 tokenizers==0.12.1 torch==1.12.1 tqdm==4.64.0 transformers==4.21.1

K Examples

Example summaries are displayed in Tables 8, 9, 10, 11, and 12. Since the documents are very long sometimes, we truncated them to the first 2500 characters. We sorted the examples by RougeL scores and show the bottom 5%, bottom 25%, top 75% and top 95% percentile.

Results on LexGLUE The size of the dots indicates the maximum sequence length



Figure 5: Results on the LexGLUE benchmark (all models). Note that the x-axis is in log-scale.

Model	Data	# Steps	Train Loss	Eval Loss
small	caselaw	50K	14.61	15.78
small	caselaw	100K	13.93	15.07
small	diverse	50K	13.75	12.70
small	diverse	100K	12.78	11.66
base	caselaw	50K	12.40	13.76
base	caselaw	100K	11.67	12.99
base	diverse	50K	10.70	10.01
base	diverse	100K	9.86	9.22

Table 7: Training and Evaluation losses for the different trained models. Note that these losses are the addition of the loss of the generator and the loss of the discriminator. Since the loss of the discriminator is much smaller, it is scaled by a factor of 50 to stabilize training.

	example (Sorted by rougeL)
Document	-(1) For parposes of subsection (a)(2) and this subsection, the term 'joint resolution' means only a joint resolution introduced by a qualifying Member specified in paragraph (2) after the date on which the report of the President mode subsection (a)(1) is received by the Congress—'(A) the matter after the resolving clause of which is as follows: 'That the Congress hereby concers in the certification of the President relating to deployment of a National Missile Defense system a submittee of Congress parameters to section 4(b) of the National Missile Defense Act of 1999; '(B) which does not have a preamble; and '(C) the tild of which is observed in the report of a National Missile Defense system as submittee or a Mathematical statical Missile Defense (A) of the National Missile Defense (A); '(A) is the case of the Herose of Representatives to the majority leader of the Hone of Representatives of the Statical Missile Defense (A); '(A) is the case of the Evolution (A) and the subsection, a qualifying Member does Acted in this paragraph i: '(A) is the case of the Evolution (Representatives) and the Statical Missile Defense (A); '(A) is the case of the Evolution (Representatives) and the Statical Missile Defense (A); '(A) is the case of the Evolution (Representatives) and the Statical Missile Defense (A); '(A) is the case of the Evolution (Representatives) and the Statical Association' (Representatives) (Representatives); '(A) is the case of the Evolution (Representatives); Reader of the Freedomic (Representatives); Reader of the Statical Missile Defense (Representatives); Reader of the Representatives of interval (Representatives); Reader of the Representatives of in
	SEC 4. LUMITATION ON OBLIGATION OF FUDDS FOR PROCUREMENT FOR NATIONAL MISSILE DEFENSE SYSTEM. (a) Limitation—No funds appropriated to the parament of Defense for procurement may be obligated for the National Missile Defense system unless—(1) the President submits to Congress a report concerning in the National Missil Defense system against countermeasures that includes a certification described in subsection (b); and (2) a joint resolution concurring in the President's certification in such report is enacted as provided for in this section. (b) President's adequate toxing program for the National Missile
Gold	National Missile Defense Deployment Criteria Act of 2001 - Amends the National Missile Defense Act of 1999 to allow deployment of a national missile defense system (system) only if: (1) the system is technologically feasible; (2) syster
	cost in relation to other Department of Defense (DOD) priorities will not lead to an overall reduction in national security by reducing resources available for other defense priorities; (3) the system will not diminish overall U.S. national security (4) the system will not threaten to obstrupt relations threads. Its assis, the Propelse Republics of China, and other andors, and (5) the thread of a long-regulation seture that (K on a nation concern is clearly demonstrated Prohibits the President from directing DOD to deploy a system unless, and until; (1) the President certifics to Congress that abeauts operation has have been met, and (2) a joint resolution is enacte concurring in the President's certifications have been met, and (2) a joint resolution is enacte concurring in the President's certification congress that abeauts operates that abeauts operating that the substruction is enacted concurring in the President's certifications explaints the system uncess, and the president scentification is enacted concurring in the President scentification of the president scentification of the system tests have been were than start been and the system conternet market in the to which the example that is the spin endocurrent in the president scentification of the system test have been were and (2) a joint resolution is enacted concurring in the President scentification of the system test have been were available to which the example that is the spin endocurrent in the system conternet saves in system ground and flight to which and the example that is the spin endocurrent in the president scentification of the system testem conservation is enacted concurring in the President scentification of the system testem conternet saves in system or the system testem conternet saves in system conternets and the spin endocurrent in the system testem conternet saves in system conternets and the system testem conternet saves in system conternets and the system testem conternet saves in the system testem conternet saves in the system t
Model	prohibits funds appropriated to the department of defense (dod) for procurement from being obligated for the national missile defense system unless the presiden certifies to congress that: (1) an adequate testing program for the system is in place to meet the threats identified in the report; and (2) an adequate ground and flight testing of the system has been conducted against the system that are likely to be used against the system and that other countries have or are likely to acquire.
Metrics	Rouge1: 40.69, Rouge2: 16.67, RougeL: 20.0, RougeLsum: 20.0, Summary length (tokens): 94
Bottom 25%	example (Sorted by rougeL)
Document	TITLE I-FEDERAL ALREPORTS SECURITY ENHANCEMENT ACT SEC: 101. SHORT TITLE This tilt may be cited as the "Federal Airports Security Enhancement Act". SEC: 102. ESTACHISUSHINENT OF AIRPORT SECURITY COMMITTEES. SEC: 103. ENLOYMENT STANDARDROT SECURITY COMMITTEES. SEC: 103. ENLOYMENT STANDARDR AND TRAINING".
	MC: 100 Latricly MEDS 104 ADD ADD SALE (100 MIN). (Identifying: "(1) Review and Recommendations—The Administrator of the Federal Aviation Administration that and the administration of the Administra
	detecting facility or procedure used or operated by an employee or agent of the Federal Protective Service. The Administrator-"(1) shall require that sufficient Federal Police Officers are posted at airport facilities to provide patrol dutie during all hours of operations as well as supervise screening personnel; "(2) shall maintain sufficient numbers of Special Agents to provid
Gold	Federal Airports Security Enhancement Act - Amends Federal aviation law to direct the Administrator of the Federal Aviation Administrator (FAA) to establish at each airport a Security Committee which shall make recommendations. To prescribe appropriate changes on provery the preformance of existing airport security procedures. Requires the Administrator, on the basis of such recommendations, to prescribe appropriate changes in prover the preformance of existing airport security procedures. Requires the Administrator of the General Protective Service employees or agents of an aircraft to existing airregaries (protecting and prover) that private and protective abares (arcif cold to pendopose) or agents, and an aircraft to existing a carrier (architecting and protective abares), special policement and nonuniformed special policements of the General Services Administrator (GSA) to appoint police officers and special agents (currently, screening is carried) on thy Service adjusce the control property. Stathlises the Section Protective Service as a separate operating service of the GSA). Sets for the Section Service police officers to be assigned to areas controlled by the United States and GSA. Calls for at least 1,000 full-time equivalent Service police officers to be assigned to areas controlled by the United States and GSA. Autointorize GSA to recover aiprorts Castrilly Section (areas controlled by the United States and GSA). Autointorize of Sate officers on the sassigned to areas controlled by the United States and GSA.
Model	table of contents: title i: federal airports security enhancement act title ii: miscellaneous provisions general federal airports security enhancement act - title i federal airports security enhancement - amends the federal aviation act of 1992 to direct the administrator of the federal aviation administration (faa) to prescribe regulations requiring screening of all passengers and property that will be carried in a port of aircraft in air transportation or intrastate air transportation. (see 102) directs the administrator to prescribe regulations requiring screening of all passengers and property that will be carried out by the federal protective service the federal bureau of investigation (fbi), the federal bureau of investigation (fbi), and one member from each local jurisdiction that the aircraft may be located in or that may have jurisdictional authority for the airport of an aircraft in air transportation or intrastate air transportation. (sec. 103) directs the administrator to prescribe regulations requiring screening of all passengers and property that will be carried out by a weapon detection facility or procedure used or operated by an employee or agent of the federal protective service. (sec. 103) authorizes the administrator to enter into agreements with state and local law enforcement authorities to obtain authority for, jointly with state and local law enforcement authorities. (
Metrics	Rouge1: 52.44, Rouge2: 22.84, RougeL: 29.7, RougeLsum: 47.8, Summary length (tokens): 256
Top 75% ex	ample (Sorted by rougeL)
Document	SECTION 1. SHORT TITLE. This Act may be called as the "Pattern and Trademark Office Authorization Act of 2002". SEC. 2. AUTHORIZATION OF AMOUNTS AVAILABLE TO THE PATENT AND TRADBRAK OFFICE. (a) In General-There are authorized to be appropriated to the fullid Status Fattern and Trademark Office for salaries and necessary expenses for each of the fiscal years 2003 through 2008 an anount equal to the fees estimated by the Secretary of Commerce to be collected in each such fiscal year, respectively, under-(1) title 3S, United Status Code; and (2) the Act entitled "An Act to provide for the registration and protection of trademarks and incommerce, to carry out the provisions of externing international conversions, and for other proposes", approved 13(b) y 1964 (51 USZ): CID (18 et saq.) (comondy referred to a such Trademark Act of 1946). (b) Estimates-Note Hart There Trays 1) y (c) each ficial year, if (c) e
Gold	Patent and Trademark Office Authorization Act of 2002 - Authorizes appropriations to the U.S. Patent and Trademark Office for salaries and expenses for FY 2003 through 2008 in an amount equal to all patent and trademark for each strain strain of the Secretary of Commerce (Isomatoric December 2000) through 2008 in an amount equal to all patent and trademark for each strain to report the Secretary of Commerce (Isomatoric December 2000) through 2008 in an amount equal to all patent and trademark for each strain to report strain of the Office of the Office (Patent and Trademark Confice of the Office (Patent and Trademark December 2000) in the Office of the Office (Patent and Trademark Patent and Trademark December 2000) in the Office of the Office (Patent and Trademark Patent and
	mai decision lavoratore ou ne patentatorni o mejori enteresta enterest enteresta enteresta enterest enteresta enteresta entere
Model	patent and trademark office authorization act of 2002 - authorizes appropriations to the u.s. patent and trademark office for fy 2003 through 2008. requires the director of the patent and trademark office to: (1) complete the development of an electronic system for the filing and processing of patent and trademark
	patent and trademark office authorization act of 2002 - authorizes appropriations to the u.s. patent and trademark office for fy 2003 through 2008. requires the director of the patent and trademark office to: (1) complete the development of an electronic system for the filing and processing of patent and trademark applications; and (2) submit an annual report to the congressional committees on progress made in implementing the 21st century strategic plan issued under the
Metrics Top 95% ex	patent and trademark office authorization act of 2002 - authorizes appropriations to the u.s. patent and trademark office for fy 2003 through 2008. requires the director of the patent and trademark office to: (1) complete the development of an electronic system for the filing and processing of patent and trademark applications; and (2) submit an annual report to the congressional committees on progress made in implementing the 21st century strategic plan issued under the federal patent and trademark programs.
Model Metrics Top 95% ex Document	Rouge1: 48.99, Rouge2: 39.86, RougeL: 44.3, RougeLsum: 48.32, Summary length (tokens): 94 mple (Sorted by rougeL) SECTION 1. SHORT ITTLE This Act may be cited as the "Guidance, Understanding, and Information for Dual Eligibles (GUIDE) Act". SEC. 2. FNDIORS, FURPOSE. (a) Findings-The Congress finds the following: (1) Nearly 8,800,000 Americans were eligible for benefits under the Medicare program and for medical assistance under Medicaid (dual eligible beneficiaries) in fiscal year 2005. Of these "dua processing of the Society
Metrics Top 95% ex	patent and trademark office authorization act of 2002 - authorizes appropriations to the u.s. patent and trademark office for fy 2003 through 2008. requires the director of the patent and trademark office to: (1) complete the development of an electronic system for the filing and processing of patent and trademark applications; and (2) submit an annual report to the congressional committees on progress made in implementing the 21st century strategic plan issued under the federal patent and trademark programs. Rouge1: 48.99, Rouge2: 39.86, RougeL: 44.3, RougeLsum: 48.32, Summary length (tokens): 94 mple (Sorted by rougeL) SECTON 1. SHORT TITLE This Art my be told as the "Guidence, Understanding, and Information for Dual Eligibles (GUIDE) Ast". SEC. 7. FNDINGS: FURPOSE. Guide as the "Guidente, Understanding, and Information for Dual Eligibles (GUIDE) Ast". SEC. 7. FNDINGS: FURPOSE. Guide as the "Guidente, Understanding, and Information for Dual Eligibles (GUIDE) Ast". SEC. 7. FNDINGS: FURPOSE. Distribution for Dual Eligibles (GUIDE) Ast". SEC. 7. FNDINGS: FURPOSE. Distribution for Dual Eligibles (GUIDE) Ast". SEC. 7. FNDINGS: FURPOSE. Distribution for Dual Eligibles (GUIDE) Ast". SEC. 7. FNDINGS: FURPOSE. Distribution for Dual Eligibles (GUIDE) Ast". SEC. 7. FNDINGS: FURPOSE. Distribution for Dual Eligibles (GUIDE) Ast". SEC. 7. FNDINGS: FURPOSE. Distribution for Dual Eligibles (GUIDE) Ast". SEC. 7. FNDINGS: FURPOSE. Distribution for Dual Eligibles (GUIDE) Ast". SEC. 7. FNDINGS: FURPOSE. Distribution for Dual Eligibles (GUIDE) Ast". SEC. 7. FNDINGS: FURPOSE. Distribution for Dual Eligibles (GUIDE) Ast". SEC. 7. FNDINGS: FURPOSE. Distribution for Dual Eligibles (GUIDE) Ast". SEC. 7. FNDINGS: FURPOSE. Distribution for Dual Eligibles (GUIDE) Ast". SEC. 7. FNDINGS: FURPOSE. Distribution for Dual Eligibles (GUIDE) Ast". SEC. 7. FNDINGS: FURPOSE. Distribution for Dual Eligibles (GUIDE) Ast". SEC. 7. FNDINGS: FURPOSE. Distribution for Dual Eligibles (GUIDE) Ast". SEC. 7. FNDINGS: FURPOSE. Distribution for Dual Eligibles
Metrics Top 95% ex Document	patent and trademark office authorization act of 2002 - authorizes appropriations to the u.s. patent and trademark office for fy 2003 through 2008. require the director of the platent and trademark office to: (1) complete the development of an electronic system for the filing and processing of patent and trademark programs. and (2) submit an annual report to the congressional committees on progress made in implementing the 21st century strategic plan issued under the federal patent and trademark programs. Rouge1: 48.99, Rouge2: 39.86, Rouge1: 44.3, RougeLsum: 48.32, Summary length (tokens): 94 mple (Sorted by rougeL) SETION 1.9007 TTLE SETION 1.9007 T

Table 8: Examples of the BillSum dataset using the model billsum-1024-256 small diverse

	example (Sorted by rougeL)
Document	SECTION 1. SHORT ITTLE. This Act may be clied as the "Health Coverage Tax Credit Extension Act of 2015". SEC. 2. EXTENSION AND MODIFICATION OF HEALTH COVERAGE TAX CREDT: (6) Extension-S-Mayarganh (B) of section 35(b)(1) of the Internal Revenue Code of 1986 is amended by striking "before January 1, 2014" and inserting "before January 1, 2020". (b) Coordination With Credit for Coverage Under a Qualifit Health Than-Subsection (g) of section 35 (b)(1) of the Internal Revenue Code of 1986 is amended by striking "before January 1, 2014" and inserting "before January 1, 2020". (b) Coordination With Credit for Coverage Under a Qualifit means that the Internal Revenue Code of 1986 is amended - 101 by exclesipanting paragraphs (11) as paragraph (13), and (2) by inserting after paragraph (10) the following new paragraphs: ('11) Election—''(A) means that the section apply of an eligible coverage month. ''B) Timing and applicability of election—''(A) and predict and clieph coverage month is that taxable year and, one made, shall be inveceed with respect to sate horts. ''(12) Conditionition with prevint usat careful - 'A) la general - A taxable section adapt paragraph (11) applies shall not be treated as a coverage month (tas defined in section 368(c)(2)) for purposes of section 368.''(2) conditionition with prevent to sate months. ''(12) Conditionition with advance payment in section apple year= ''(11) section advance as a coverage month (tas defined in section 368(c)(2)) for purposes of section 368.''(2) for purposes of section 368.'' (2) conditionition with advance payment in section apple year=) ''(11) section advance as a coverage month (tas defined in section 368(c)(2)) for purposes of section 368.'' (2) conditionition with advance payment in section advance payme
Gold	Health Coverage Tax Could Elsension Act of 2010 This bill extends the tax credit for health insurance costs of a stargayer and qualifying family numbers through 2010. The tax credit for health insurance costs of a stargayer and qualifying family numbers through 2010. The tax credit for health insurance costs of a stargayer and qualifying family numbers through 2010. The tax credit for health insurance costs of a stargayer and qualifying family and the stargayer in the stargayer and qualifying family insurance costs of a stargayer in the stargay in the stargayer
Model	health coverage tax credit extension act of 2015 this bill amends the internal revenue code, with respect to health care coverage, to: (1) extend through 2020 th tax credit for advance payments to individuals, (2) allow advance payments of advance payments of advance payments, and (3) extend through 2018 the tax credit for advance payments of advance payments of advance payments to individuals.
Metrics	Rouge1: 26.37, Rouge2: 11.07, RougeL: 21.25, RougeLsum: 25.64, Summary length (tokens): 82
Bottom 259	6 example (Sorted by rougeL)
Document	SECTION 1. EXTENSION (a) In General-Chapter 5 of subtile B of the Agricultural Markning Act of 1946 (7 U.S. C. 1656 et sca) 1st annotadely adding at the earth feedborning name acrises. "SEC: 260. TEEMINATION OF AUTHORT "The authoring provided b mits subtile terminates on September 30. 2010". (b) Conforming Amendment and Extrainon-Section 94.0 ft Li-basteck A Mandaroy Reporting and (1996 (7 U.S. C. 1655 et acr. Pbil): Level 96-78) is annotable by striking "terminate on September 30. 2005" and inserting "(other than accions 911 of subtile A and the amendments made by that section 94.0 ft Li-basteck A Mandaroy Reporting and (1996 (7 U.S. C. 1655 et acr. Pbil): Level 96-78) is annotable to read as follows: "(1) Base Market HeysSection 231(1) of such A tar(7 U.S. L. 1655(1)) is amended to read as follows: "(3) Baser, "The term hear market hey "The term hear market hey". The term hear market hey "The term hear market hey" metas as exaultip-inter mate swine". (c) Packer of Soxs and Baser-Section 231(1) of such Act (7 U.S. L. 1655(1)) is amended to read as follows: "(3) Baser, "The term hear market hey", The term incluids col)". (7) a swine proceeding plant that studgethered an average of at Li-basteck Market hey (-1) as swine proceeding plant that studgethered an average of at Li-basteck Market hey (-1) as swine proceeding plant that studgethered an average of at Li-basteck Market hey (-1) as swine proceeding plant that studgethered an average of at Li-basteck Market Market Market (-1) (A strike they preceding five calendar years; and "(ii) a proceeding the second 201000) sors, boars, or any combination thereof, per year during the immediately preceding five calendar years; and "(iii) a proceeding the term industed) sort of the processing plant (B studgethered an average of at Li-basteck Market Market Market (-1) (A strike they proceeding five calendar years; and "(iii) a proceeding the term) and (-1) his printing or person" after "plant terminated strike thereoffice and the processing plant strike studgethere an a
Gold	Amends the Agricultural Marketing Act of 1946 to extend the provisions of the Livestock Mandatory Price Reporting Act of 1999 through September 30, 2010. Amends swine price reporting provisions.
Model	amends the agricultural marketing act of 1946 to terminate the authority of the secretary of agriculture (usda) to: (1) livestock processing plant processing plant slaughter, and (2) slaughtering plant slaughter. amends the agricultural marketing act of 1946 to: (1) revise minimum reporting requirements; and (2) revise reporting requirements. amends the agricultural marketing act of 1946 to: (1) revise minimum reporting act of 1946 to: (1) revise reporting requirements; and (2) revise reporting requirements; and (3) revise reporting requirements; and (4) revise reporting requirements; and (4) revise reporting requirements; and (5) revise reporting requirements; and (5) revise reporting requirements; and (6) rev
Metrics	Rouge1: 33.66, Rouge2: 18.18, RougeL: 31.68, RougeLsum: 29.7, Summary length (tokens): 105
Top 75% ex	ample (Sorted by rougeL)
	This Act may be cited as the "Maritime Administration Authorization Act for Fiscal Year 2001". SEC: 2. AUTHORIZATION OF APPROPRIATIONS FOR THE SICAL YEAR 2001. Funds are breely authorized to be appropriated, as Appropriations Acts may provide, for the use of the use of the Department of Transportation for the Maritime Administration authorized by title XI of the Mechant Maritine Administration authorized by title XI of the Mechant Maritine Administration authorized by title XI of the Mechant Maritine Administration authorized by title XI of the Mechant Maritine Administration as follows: (1) For expenses necessary for operations and training activities, not to exceed \$902,000,000 to be available unit equended. In addition, for administrative expenses related to long aurantee commitments under title XI of that Act, \$4,179,000. SEC: 3. AMENDARISTS TOT TILE FOR THE MERCHANT MARINE ACT, 1995. (a) Talle IX of the Mechant Maritine Act, 1956 (40 LSC, App. 10) et exc) is amended by adding at the end thereof the following: "SEC: 9.10 DECIMENTING TOT FECETARIA DRY CARGO VESSEL3. WHICH MARITING ADDITION OF THE MERCHANT MARINE ACT, 1995. (b) Talle IX of the Mechant Maritine Act, 1956 (40 LSC, App. 10) et exc) is amended by adding at the end thereof the following: "SEC: 9.10 DECIMENTING TOT FECETARIA DRY CARGO VESSEL3. WHICH MARITING ADDITION OF THE MERCHANT MARINE ACT, 1995. (b) Talle IX of the Mechant Maritine Act 1996 (b) Talle Act, 1996 (b) Talle Act, 1996 (b) Talle Act, 1997 (b)
Gold	(Sec. 3) Amends the Merchant Marine Act, 1996 to declare that certain restrictions concerning a vessel built in a forigin country shall not apply to a newly constructed divbuilk or breakbuilk vessel over 7,500 deadweight tons that has beed delivered from a foreign shipyard before the cartier of two specified dates. Deems U.Sbuilt any vessel integle contracted for or onstruction in a foreign shipyard before the earlier of two specified dates. Deems U.Sbuilt any vessel integle contracted for or delivered and documented under U.S. Law, if certai contains earlier to the Secretary of Tansportation, to initiate dates issues in all appropriate interminational forms to establish an international standar for the secretary of Tansportation to develop, and report to specified congressional committees on a program for the scraphing of obselte National Marinite the Secretary of Tansportation to develop, and report to specified congressional committees on a program for the scraphing of obselte National Marinite the Secretary of Tansportation to develop, and report to specified congressional committees on a program for the scraphing of abselte National Marinite Marinitariand for the scraphing of Abselte National Defense Reserve Fleet Vessels. Amends the Secretary to dispose of the maticos of the environment, and safety concerns. Amends Federal law is an of a program for the sational Defense Reserve Fleet Vessels. Amends the Satistican Defense Reserve Fleet vessels. The National Defense Reserve Fleet that may be scraphing of National Defense Reserve Fleet Vessels. Amends the Satistican Defense Reserve Fleet vessels. Amends the scraphing of the science Satistican Defense Reserve Fleet vessels. The National Defense Reserve Fleet that may be scraphing of National Marinite Rationareal marking and anticine and marinite marking and marki
Model	maritime administration authorization act for fiscal year 2001 - authorizes appropriations for the department of transportation (dot) for fy 2001 for: (1) operation and training activities; (2) training activities; and (3) administrative expenses amends the merchant marine act, 1936 to make appropriations for fy 2001 for (1) operation 2001 for the maritime administration.amends the merchant marine act, 1936 to apply certain restrictions concerning a vessel located in a foreign soluty to newly constructed dry or breakable vessel over seven.500 feet that has been delivered from a foreign shipyard or contracted for construction in a foreign shipyard before the earlier of: (1) one year after enactment of this act, or (2) the effective date of the international maritime administration. Liters the secretary of transportation to initiate discussions in all appropriate international forums in order to establish an international standard for the scrapping of vessels in a safe and environmentally sound manner. directs the secretary of state to initiate discussions in all appropriate international standard for establish an international standard for the scrapping of vessels in a safe and environmentally sound manner.
Metrics	Rouge1: 61.19, Rouge2: 41.5, RougeL: 47.76, RougeLsum: 57.21, Summary length (tokens): 222
Top 95% ex	ample (Sorted by rougeL)
Document	SECTION 1. SMALL BUSINESS EXPENSING PROVISIONS MADE FREMANISNT. (a) Increase in Small Business Experiments Made Fernmannet. (b) In general-backetion (b) of section 179 of the Internal Revenue Code of 1986 (relating to limitations) is amended-(A) by striking "\$25,000 (is125,000 in the case (a) increase in Small Business Experiments date 2006 and before 2011)" in paragraph (1) and inserting "\$55,000", and (a) by striking "\$50,000" (\$50,000", and (a) by striking "\$50,000" (\$50,000", a) and (a) by striking "\$50,000", and (a) by striking "\$50,000" (\$50,000", a) and (a) by striking "\$50,000", and (a) budies by an angering (b) (a) and (a) by striking "\$50,000", and (a) budies by striking and (
Gold	amount of the ded Amends the Internal Revenue Code to: (1) increase and make permanent the expensing allowance for depreciable business assets; and (2) allow a tax deduction, up to \$10,000, for the purchase of a motor vehicle manufactured in the Unite
Model	States. Terminates such as deduction after 2010. amends the internal revenue code to make permanent: (1) the increased expensing allowance for depreciable business assets; and (2) the tax deduction for th purchase of manufactured manufactured automobiles.
Metrics	Rouge1: 72.0, Rouge2: 46.58, RougeL: 64.0, RougeLsum: 64.0, Summary length (tokens): 40

Table 9: Examples of the BillSum dataset using the model billsum-1024-256 base diverse

Document	example (Sorted by rougeL) section 1. Short TITLE
Document	SECTION 1 SHORT ITTLE. This Act may be cited as the "Public Health Equity Act". SEC. 2 INDORSO,
	Congress finds that- (1) all communities and individuals are entitled to protection from occupational and other exposure to substances that are hazardous to the public health; (2) hazardous substances have had a dispropertionate impact on the public health of poor and ethnic minority communities and individuals, resulting in exclusion from participation in, denial of benefits under, and discrimination under, programs and activities receiving Federal financial assistance; and (3) eace Federal agency has no obligation to ensure that all federally assisted programs and activities that affect human health do not directly or through contractual arrangements use etiretian, methods, or partices that cause discrimination on the
	ground of race, color, or national origin. SEC. 3. PULLICH HEALTH EQUITY. The Public Health Service Art (42 U.S.C. 201 et seq.) is amended by adding at the end thereof the following new title: "TITLE XXVII-PUBLIC HEALTH FOURTY
	"As used in this tille: "(1) Activity; program.—The term 'program or activity' means any operation of—"(A)(i) a department, agency, special purpose district, or other instrumentality of a State or of a local government; or "(ii) the entity of suc
	State or local government that distributes such assistance and each such department or agency (and each other State or local government that intributes such assistance is extended, in the case of assistance to a State or local government that (and the such assistance) is extended. In the case of assistance is or local government that (and the such assistance) is extended. In the case of assistance is exte
Gold	Public Health Equity Act - Amends the Public Health Service Act to add a new tide, title XXVII: Public Health Equity. Prohibits, under such title, acts with discriminatory effect as well as intentionally discriminatory acts that result in disproportionate exposure to a covered substances, contaminants, or chemicals listed, identified, or defined is specified laws.
Model	public health equity act - amends the public health service act to require the president to ensure that no person shall be excluded from participation in, be denice
	the benefits of, or being subject to discrimination under, any program or activity on the ground of race, color, or national origin. requires the president to ensure that no person shall be excluded from participation in, be denied the benefits of, or being subject to discrimination under, an
	program or activity on the ground of race, color, or national origin. requires the secretaries of labor, health and human services, the administrator of the health and human services, the administrator, and any other head of
	federal agency with responsibility for providing federal financial assistance to a program or activity to issue regulations implementing such nondiscriminatio requirements. requires such regulations to: (1) declare that no person shall be excluded from participation in, be denied the benefits of, or be subject to discrimination unde
	any program or activity on the ground of race, color, or national origin; and (2) address actions of programs or activities that result in disproportionate exposur to a covered substance on the basis of race, color, or national origin.
Metrics	Rouge1: 28.89, Rouge2: 20.15, RougeL: 22.96, RougeLsum: 26.67, Summary length (tokens): 239
	example (Sorted by rougeL)
Document	SECTION 1. SHORT TITLE; REFERENCES TO TITLE 38, UNITED STATES CODE. (a) Short TitleThis Act may be cited as the "Veterans Programs Improvement Act of 2003". (b) ReferencesExcept as otherwise expressly provided, wherever in this Act an amendment is expressed in terms of an amendment to a section of
	other provision, the reference shall be considered to be made to a section or other provision of title 38, United States Code. SEC. 2. INCREASE IN RATES OF DISABILITY COMPENSATION AND DEPENDENCY AND INDEMNITY COMPENSATION.
	(a) Rate Adjustment—The Secretary of Veterans Affairs shall, effective on December 1, 2003, increase the dollar amounts in effect for the payment of disability compensation and dependency and indemnity compensation by the Secretar as specified in subscription (a) December 1, 2003, increase of the dollar amounts to be increased pursuant to subsection (a) and the following: (1) Compensation—Each of the dollar amounts to be increased pursuant to subsection (a) and the following: (1) Compensation—Each of the dollar amounts to infect the other secret and the following: (1) Compensation—Each of the dollar amounts to infect the other secret amount and infect the other secret amount in effect the following: (1) Compensation—Each of the dollar amounts to infect the following: (1) Compensation—Each of the dollar amounts to infect the following: (1) Compensation—Each of the dollar amounts infect the following: (1) Compensation—Each of the dollar amounts infect to the following: (1) Compensation—Each of the dollar amounts infect the following: (1) Compensation—Each of the dollar amounts infect to the following: (1) Compensation (2) Compens
	compensation for dependentsEach of the dollar amounts in effect under section 1115(1), (3) Citofing allowanceTe hdefinal amount in effect under section 1162, (4) New dic ratesEach of the dollar amounts in effect under section 1115(1), (3) Citofing allowanceTe hdefinal amount in effect under section 1116(1), (3) Additional di c fu surviving spousses with minor childrenThe dollar amounts in effect under section 1111(1), (3) Citofinal amount in effect under section 1116(1), (3) Citofinal amount in effect under section 1116(1), (3) Citofinal amount in effect under section 111(1), (3) Citofinal amount i
	for disability-Each of the dultar amounts in effect under subsections (c) and (d) of section 1311. (8) DIC for dependent childrenEach of the dultar amounts in effect under sections 1314, (c) Determination of Increase -(1) Ti increase under subsection (a) shall be made in the dollar amounts specified in subsection (b) as in effect on November 30, 2003; (c) Except as provided in paragraph (3), each such amount shall be increased by the same percentage as th percentage by which benefit amounts payable under title I of the Social Security Act (42 U.S.C) of use Act (42 U.S.C)
C-14	
Gold	Veterans Programs Improvement Act of 2003 - Directs the Secretary of Veterans Affairs to increase, as of December 1, 2003, the rates of veterans' disability compensation, additional compensation for dependents, the clothing allowance f certain disabled adult children, and dependency and indemnity compensation for surviving spouses and children.
	Makes the effective date for the award of death pension the same as that for the award of death compensation or dependency and indemnity compensation. Excludes lump-same from income for purposes of eligibility for veterans' pensions.
	Prohibits the payment of veterans' disability compensation for an alcohol- or drug-abuse related disability even if the the alcohol or drug abuse is secondary to a service-connected disability. Provides alternative beneficiaries for National Service Life Insurance and United States Government Life Insurance proceeds when the first beneficiary does not make a claim.
	Provides burial benefit eligibility for a veteran's surviving spouse who remarries following the veteran's death. Makes permanent the authority for the State cemetery grants program.
	Repeals the Department of Veterans Affairs Education Loan program. Includes self-employment training under the Montgomery GI Bill.
	dependency and indemnity compensation (dic) through the department of veterans affairs (va), to: (1) increase the rates of disability compensation an dependency and indemnity compensation; (2) provide for additional compensation for dependents; (3) provide for additional compensation; (4) exclude lump-sum sales of any life insurance policy or policies on a veteran for purposes of pension benefits; (5) exclude lump-sum sales of any life insurance policy or policies on a veteran for purposes of pension benefits; (6) exclude lump-sum life insurance proceeds from the determinations of annual income for pension purposes; (7) provide for alternative beneficiaries for certain veterans' life insurance policies on a veteran's service-connected disability; an (8) authorize the secretary to approve a program of self-employment on-employment in the department of veterans affairs education loan program.amends th veterans' advisory committee on education to: (1) repeal the requirement that a claimant and the claimant's representative is necessary to complete an applicatio is not received by the secretary to approve a program of self-employment on-employment in the department of state cemetery grants program; an (3) authorize the secretary to approve a program of self-employment on-employment in the department of america known as the department of veterans affairs education low program of self-employment on-employment in the department of america known as the department of veterans affairs education low program of self-employment on-employment in the department of america known as the department of veterans affairs education lowen endown as the department of veterans affairs education lowen as the de
Metrics	(s) autorize the secretary to approve a program of sen-employment on employment in the department of america known as the department of veterans analys Rouge1: 60.71, Rouge2: 29.79, RougeL: 33.88, RougeLsum: 50.82, Summary length (tokens): 297
•	mple (Sorted by rougeL)
Document	
	SECTION 1. SHORT TITLE. This Act may be inde as the "Cameron Gubransen Kids and Cars Safety Act of 2003". SEC. 2. EVALUATION OF DEVICES AND TECHNOLOGY TO REDUCE CHLD INURY AND DEATH FROM PARKED OR UNATTENDED MOTOR VEHICLES. (a) In General-The Secretary of Transportation shall evaluate (b) devices and technologies intended to reduce the incidence of child injury and child death occurring outside of parked motor vehicles in nontraffic, noncrash events, includin backing-over incidents, that are caused by such vehicles, and determining which of those methods is the most effective; and (2) currently available technology to prevent injury and death of children left unattended inside of parked motor vehicles, including injury of each those to hyperfermin, power windows, or power surronds, (b) Report-The Secretary of Transportation shall subations of the evaluation under this section to the Compression by not later than one year after the date of the enactement of this Act, regrarding Power Windows. The Secretary of Transportation shall by not later than 6 months after the submission of the report under subsection (b) complete any rulemaking begun before that of the anot transfing power Windows. The Secretary of Transportation shall subations in other dates to the child data be to include, and complete any rulemapsortanis shall subations in other data. The Secretary of Transportation shall by not later than 6 months after the submission of the report under subsection (b) complete any rulemating begun before that one from that regrarding power Windows. The Secretary of Transportation shall subations and the date of the enactement of this Act regrarding power Windows. The Secretary of Transportation shall advates to include, and collect that regarding, the numbers and types of injuries and defans in nontraffic, noncrash vents. (2) Whether here was an operation of each notary vehicle in sub-that shalt shalts the following information; (1) The types, makes, models, and model yars of moore vehicles in noter the eac
Gold	This Act may be cide at the "Cameron Culbramen Kids and Cans Safety Act of 2005". SEC: 2. EVALUATION OF DEVICES AND TECHNOLOGY TO REDUCE CHILD JUNURY AND DEATH FROM PARKED OR UNATTENDED MOTOR VEHICLES. (a) In General-The Secretary of Transportation shall evaluate-(1) devices and technologies intended to reduce the incidence of child injug and child death occurring outside of parked motor vehicles in nontrafic, nonerash events, includit bucking-wer incidents, that are caused by such vehicles, and determining without of those methods is the most effective, and child death and child death occurring outside of parked motor vehicles in nontrafic, nonerash events, includit bucking-wer incidents, that are caused by such vehicles, and determining without of those methods is the most effective, and click and and without a report on the findings and determining without of those methods is the most effective, and click and and without a report on the findings and determining without exhibitions of the evaluation under this section to Congress by on black for RT MACKED NPT HEVIDERS AND DEVIDENT DEVIDENT ACT ORAPICATE DEVIDENT ACT Transportation and by not later than one or part after that on of the act top length of the matching Regarding power windows and power windows. The Secretary of Transportation and by not later threat mode this act of the char of the act top length and power windows. The Secretary of Transportation and by not later than formation. The Secretary of Transportation shall collect and include in such detabase the following information: (1) The types makes, models, and mode yeas of more vehicles involved in nontrafic, nonerash vevents involving information: (1) The types makes, models, and mode yeas of noner vehicles involved in nontrafic, nonerash vevents involving information: (4) The age of each operator of such motor vehicles. (4) The age of each individual who sufficed injury or death in such events. (5) Whether each motor vehicles in nontrafic, nonerash vector individual sub ostifiered injury
Gold	This Act may be cited as the "Cameron Culbraness Rids and Cans Safety Act of 2005". SEC: 2 EVALUATION OF DEVICES AND TECHNOLOGY TO REDUCE CHILD JUNURY AND DEATH FROM PARKED OR UNATTENDED MOTOR VEHICLES. (a) In General-The Secretary of Transportation shall evaluate-(1) devices and technologies intended to reduce the incidence of child injury and child death of children left unattended inside of parked motor vehicles in nontraffic, noncrash events, includin bucking-wer incidents, that are caused by such vehicles, and determining witch of those methods is the most effective, and child death cocarring outside of parked motor vehicles in nontraffic, noncrash events, includin bucking-wer incidents, that are caused by such vehicles, and determining witch of those methods is the most effective, and child death such on any and extent of this Act (1) Completion of Rulemanking Regarding Power Windows, -The Secretary of Transportation shall should a report on the findings and determining witch child be native the tide of the enactment of this Act (1) Completion of Rulemanking Regarding Power Windows, -The Secretary of Transportation shall should are than on great reflect any rulemaking legan before the date of the enactment of this Act (1) Completion of Rulemanking Regarding Power Windows, -The Secretary of Transportation shall should are than 6 months after the submission of the requires and the state than the submission of the evaluation under this act engrading power windows and power windows witches. 2C: J. DATABASE FOR TRACKING THE VIMBER AND DEATIFIES NO DEATIFIES NO NOVERAFIES (1) (1) (1) (1) (1) (1) (1) (1) (1) (1)
	This Act may be cited as the "Cameron Gulbrames Rids and Can Safety Act of 2007". SEC: 2 EVALUATION OF DEVICES AND TECHNOLOGY TO REDUCE CHILD INURY AND DEATH FROM PARKED OR UNATTENDED MOTOR VEHICLES. SEC: 2 Evaluation of Tanaportation shall evaluate-(1) devices and technologies intended to reduce the incidence of child injury and data deduce of parked motor vehicles in nontraffic, nonerash events, including how the structure of the second seco
Model	This Act may be cide at the "Cameron Gulbraness Rids and Cars Safety Act of 2007". SEC: 2 EVALUATION OF DEVICES AND TECHNOLOGY TO REDUCE CHILD JOILUTY AND DEATH FROM PARED OR UNATTENDED MOTOR VEHICLES. SEC: 2 EVALUATION OF DEVICES AND TECHNOLOGY TO REDUCE CHILD JOILUTY AND DEATH FROM PARED OR UNATTENDED MOTOR VEHICLES. SEC: 2 EVALUATION OF DEVICES AND TECHNOLOGY TO REDUCE CHILD JOILUTY AND DEATH FROM PARED OR UNATTENDED MOTOR VEHICLES. SEC: 2 EVALUATION OF DEVICES AND TECHNOLOGY TO REDUCE CHILD JOILUTY AND DEATH FROM PARED OR UNATTENDED MOTOR VEHICLES. SEC: 2 EVALUATION OF DEVICES AND TECHNOLOGY TO REDUCE CHILD JOILUTY AND DEATH FROM PARED OR UNATTENDED MOTOR VEHICLES. SEC: 2 EVALUATION OF DEVICES AND TECHNOLOGY TO REDUCE CHILD JOILUTY AND DEATH FROM PARED OR UNATTENDED MOTOR VEHICLES. SEC: 2 EVALUATION OF DEVICES AND TECHNOLOGY TO REDUCE CHILD JOILUTY AND DEATH SIN PARED OR UNATON. The Secretary of Transportation shall what and or which the date of the enterment of this Act regramment on All prover Windows: The Secretary of Transportation shall by not later than 6 months after the submission of the evaluation of the calculation and power Windows: And types of jupites and deaths in nonerantific, nonecraft vented is and prover Windows: And types of jupites and deaths in nonerantific, nonecraft vented is involved in the secretary of Transportation shall by not later than 6 months after the submission of the requirement of the Act regramment of
Model Metrics	This Act may be cide at the "Cameron Gulbraness Rids and Care Safety Act of 2007". SEC: 2 EVALUATION OF DEVICES AND TECHNOLOGY TO REDUCE CHILD INURY AND DEATH FROM PARKED OR UNATTENDED MOTOR VEHICLES. (a) In General-The Secretary of Transportation shall evaluate-(1) devices and technologies intended to reduce the incidence of child injury and data document of the intermation of the most effective, and (2) currently available technology to preven injury and death of children left unattaneode inside of parkad most or whicles in nontraffic, noncrash events, includin thorking-nov michaely to prevent injury and death of children left unattaneode inside of parkad most or whicles in nontraffic, noncrash events, includin the complete any relation of the ensement of this Act (2) Completion of Rolemarking Regarding Power Windows, and Power Windows, and Power Windows, and Power Windows, SEC: 3. DATABASE FOR TRACKING THE NUMBER AND TYPES OF INURES AND DEATHS IN NONTRAFFIC, NONCRASH EVENTS. (4) Establishment. The Secretary of Transportation shall collect and include in such database to include, and collect data regarding, the numbers and types of injuries and deaths in nontraffic, noncrash events involvin motor vehicles, (b) Included Information. The Secretary of Transportation shall collect and include in such database to include, and collect data regarding the way and operation database since of a stall cover stalls, and collect data regarding the secret stalls. The secretary of Transportation shall collect and include in such database the following information. The Secretary of Transportation shall collect and include in such database the following information, and excets. (3) Mether ack motor vehicles, in non-traffic, non-crash coverts. (2) Whether levs an a operator for the score tary of Transportation to: (1) evaluate devices and technologies to reduce child injuries and deaths in nontraffic, non-crash events. (2) Information within the stark of the score tary of Transportation to: (1) evaluate devices and te
Model Metrics Top 95% exa	This Act may be cide at the "Cameron Cultimates Rids and Care Safety Act of 2007". SEC: 2 EVALUATION OF DEVICES AND TECHNOLOGY TO REDUCE CHILD INJURY AND DEATH FROM PARKED OR UNATTENDED MOTOR VEHICLES. (a) In General-The Secretary of Transportation shall evaluate-(1) devices and technologies intended to reduce the incidence of child impay and data devices for ordered in jung and data devices for ordered injury and death of children left unattended inside of parks of most effective, and (2) carerently available technology to preven impay and death of children left unattended inside of parks of most effective, and (2) carerently available technology to preven impay and death of children left unattended inside of parks of most effective and (2) carerently available technology to preven indices and the transmission of the regulation of the regulati
Model Metrics Top 95% exa	This Act may be cide at the "Cameron Gulbramen Kids and Care Safety Act of 2007". SEC: 2 EVALUATION OF DEVICES AND TECHNOLOGY DI REDUCE CHILD JULIUP AND DEATH FROM PARKED OR UNATTENDED MOTOR VEHICLES. SEC: 2 EVALUATION OF DEVICES AND TECHNOLOGY DI REDUCE CHILD JULIUP AND DEATH FROM PARKED OR UNATTENDED MOTOR VEHICLES. (a) In General-The Secretary of Transportation shaft evaluate-(1) devices and technologies intended to reduce the incidence of child injup: and child death of child and of child and of the child of child inside of parked motor vehicles in sentimics. Motor and the secretary of Transportation shaft evaluate-(1) devices and technologies intended to reduce the incidence of child injup: child the intervention of the evaluation of Relemaking Regarding Power Windows. SEC: 3 DATABASE FOR TRACKING THE NUBBER AND TYPES OF INIUES AND DEATHS IN NONTRAFFIC. NONCKASH EVENTS (6) Elabolishame. The Secretary of Transportation shaft evaluate- in the data of the entities of the origin of each due to the entities of the chart of the child transportation of the reputation of the child transportation of the reputation of the child transportation of the reputation of the reputation of the child transportation of the child transportation of the reputation of the reputation of the reputation of the child transportation of the reputation of the reputation of the child transportation of the reputation of the reputation of the child transportation of the reputation of the reputatio
Gold Model Metrics Top 95% exa Document	This Act may be cide at the "Cameron Gulbramen Kids and Can Safety Act of 2007". SEC: 2: EVALUATION OF DEVICES AND TECHNOLOGY DEBUCE CHILD NURLY AND DEATH FROM PARKED OR UNATTENDED MOTOR VEHICLES. (a) In General-The Secretary of Transportation shall evaluate-(1) devices and technologies intended to reduce the incidence of child injury and child adath occurring outside of parked motor vehicles in nontraffic, noncrash events, including to Campress by not later than one year after the date of the enactement of this Act regrest realisment of the Act regrest realisment regrest realisment regrest realisment of the Act regrest realisment regrest realisment regrest realisment of the Act regre
Model Metrics Top 95% exa Document	This Act may be cide at the "Cameron Cultomases Risks and Care Safety Act of 2007". SEC: 2 EVALUATION OF DEVICES AND TECHNOLOGY TO REDUCE CHILD INUURY AND DEATH FROM PARKED OR UNATTENDED MOTOR VEHICLES. (a) In General-The Secretary of Transportation shall evaluate-(1) devices and technologies intended to relate the instance of child injury and data devices of parked motor vehicles in nontraffic, nonerash events, includin tacking-over indexings your indexing to park briefs, and determining with he instance interview. The secretary of Transportation shall by net later than on source shall by net later than on source and the maxement of this Act regranding power Windows: Nuclea. SEC: 3 DATABASE FOR TRACKING THE NUMBER AND TYPES OF INURES AND DEATHS IN NONTRAFFIC, NONCRASH EVENTS. (b) Evaluation of each network which is a share water. (b) The age of each preserve indox and power windows whiches. (b) Evaluation of each network which is a share water. (b) The age of each preserve indivises, and collect data regarding, the numbers and types of injuries and deaths in nontraffic, nonerash events, involvin motor vehicles. (b) Included Information. The Secretary of Transportation shall collect and include in such deaths. Secretary of Transportation shall collect and include in such deaths. (c) The age of each notevehicles, in non-traffic, none-rate vents. (c) Whether dev sus an operator for the such vents. (c) Transportation traffic alternation of such notor vehicles. (d) Included Individual was and the vents. (c) evaluate technologies for deatecting and prever traindows or individual vasion of the such vents. (c) evaluate technologies for deatecting and prever traindows or prever windows and power windows and objects behind motor vehicles in non-traffic, none-rate events or individue share contraindow (c) statistica and notectics in non-traffic, none-rate events or individue share the such and events or require devices for detecting individual was deathed in such events. (c) evaluate the individual was and eve

Table 10: Examples of the BillSum dataset using the model billsum-4096-1024 base diverse

Bottom 5%	example (Sorted by rougeL)
Document	this study is an extension of a report on patients with type 1 diabetes at children 's hospital of new orleans (14) and was approved by the institutional review board at louisiana state university health sciences center , new orleans , louisiana glucose data were downloaded from patient meters at each clinic visit, meter model and sampling protocols varied by patient preference and insurance provider , an average of three glucose measurements per day were recorded in a study using a similar 4 minor monitoring protocols varied by patient preference and insurance provider , an average of three glucose measurements per day were recorded in a study using a similar 4 minor monitoring protocols varied by louise (15), at less measurements per day were recorded in a study using a similar 4 minor 20 patients collected at 16/2 clinic visit, matering and hybrid height by patient preference and calculate hemoglobin glycations (15) at the shifted at 16/2 clinic visit is described betwere (14), he same data were used to calculate hemoglobin glycation index (1g) and 10 ord/we patients in thory commercial bacharders in the one-, mortane - and ligh-hg is topposed on mean hybrid ta lob by into the regression equation. Just addece and light by grigorous data were calculated by subtrating predicted at le from observed at le mean eline visit. Tatients were divided in thory -, mortane - and light-hg is topposed on mean hybrid in the pregression equation (egg [mg/d] = (28.7 at (16 %)] 46.7, r = 0.027 (1). a mean blood glucose index (11) do 10.26, n = 68. (thight) 5.02.08, n = 67. (see each clinic visit, matering observed mig from ease. each clinic visit, matering observed m
Gold	objectivethe al c - derived average glucose study recommended reporting al c in estimated average glucose (eag) equivalents . we compared eag with self - monitored mean blood glucose (mbg) to determine whether eag is systematically biased due to biological variation in the relationship between mbg and al c-research design and methodsmbg and al ce were recorded from charts of 202 pediatric type I diabetic patients at 1612 clinic visits , raiterns were divided into group with low . moderate , or high al c bias based on a hemoglobin by givanion index (http://substithem eans at viaules for mbg versues eag were as follows : teal of population, 1942 H viau - higf or page 137 vs. 1632 on grd 137 vs. 1632
Model	objectiveto investigate the relationship between hemoglobin glycated hemoglobin (hba1c), hemoglobin glycated hemoglobin (hba1c) in children with type 1 diabetes research design and methodsalc was measured by national glycohemoglobin standardization program (ngsp)-approved immunoassays at the children's hospital of new orleans. hba1c was measured by national glycohemoglobin standardization program (ngsp)-approved immunoassays at the children's hospital of new orleans. hba1c was measured by national glycohemoglobin standardization program (ngsp)-approved immunoassays at the children's hospital of new orleans. hba1c was measured by national glycohemoglobin standardization program (ngsp)-approved immunoassays at the children's hospital of new orleans. hba1c was measured by national glycohemoglobin standardization program (ngsp)-approved immunoassays at the children's hospital of new orleans. hba1c was measured by national glycohemoglobin standardization program (ngsp)-approved immunoassays at the children's hospital of new orleans. hba1c was measured by national glycohemoglobin standardization program (ngsp)-approved immunoassays at the children's hospital of new orleans. ba1c was measured by national glycohemoglobin standardization program (ngsp)-approved immunoassays at the children's hospital of new orleans. Appulation regression equation [a1c (%) = [0.021 mbg (mg / d] + 4.3, r = 0.92) + 4.3, r = 0.92 (mg / d] + 4.3, r = 0.92 (mg / d] + 4.3, r = 0.92) has derived using mean hba1c (%) = [0.021 mbg (mg / d] + 4.3, r = 0.92) + 4.3, r = 0.92 (mg / d] + 4.3, r =
Metrics	Rouge1: 19.4, Rouge2: 6.06, RougeL: 10.03, RougeLsum: 13.38, Summary length (tokens): 256
	example (Sorted by rougeL)
Document	the principal aim of this laboratory is the synthesis of conjugated unsaturated keenes as candidate antineoplastic agars, these compounds interact with thiols but in general, they have little or no affinis for animo and hydroxyl groups which are found in models and the synthesis of conjugated unsaturated keenes as candidate antineoplastic agars, these compounds interact with thiols but in general, they have little or no affinis for animo and hydroxyl groups which are found in models are they as not have the generation is provide to a subsequent cytotoxic effect than various non - malignant cells (5, 6), hence by mounting the 1.5 diaryl-3-oxo-1.4 pertauting they provide the devolutent of 3.5 shiftsenzitient-0 perturbations and the devolute of a subsequent cytotoxic effect than various non- malignant cells (5, 6), hence by mounting the 1.5 diaryl-3-oxo-1.4 pertauting they periods in a devolute with discussional periods the devolutent of 3.5 shiftsenzitient-0 periods are also as a single subsequent cytotoxic effects are also as a close of a sequential periods and e cent - lymphocyles as well as municital [10] tymphocyle kakemine cells (7, 8), the hypothesis of sequential cytotoxicity was advanced than the 1.5 diaryl-3-oxo-1.4 periodicelly generations is of too the devolutent of 3.5 shiftsenzition-0.4 periodicelly generations is a sequential instruction, and the sequence structure constraints in order too analyses in a devolution of the sequence structure constraints in the sequence structure constraints and an antice structure structure structure structure structure
Gold	1-{4-(2-alkylaminocthoxy) phenykatoonyl3-3-bis(aryliden-)-piperidones are a novel class of potent cytotoxic agents. hese compounds demonstrate low micromolar to submicromolar ic50 values against human molt 4c8 and cent 1 lymphocytes and murine leakemia 11210 cells. in hits study, a comparative gast imersigation was performed on a series of 3.5-bis (aryliden-)-piperidones using different chemometrix tools to develop the best predictive models for further development of analogs with improved cytoxicity. all the gart models were studiaded by internal validation tests. In degar models over aluaded to histendia validation tests. In degar models over aluaded to h
Model	in order to evaluate the physicochemical properties of the 3,5-bis(benzylidene)-4-piperidone derivatives 1 - 7 and 2 - 7 were synthesized. the compounds 1 - 7 showed potent cytotoxic property with ic50 values of 3 - 6. the compounds 1 - 7 showed potent cytotoxic property with ic50 values of 3 - 6. the compounds 1 - 7 showed potent cytotoxic property with ic50 values of 3 - 6. the compounds 1 - 7 showed potent cytotoxic property with ic50 values of 3 - 6. the compounds 1 - 7 showed potent cytotoxic property with ic50 values of 3 - 6. the compounds 1 - 7 showed potent cytotoxic property with ic50 values of 3 - 6. the compounds 1 - 7 showed potent cytotoxic property with ic50 values of 3 - 6. the compounds 1 - 7 showed potent cytotoxic property with ic50 values of 3 - 6. the compounds 1 - 7 showed potent cytotoxic property with ic50 values of 3 - 6. the compounds 1 - 7 showed potent cytotoxic property with ic50 values of 3 - 6. the compounds 1 - 7 showed potent cytotoxic property with ic50 values of 3 - 6. the compounds 1 - 7 showed potent cytotoxic property with ic50 values of 3 - 6. the compounds 1 - 7 showed potent cytotoxic property with ic50 values of 3 - 6. the compounds 1 - 7 showed potent cytotoxic property with ic50 values of 3 - 6. the compounds 1 - 7 showed potent cytotoxic property with ic50 values of 3 - 6. the compounds 1 - 7 showed potent cytotoxic property with ic50 values of 3 - 6. the compounds 1 - 7 showed potent cytotoxic property with ic50 values of 3 - 6. the compounds 1 - 7 showed potent cytotoxic property with ic50 values of 3 - 6. the compounds 1 - 7 showed potent cytotoxic property with ic50 values of 3 - 6. the compounds 1 - 7 showed potent cytotoxic property with ic50 values of 3 - 6. the compounds 1 - 7 showed potent cytotoxic property with ic50 values of 3 - 6. the compounds 1 - 7 showed potent cytotoxic property with ic50 values of 3 - 6. the compounds 1 - 7 showed potent cytotoxic property with ic50 values of 3 - 6. the compounds 1 - 7 showed potent cytotoxic property with ic50 valu
Metrics	Rouge1: 21.38, Rouge2: 5.06, RougeL: 16.35, RougeLsum: 16.35, Summary length (tokens): 253
Top 75% ex	umple (Sorted by rougeL)
Document	educational evaluation (ee) is a formal process performed to evaluate the quality of effectiveness and/or value of a program. To process, good or curiculum. 12 it deals with data collection and assessment of the programs of the programs. To they considering more principles related to or clucational measuress and/or value of a program. To diverse performant on the collection or enary results in a better understanding of such programs. To diverse person do assessment of the programs of the programs of the programs of the programs. To diverse the collection and assessments and data collection, ee may results in a better understanding of such programs. To diverse the collection and assessment of the programs of the programs of the collection and assessment of the programs of the programs. To diverse the coll and the program calculation and and erection and the collection and assessment of the programs of the programs of the collection and assessment of the programs of the programs of the collection and prove the second provide a scientific provide
Gold	background : educational evaluation is a process which deals with data collection and assessment of academic activities progress . in this research , educational evaluation of detristry school of tehran university of medical sciences , which trains students in undergraduate and residency courses , was studied. methodschris descriptive study was done with a model of educational evaluation in ten steps and 13 fields including purposes and mission objectives , management an organization , academic board methores , students , human resources and support , educational , research , health and treatment paces , educational , dispositi , research and laboratory tools , chalt and treatment programs and courses , process of teaching and learning , evaluation and assessment , alumni , and parinets satisfaction , data were collected using observation , interviews , questionnaires , and checkitss results evaluation and the research (alumitor). Health and treatment mainly qualitative reset, results of some fields such as treatment quality of patients and elaction of the study were missions , detactional and research fields with size which is relatively destine. In the case of chaltering of the study metal mainly qualitative results of chaltering of chaltering of the study were missions , detactional and research fields with its relatively destines and elaction of the study were missions , detactional and research fields with size study evaluation of the study were missions , detactional and research fields were identified as the weeks retractive and paint one serioos attention.
Model	background : educational evaluation (ee) is a formal process performed to evaluate the quality of effectiveness and/or value of a program, process, goal or curriculum, it deals with data collection and assessment of the progress of academic programs. the objective of this descriptive cross - sectional study was to assess the quality of education provided at dental school of tehran university of medical sciences. materials and methods : in this descriptive cross - sectional study, the efficacy of provided educational programs provided at dental school of tehran university of medical sciences educational groups in 2010 through this descriptive cross - sectional study was assessed in 13 fields. the data were analyzed using spss version 20 results: the overall quality of education may the quality of education the most important quality of education as the quality of education. the most important quality of education, the most important quality of education. the most important quality of education, the most important quality of education was the quality of education may the quality of education. the most important quality of education was the quality of education was the quality of education. the most important quality of education was the quality of education was the quality of education. the most important quality of education was the quality of education was the quality of education. the most important quality of education was the quality of education was the quality of education.
Metrics	Rouge1: 39.32, Rouge2: 12.68, RougeL: 27.67, RougeLsum: 34.47, Summary length (tokens): 256
	umple (Sorted by rougeL)
Document	the staty population included patients over 18 years old who had an initial cabp or combined calp and open chest andic valve replacement, even $(n, 1)$, from april 1. 1998 to exclose 31. 2011 in ontario- the date of first catilities units cabe of combined calp and open chest andic valve replacement, even $(n, 1)$, form april 1. 1998 to exclose 31. 2011 in ontario- the date of first catilities units cabe of combined or patients of the respect to major outcomes, and 5 years for normality . representive data were included for fract states are excluded. cardiac care network of ontarios (ccn) data were used to dentify baseline characteristics such as catilica ejection fraction mumber of graft by based, prior more cased in states (and in a dentify the respect to major outcomes, and 5 years for normality). representive data were included for first com such distates used as catilica ejection fraction mumber of graft by based, prior more cased in an dentify the states were excluded as were excluded as were excluded in the state of the state information (clh) (disharge abstract database, national ambelation case network of the state states were calibated as were calibated as exclused in the respect to major constraints were have based on the state state exclusion to a state state exclusion and the state state exclusion of the state providents. Junit state (July
Gold	backgroundhe obesity paradox reflects an observed relationship between obesity and decreased morbidity and mortality , suggesting improved health outcomes for obese individuals , studies examining the relationship between high body mass index (bmi) and adverse outcomes after cardiac surgery have reported conflicting results methods and resultsthe study population (n re78762) was comprised of adult parients who had undergone first - time coronary attery bypass cardialy coro tarbies day / active value requestions day / sort value (1) in outroi () data from the institute for clinical evaluative sciences). perioperative outcomes and <i>J</i> -yaar totality among Prio- efficience (1) in outroi () data from the institute for clinical evaluative sciences). perioperative outcomes and <i>J</i> -yaar totality among Prio- efficience on the relationship between bin and adverse outcomes. The evaluation sciences). perioperative outcomes and <i>J</i> -yaar totality among Prio- efficience on the relationship between bin and adverse outcomes. The event of the adverse outcomes and evert or solid adverse outcomes and evalue to a solid obsect and a solid and morbid () obsect and and evalue (1) (9.80 to 902.) (9.766 (9.9% ci. 730 to 7.81), and 9.006 (9.9% ci. 80.51 negotive) and obsect merel and and particulate and preves and there are priorparitive contonality and indiverse conclusions. Soverweight and obsect prior mere priorparitive contones there endines and and are reasoned and evaluated and excellence and and evaluated and morbid () obsect and and particulate and and solid () and prior particulate and and solid () and and and morbid () and solid () and
Model	backgroundthe aim of this study was to investigate the relationship between body mass index (bmi) and mortality in patients undergoing coronary artery bypass grafting (cabg) or open chest aortic valve replacement (avr), methods and results we identified 1 year patients undergoing primary cabg or combined cabg and open chest aortic valve replacement (avr) from april 1, 1998 to october 31, 2011. patients were idvided into groups : underweight (bmi < 20 kg / m 2), norma weight (bmi 25.0 to 29.9 kg / m 2), overweight (bmi 25.0 to 29.9 kg / m 2), obsee (bmi 30.0 to 29.9 kg / m 2), and morbidly obsee (bmi > 34.9 kg / m 2) patients were followed for 1 year with respect to major outcomes, and 5 years for mortality. cox proportional hazards regression analysis was used to investigate multiple confounders on the relationship between bmi and mortality, providing hazard ratios and 95% confidence intervals (ci), patients with bmi < 20 kg / m 2 mer more likely to die during the 5 years of follow - up (hazard ratio [hr]
Metrics	Rouge1: 56.82, Rouge2: 26.29, RougeL: 40.91, RougeLsum: 50.0, Summary length (tokens): 256

Table 11: Examples of the PubMed dataset using the model pubmed-4096-512 small diverse

	example (Sorted by rougeL)
Document	in august, 4 months before presentation, a 35-year-old while woman of scots and enginh descent developed redisiburine for several days. Bolloweld by eruption of vesicles and biaters on the dorsal surfaces of her hands and fingers. Its sides of her nose, and ther upper atterior best, haves, and legs, she worked as a mindex-gaine contraint and noticed that estimations occurred on auces scoped of sounight, but any patiention of surfaces of her hands and fingers. Its sides of her nose, and her upper atterior best, haves, and and hands electronic and notice day has based dowly, of the wink having duringed from tobaccurred or allowed the patients of house flows of the start with here are have here hereds. She consumed there glasses of wine rady week and handsel electronic cigareties for approximately 6 months, having duringed from tobaccurred cigareties. She patientic cigareties for approximately 6 months, having duringed from tobaccurred cigareties. She patient here are with here are with the hereds. She consuming the patients and handsel electronic cigareties for approximately 6 months, having duringed from tobaccurred cigareties. She patients have here and affering her and notice and startisticat descripted above. Lesions were most prominent on the doral surfaces of the hands and finger (fing. milia were scatcurred with biallow stile contraints and there were scatcured with a startistic continue there are solved with and there (finger (finger) there were vesicles, hulls with erythemators haves, some a large sh l cri in dimatory and deler biolons in various stages of erosion resolution, and scattured with advecurred with end there with erythemators haves, some a large sh l cri in dimatory and deler biolons in various stages of erosion resolution and scatture and a larger (finger) there were vesicles, hulls with erythemators haves, older lesions in various stages of erosion resolution, scarring and puncture milia, all hypical of pert. (h) after tori depletion with philosophilosony, there were occites, hull
Gold	pephysis catanas tuda (pc) is characterized by docensed supportphysingen docaboxybase activity in hepatoxytase, supportphysini and hepatoxanboxyl pephysini and successativity in securated and a substance to the product of the produ
Model	a 35-year - old white woman of scots and english descent developed reddish urine for several days followed by eruption of vesicles and blisters on the dorsa surfaces of her hands and fingers, the sides of her nose, and her upper anterior chest, knees, and legs. skin biopsy specimens revealed linear glassy iga and igg deposition along the epidermal basement membrane zone and superficial dermal blood vessels, forming a doughnut pattern. the patient was treated with levonorgestrel - releasing intrauterine contraceptive system (skyla). eight months after presentation, she underwent implantation of a levonorgestrel - releasing intrauterine contraceptive system (skyla). eight months after presentation, she underwent implantation of a levonorgestrel - releasing intrauterine contraceptive system (skyla).
Metrics	Rouge1: 20.56, Rouge2: 0.94, RougeL: 11.22, RougeLsum: 17.76, Summary length (tokens): 176
Bottom 25%	example (Sorted by rougeL)
Document	this was a multicenter, community - hased, netrospective observational study of patients with pass, ranging from 8 to 20 mm in dumeter, proving to 18 geographically representative compatient pulmonary clinics across the united states the study was approved 11 5 lists by carring interview how rad and at the study was approved 11 5 lists by carring interview how rad and the study was approved 11 5 lists by carring interview how rad and the study was approved 11 5 lists by carring interview how rad and the study was approved 11 5 lists by carring interview how rad and the study was approved 11 5 lists by carring interview how rad and the study was approved 11 5 lists by carring interview how rad and the study was approved 11 5 lists by carring interview how rad and the study was approved 11 5 lists by carring interview how rad and the study and provide 1 lists underging undification review even bene on participation based on the following criteria (1) management of patients with participations (2) availability of perform data abstraction, investigators targeted enrollment of geographically drivere patients to limit the potential bias associated with differences in practice patterns and to account for variation in disease prevalence (eq. hilling and scheduling systems) using five who met the criteria to minimize selection bias, this site were approximated costs form (2) availability of lists. Underging the study system is listing the display is not indicated accost, and the study energy to display systema is single (1) substance instance is deviated bias approximate and the site of the study and approximate study of the study systema is single (1) substance is application of diseases. This is the study and application of diseases is the study application of diseases is the study and application of diseases is the study and application of diseases is the study and application of diseases applica
Gold	background: pulmonary nodules (pas) are a common reason for referral to pulmonologists . the majority of data for the evaluation and management of pas is derived from studies performed in academic medical centers. Inits is known about the prevalence and diagnosis of pas, the use of diagnostic testing , or the management of pas by community pulmonologists . mitmoship size resolution of passing and the prevalence of malignary prevalence of malignary prevalence of malignary prevents of the management of passing and the prevalence of malignary prevalence o
Model	objectivethe aim of this study was to document the prevalence of cancer (25%) in patients with intermediate - sized pulmonary nodules referred to community pulmonologists.materials and methodsthis was a multicenter, community - based, observational study of patients with intermediate - sized pulmonary nodule who were referred to pulmonologists. patients were categorized into three groups by the most invasive procedure performed during management, as follows surveillance (serial imaging), biopsy (ct - guided transthoracic needle aspiration [tma] or bronchoscopy), or surgery (including mediastinoscopy, video assisted thorascopic surgery, and/or thoracotomy), patients were categorized into three groups by the most invasive procedure performed during management, as follows : surveillance (serial imaging), biopsy (ct - guided transthoracic needle aspiration [tma] or bronchoscopy), or surgery (including mediastinoscopy, video - assisted thorascopic surgery, and/or thoracotomy), patients were categorized into three groups by the most invasive procedure performed during management, as follows : surveillance (serial imaging), biopsy (ct - guided transthoracic needle aspiration [tma] or bronchoscopy), or surgery (including mediastinoscopy, video - assisted thorascopic surgery, and/or thoracotomy). multivariate logistic regression was performed to identify factors associated with the use of an invasive diagnostic procedure, resultsof the 377 patients included, 283 (75%) had a nodule that was benign, and 94 (25%) had a malignant nodule. the overall accuracy of pet scanning was 74%, with a false - positive (fn) rate of 9%, nodules measuring > 11 to 15 mm (n = 48) had fin and fp rates of 9% and 36%, respectively.conclusionsthe prevalence of cancer in patients with intermediate - sized nodules was 25%, the rate of surgical resection for benign disease varied from 9% to 23% in screening trials and surgical series.
Metrics	Rouge1: 45.58, Rouge2: 9.56, RougeL: 18.37, RougeLsum: 38.1, Summary length (tokens): 470
Top 75% ex	ample (Sorted by rougeL)
Document	a total of (217 deal birds were shipped at 6e to the tropical medicine institute "pedro board" and identified by entitled by geness, brain, and kidneys were ensured and tested for wm by using reserve transcriptione polymerase chain encoder (1 - geness) (217, 1 - geness), we extracted by using the quarty wirt and kidney (quere, inc., values), "penetroms vn221 (5 + typicgregacetacine), where were lost 0 detect min man, a second r - per with primers wn0483 (5 - second-agoccase), and wn0794 (5 - geness), and the same man preparation. serum specimens from horses in havana and havana province were tested for autibulice to wm by using a concertage constance. The same start (1 - geness) were set to 0 detect min at the same man preparation. serum specimens from horses in havana and havana province were tested for autibulice to wm by using a concertage contradacettrates say (eilar) with monocloal antibulice to a same start (1 - geness) were set to 0 detect min at the same man preparation. serum specimens from horses in havana and havana province were used to detect min at the same say (eilar) with minocloal antibulice to a same start (1 - geness) were say (eilar) with minocloal antibulice to a same say (eilar) with minocloal antibulice to a same start (1 - geness) were same say (eilar) with minocloal antibulice to a same say (eilar) with minocloal antibulice to the same say (eilar) with minocloal antibulice to a same same say (eilar) with minocloal antibulice to same say (eilar) with minocloal antibulic
Gold	a surveillance system to detect west nile virus (ww) was established in cuba in 2002, wav infection was confirmed by serologic assays in 4 asymptomatic horses and 3 humans with encephalitis in 2003 and 2004. these results are the first reported evidence of www activity in cuba.
Model	we report the first evidence of antibodies to west nile virus (wnv) in horses and humans in cuba. antibodies were detected by plaque reduction neutralization test in 10 of 1,217 dead birds tested, these findings provide evidence that wnv and slev may co - circulate in cuba.
Metrics	Rouge1: 51.06, Rouge2: 17.39, RougeL: 31.91, RougeLsum: 46.81, Summary length (tokens): 64
-	ample (Sorted by rougeL)
Document	intra articular injections of cordicoteroids have been used for several decades in the management of inflammatory and degenerative joint conditions when first – line conservative therapies such as rest, ice, and anti - inflammatory medications fail to provide adoptate symptome relicf. Jussel in part on this log history of accessful attribution coupled with the findings of several radionized controlled trips. consensus attenties and media - analyses have concluded the inflammatory medications fail to acceptate as an effective treatment for the constant of the inflammatory and degenerative joints when here the constructives the verse observability. In the inflammatory and degenerative joints who have been conservative the terms for the constant fails and the inflammatory and degenerative joints who have been conservative the approximation of the part in the pain associated with observability to constructivity in nonpharmacologic therapy and simple analgesis e.g. acetaminophen. Traditionally, intra - articular injections have been performed using antonical landmarks to identify the correct major gravity of needle placement. Jowever, afferent annotical- gravitation in the provide adaptive structure and the accession that articular injections in the application and directly valuation due area of interest. Jowever, and variations in analow are common sepsis, and structure articular articular articular injections. In the first the test possible model have a structure and variations in tanking model have the accession that articular injections. In the first the specifications constructure and and articular articular injections. The accession of the specifications constructure and and articular articular injections are accessible model. The accessible the specifications constructure and the applications articular injections are accessible model have and the specification constructure and the provemation articular injections are accessible model have and the accessible model have accessible and thave the specification an
Gold	intra - articular corticosteroid and hyduronic acid injections provide short - term symptom amelioration for arthritic conditions involving structural damage or degenerative changes in the knee . conventional pulpation - guided anatomica injections frequently result in inaccurate needle placement into extra - articular tissue and adjacent structures. the purpose of this review was to determine the effect of ultrasonal guidance on the accuracy of needle placement, clinica oncorners , and cost - effectiveness in comparison with anatomical landmark - guided intra - articular lang jointi njecicions with particular emphasis on the lanes. at solid of 11 placent studies were to define a free structure in the gradient of the lanes. The effectiveness in comparison with anatomical landmark - guided intra - articular lang jointi njecicions with particular emphasis on the lanes. at solid of 11 placent studies were to define a free structure in the lanes in the lane in the lanes. The effectiveness in comparison with anatomical landmark - guided intra - articular lang jointi njecicions with the structure in the lange of the lange intra - guides with advaced (95% versus 73% p. e COOI) and the lange structure in the data extra of the advaced (95% versus 73% p. e COOI) and the lange structure in the lange entra - articular joint space of lange - intra - articular joint space of lange - intra - articular joint space of lange - provide lange and housider (93% versus 73% p. e COOI) and the lange structure intra - articular joint space of lange - joint including the knee . the enhanced integration accuracy achieves in the lange intra - articular joint space of lange joint including the knee . the enhanced integration accuracy in the lange intra - articular joint space of lange joint including the knee . the enhanced integration accuracy achieves in the lange intra - articular joint space of lange joint including the knee . the enhanced integration accuracy in the lange intra - articular joint space of lange joint including t
Model	objectivethe purpose of this review was to determine the effect of ultrasound guidance on the accuracy of needle placement, clinical outcomes, and cost effectiveness in comparison with anatomical landmark - guided intra - articular large joint injections, with particular emphasis on the knee.methodsthe peer reviewed literature was searched using medline and relevant bibliographies published in english language journal strong the output output documents of 1, 2011 that compared the accuracy of intra - articular injections, with particular emphasis on the knee.results: total of 13 studies met the entry criteria. needle placement accuracy ranged from 63% to 100% with ultrasound and from 39% to 1000 with conventiona anatomical guidance. imaging guidance improved the accuracy of intra - articular injections of the knee (96.7% versus 81.0%, $p < 0.001$) and shoulder (97.3% versus 65.4%, $p < 0.001$). imaging guidance improved the accuracy of intra - articular injections of the knee (96.7% versus 81.0%, $p < 0.001$) and shoulder (97.3% versus 65.4%, $p < 0.001$). in particular, ultrasound guidance of knee injections resulted in better accuracy than did anatomical guidance (95.8% versus 81.0%, $p < 0.001$). in particular, ultrasound guidance of knee injections resulted in better accuracy than did anatomical guidance (95.8% versus 81.0%, $p < 0.001$). in particular, ultrasound guidance of knee injections resulted in better accuracy than did anatomical guidance (95.8% versus 77.8%, $p < 0.001$). in particular, ultrasound guidance of knee injections resulted in better accuracy than did anatomical guidance (95.8% versus 77.8%, $p < 0.001$). in particular, ultrasound guidance of knee injections resulted in better accuracy than did anatomical guidance (95.8% versus 77.8%, $p < 0.001$). in particular, ultrasound guidance of knee injections resulted in better accuracy than did anatomical guidance (95.8% versus 77.8%, $p < 0.001$). in particular ultrasound guidance of knee injections resulted in better accuracy than did anatom

Table 12: Examples of the PubMed dataset using the model pubmed-4096-512 base diverse

L Data Details

We used our own tokenizer to calculate the number of tokens. In Tables 6, and 7 we show the data length distributions for the BillSum train and test splits. In Tables 8, 9, and 10 we show the data length distributions for the PubMed train, validation and test splits.



Figure 6: Histograms for the BillSum training set (18949 samples).



Figure 7: Histograms for the BillSum test set (3269 samples).



75-Quant: 265, 95-Quant: 329, Max: 506

Figure 10: Histograms for the PubMed test set (6658 samples).

75-Quant: 3964, 95-Quant: 6985, Max: 48750