# Relation-Guided Pre-Training for Open-Domain Question Answering 

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#### Abstract

Answering complex open-domain questions requires understanding the latent relations between involving entities. However, we found that the existing QA datasets are extremely imbalanced in some types of relations, which hurts the generalization performance over questions with long-tail relations. To remedy this problem, in this paper, we propose a Relation-Guided Pre-Training (RGPT-QA) framework ${ }^{1}$. We first generate a relational QA dataset covering a wide range of relations from both the Wikidata triplets and Wikipedia hyperlinks. We then pre-train a QA model to infer the latent relations from the question, and then conduct extractive QA to get the target answer entity. We demonstrate that by pretraining with propoed RGPT-QA techique, the popular open-domain QA model, Dense Passage Retriever (DPR), achieves $2.2 \%, 2.4 \%$, and $6.3 \%$ absolute improvement in Exact Match accuracy on Natural Questions, TriviaQA, and WebQuestions. Particularly, we show that RGPT-QA improves significantly on questions with long-tail relations.


## 1 Introduction

Open domain question answering is a challenging task that answers factoid questions based on evidence in a large corpus (e.g., Wikipedia). Most open-domain QA systems follow retriever-reader pipeline (Chen et al., 2017), in which a retriever selects a subset of candidate entities and associated passages from the corpus that might contain the answer, then a reader extracts a text span from the passages as the answer. This process involves multiple entities that are relevant to answer the question. The QA system is required to extract these entities from the question and passages and identify the (latent) semantic relations between these entities in order to answer the question. For example, to

[^0]answer the following question: "Where did Steph Curry play college basketball at?", the QA model is required to reason the implicit relation triplet〈Steph Curry, Educated At, Davidson College〉 to identify the correct answer.

To capture the relation knowledge required to answer questions, most QA systems rely on humanannotated supervised QA datasets. However, it is expensive and tedious to annotate a large set of QA pairs that cover enough relational facts for training a strong QA model. In addition, we showed that even for a large QA dataset like Natural Questions (Kwiatkowski et al., 2019), its training set only covers $16.4 \%$ of relations in WikiData (Vrandecic and Krötzsch, 2014) knowledge graph. Moreover, for those covered relations, the frequency distribution is imbalanced, i.e., $30 \%$ of relation types appear only once. Consequently, for the questions involving infrequent (a.k.a, long-tail) relations in the training set, the QA exact match accuracy is $22.4 \%$ lower than average. Such a biased relation distribution of existing QA datasets severely hurts the generalization of trained QA systems.

To improve the open-domain QA systems for questions with long-tail relations, in this paper, we propose RGPT-QA, a simple yet effective RelationGuided Pre-training framework for training QA models with augmented relationa facts from knowledge graph. The framework consists of two steps: 1) generate a relational QA dataset that covers a wide range of relations without human labeling; 2) pre-train a QA model to predict latent relations from questions and conduct extractive QA.

The key of our framework is to generate a relational QA dataset that align entities in Wikipedia passages with structured knowledge graph (e.g., WikiData). We call such a dataset Grounded Relational Wiki-Graph. In this graph, each edge indicates the relationship of two connected entities, and the edge is linked to a passage in Wikipedia describing this relationship. As WikiData knowledge


Figure 1: Cumulative distribution function (CDF) of relation frequency in Natural Question Training set.
graph also suffers from low coverage of long-tail entities and relations, we further convert hyperlinks in Wikipedia into knowledge triplets without specifying relation labels. Next, we link each relation triplet to a Wikipedia passage to help generate natural questions. We assume that if one passage in the Wiki-page of source entity contains the target entity, then the context in this passage describes the relationship between the two entities. With the constructed graph, we use a template to synthesize question and answer pairs and then pre-train the QA model to capture the relational facts for answering complex open-domain questions.

As a pre-training method, RGPT-QA can be incorporated with any open-domain QA system. In this paper, we utilize the recently developed Dense Passage Retriever (DPR) (Karpukhin et al., 2020) as the base QA system to evaluate the proposed pretraining effectiveness. Experimental results show that RGPT-QA enhances DPR's Exact Match accuracy by $2.2 \%, 2.4 \%$, and $6.3 \%$ on Natural Questions, TriviaQA and WebQuestions respectively. Compared with the existing QA pre-training methods (Lee et al., 2019; Guu et al., 2020a; Lewis et al., 2019), RGPT-QA explicitly captures a wide range of relational facts and thus achieves better performance. Moreover, for the questions containing long-tail relations in Natural Questions, the performance is improved by $10.9 \%$, showing that RGPT-QA alleviates the unbalanced relation distribution problem in the existing QA datasets.

The key contributions of this paper are:

- We propose RGPT-QA, a pre-training method to inject knowledge from relational facts in knowledge graph into QA models.
- RGPT-QA enhances the performance of a popular QA model, i.e., DPR, especially on the questions with long-tail relations.


Figure 2: Exact Match accuracy of a trained DPR model in validation set with different relation frequency in training set.

## 2 Preliminary and Empirical Analysis

In this section, we firstly introduce the retrieverreader pipeline for open-domain QA , and then we analyze how the relation distribution in existing QA datsets influence generalization performance.

Open-Domain Question Answering. We focus on open-domain question answering that requires to extract answer from a large corpus (e.g. Wikipedia) $\mathbb{C}=\left\{p_{i}\right\}_{i=1}^{N}$ containing $N$ passages. Most open-domain QA systems follow a retrieverreader pipeline proposed by Chen et al. (2017). Given a factoid question $q$, the QA system first retrieves $K$ relevant passages $\left\{p_{j}\right\}_{j=1}^{K}$ from the corpus $\mathbb{C}$. Then a reading comprehension module extracts a text span $w_{\text {start }}, \ldots, w_{\text {end }}$ from one of these retrieved passages as the answer $a$ to the question. Some QA dataset annotated the passage where the answer $a$ is derived. We called this passage ground truth passage.

For the retriever, earlier systems utilize termbased retrieval methods, such as TF-IDF and BM25, which fails to capture the semantic relationship between question and passage beyond lexical matching. Recent studies (Lee et al., 2019; Karpukhin et al., 2020; Dhingra et al., 2020) use BERT-like pretrained language model to encode the question and passages independently into dense representations, and use maximum inner product search (MIPS) algorithms (Shrivastava and Li, 2014) to efficiently retrieve the most similar passage for each question. In this paper, we utilize Dense Passage Retriever (DPR) (Karpukhin et al., 2020) as the base QA model.

Relation Bias of Existing QA Datasets. We first explore how much relational knowledge between entities is required to answer the questions in the existing open-domain QA dataset. We con-
duct an empirical study to analyze the relation distribution in Natural Questions, one of the largest open-domain QA datasets, and how it influences QA model's performance.

For each question in Natural Question training set, we first select the entity that the ground-truth passage is associated with. We then combine the entity with the answer as an entity pair, and check whether we can find a relation triplet in WikiData describing the relation between these two entities. Out of 58,880 training QA pairs, there are 23,499 pairs that could be aligned. The aligned QA pairs cover 329 relations, which accounts for $16.4 \%$ of the total 2,008 relations in WikiData. For most unaligned QA pairs, the answers are not entities and thus cannot be aligned to the graph.

In addition to the low relation coverage issue in Natural Question, we also find that the relation distribution is imbalanced. As showed in Figure 1, $90 \%$ of relations have frequency less than 41 , and $30 \%$ of relations appear only once. On the contrary, the most frequent relation "P161 (cast member)" appears 1,915 times out of 9,238 aligned QA pairs. A complete list of all these relations with aligned QA pairs is shown in Table 6-9 in Appendix.

We then study whether the imbalanced relation distribution influences the performance of QA models trained on these datasets. We use a DPR model trained on training set of Natural Questions and then calculate the Exact Match accuracy in validation set of each aligned QA pairs. We then analyze the correlation of the accuracy with the relation frequency in training set. As illustrated in Figure 2, the validation set accuracy is overall proportional to the relation frequency in training set. For those relations with frequency less than 5 , the average accuracy is only $20.3 \%$, much lower than the average accuracy $42.7 \%$ over all samples in validation set. This shows that the relation bias in existing QA datasets severely influences the generalization of QA models to questions with long-tail relations.

## 3 Method

In this section, we will discuss RGPT-QA framework in: 1) how to generate relational QA dataset for the pre-training purpose; and 2) how to construct a self-training task to empower QA model to capture relational facts.

| \# of linked Entity | $5,640,366$ |
| :--- | :---: |
| \# of relation labels | 2,008 |
| \# of labelled triplet | $14,463,728$ |
| \# of unlabeled triplet (hyperlink) | $66,796,110$ |
| \# of grounded descriptions per triplet | 1.25 |

Table 1: Statistics of Grounded Relational Wiki-Graph.

### 3.1 Construct QA Pre-Training Dataset

To help QA model capture the knowledge from relation facts required to answer open-domain questions, we first focus on generating QA pre-training dataset, in which there exist relation connections between the source entity in questions to the target answer. Specifically, each QA pair datapoint $d=\left\langle\langle s, r, t\rangle, q, p^{+}\right\rangle$consists of three components: 1) relational triplet $\langle s, r, t\rangle$, in which $r$ denotes the relation between source entity $s$ and target entity $t$;2) question $q$ in natural language asking which entity has relation $r$ to source entity s , with target entity $t$ as the correct answer; 3) positive context passage $p^{+} \in \mathbb{C}[s]$, a passage from source entity's Wiki-page that contains the target answer $t$.

Grounded Relational Wiki-Graph. To generate QA pre-training dataset, leveraging the relation triplets in knowledge graph, e.g., WikiData, is a natural choice to define questions that require relation reasoning. We therefore construct Grounded Relational Wiki-Graph, in which each relation triplet $\langle s, r, t\rangle$ is linked to a set of description passages $\{d e s c .(s, t)\}$ in the Wiki-page of entity $s$. These descriptions would be later utilized to generate questions $q$ and positive context passages $p^{+}$.

To construct such a graph, we use the 2021 Jan. English dump of Wikidata and Wikipedia. For each Wikipedia hyperlink $\langle s, ?, t\rangle$ (? denotes the relation is unlabeled), the passage containing anchored text to $t$ in the Wiki-page of $s$ naturally fits our requirement for desc. $(s, t)$. For each WikiData relation triplet $\langle s, r, t\rangle$, if the two entities are linked by a hyperlink in Wikipedia, we label the relation of the aligned hyperlink as $r$. For the other triplets $\langle s, r, t\rangle$ without alignment with hyperlinks, we extract all mentioning of target entity $t$ from the Wiki-page of $s$, and use the context passage as desc. $(s, t)$. The dataset statistics are shown in Table 1.

Relational QA Pair Generation In the following, we introduce the details to generate the relational QA pair from the constructed graph.

Recent unsupervised QA studies (Li et al., 2020; Pan et al., 2020) revealed that if the question $q$ and


Relation Triplet: <Stephen Curry, ?, Splash Brothers>
Question: [<MASK(r)>] of [Stephen Curry] which [the <MASK( t$)>$ which are a duo of American basketball players consisting of Stephen Curry and Klay Thompson].
True Passage: ...<MASK(s)> and teammate Klay Thompson have earned the nickname of the Splash Brothers... Answer (Target Entity): Splash Brothers
Self-Distilled Relation Label: nickname ( $p=0.18$ ), member of team ( $P=0.11$ ), position on a team ( $P=0.09$ )......

Figure 3: Example of a generated relational QA pair from Grounded Relational Wiki-Graph.
context passage $p^{+}$share a large lexical overlap, then the QA model could utilize low-level lexical patterns as shortcuts to find the answer. These shortcuts hinder the model from learning to comprehend the passages and answer the questions, hurting model's generalizability. To avoid this lexical overlap issue, we aim to generate questions from a passage that is different from the context passage $p^{+}$.

We first select all the entity pairs $\langle s, t\rangle$ that have mutual links in the Grounded Relational WikiGraph, with $\operatorname{desc} .(s, t)$ and $\operatorname{desc}(t, s)$ in part of Wikipage of $s$ and $t$ respectively, describing the relationship between the two entities. Without loss of generality, we denote $s$ as source entity and $t$ as the target answer. The passage desc. $(s, t)$ containing target answer $t$ can be used as the positive passage $p^{+}$.

Next, we generate a question that is lexically different from $p^{+}$using the following template:

$$
q(s, r, t)=[\operatorname{MASK}(r)] \text { of }[s] \text { which }[\operatorname{desc} .(t, s)] ?
$$

in which MASK $(r)$ is a relation mask token. As desc. $(t, s)$ contains source entity $s$, it provides information to describe the relationship between $s$ and $t$, based on which the QA model should learn to infer the latent relation $r$, and retrieve positive passage $p^{+}=\operatorname{desc} .(s, t)$ and extract answer entity $t$. In addition, as desc. $(t, s)$ and desc. $(s, t)$ come from different Wiki-page, our question generation
procedure can avoid the lexical overlap issue that often occur in prior Unsupervised QA methods.

Mask Target Answer. As description desc. $(t, s)$ is from target answer $t$ 's wiki-page, it often contains the name of entity $t$. We thus need to mask $t$ from the question. Otherwise, the pre-trained model can simply identify the answer to a question based on the local patterns.

As an example, in Figure 3, we show how to generate question for triplet〈Stephen Curry, ?, Splash Brothers〉. We firstly retrieve two descriptive passages desc. $(s, t)$ and desc. $(t, s)$ in two entities' wiki pages. Using the template, we generate the question along with the ground-truth passage. We then mask out the target entity in question and source entity in true passage (will discuss later in retrieval pre-training) to avoid shortcut. A list of generated relational QA pairs are shown in Table 10 in Appendix.

### 3.2 Relation-Guided QA Pre-Training

With the generated relational QA dataset, we introduce how to pre-train both retriever and reader components in the QA model.

### 3.2.1 Relation Prediction Pre-Training

Our generated QA dataset contains the relation label $r$ between the source entity $s$ and the answer target $t$. Therefore, we design a self-training task to guide the model to predict the latent relation
in question, which can benefit both retriever and reader. Specifically, we adopt a linear projection layer $\mathrm{L}_{R}(\cdot)$ over the BERT $_{\text {[CLS] }}$ token embedding to predict the relation over the WikiData relation set. The pre-training loss of relation prediction is:

$$
\mathcal{L}_{\mathrm{rel}}=\frac{1}{B} \sum_{q}-\log P(r \mid q ; \theta),
$$

Self-Distillation for Unlabelled Relation The hyperlinks in wikipedia also provide valuable implicit information about the relations between entities. To leverage them, we use the trained relation predictor at each epoch with fixed parameter $\hat{\theta}$ as teacher model to assign soft label and then progressively train the relation predictor as student model based on the assigned labels in the next epoch. This approach is referred to as self-distillation in the literature (Xie et al., 2020; Chen et al., 2020). We minimize this self-distillation loss as:

$$
\mathcal{L}_{\text {distill }}=\frac{1}{B} \sum_{q} \sum_{\hat{r}}-\log P(\hat{r} \mid q ; \theta) \cdot \operatorname{sg}(P(\hat{r} \mid q ; \hat{\theta}))
$$

where $\operatorname{sg}(\cdot)$ denotes the operation of stop gradient, which avoids back propagation to the teacher network with fixed parameter $\hat{\theta}$. $\hat{r}$ is enumerating all the relation labels.

As the relation predictor at early stages cannot give a reasonable prediction, we put a dynamic weight schedule to $\mathcal{L}_{\text {distill }}$ by a time-dependent weighting term $1-e^{- \text {epoch }}$, which ramps up from zero to one. Combing the weighted self-distillation $\operatorname{loss} \mathcal{L}_{\text {distill }}$ with the supervised relation loss $\mathcal{L}_{\text {rel }}$, we get the final relation loss $\hat{\mathcal{L}}_{\text {rel }}$ to train the model capturing all relational facts covered in the Grounded Relational Wiki-Graph.

### 3.2.2 Dense Retrieval Pre-Training

The goal of dense retrieval pre-training is to get a question encoder $\mathrm{Enc}_{Q}$ and a passage encoder $\operatorname{Enc}_{P}$ to map questions and all passages in the Wiki Corpus $\mathbb{C}$ into an embedding space, such that each question $q$ is close to its ground-truth positive context passage $p^{+}$in the embedding space. The objective is as follows:

$$
\begin{equation*}
P_{\text {retr }}\left(p^{+} \mid q, \mathbb{C}\right)=\frac{\exp \left(\operatorname{sim}\left(q, p^{+}\right)\right)}{\sum_{p \in \mathbb{C}} \exp (\operatorname{sim}(q, p))} \tag{1}
\end{equation*}
$$

where $\operatorname{sim}(q, p)$ is the cosine similarity between the normalized embeddings of question and passage.

Two-Level Negative Passage Sampling. As we cannot enumerate all other passages in the denominator of $\mathrm{Eq}(1)$, we need to sample a set of negative passages for contrastive learning. Previous studies (Karpukhin et al., 2020) have revealed that it is essential that the sampled negative passages should be hard enough to train the retriever. As the question and passage embeddings are encoded independently, DPR can efficiently calculate the similarity of each question to all passages in the batch via dot product. Based on this property, as long as the passages within a batch are similar to each other, they serve the hard cases of negative passages to others. We thus propose a two-level negative passage sampling strategy to construct hard cases for training the retriever in the following.

We first sample at the level of entity. Given a set of randomly sampled $b$ entities, we adopt random walk from these seed entities over the Grounded Relational Wiki-Graph to get $B$ entities. As the connected entities have a relationship, their true passages are also semantically similar, and thus serve as good negative samples. We then conduct sampling at the level of passage. For each source entity $s_{i}$ with positive passage $p_{i}^{+} \in \mathbb{C}\left[s_{i}\right]$, we randomly pick $K$ other passages from the same Wiki-page to form a negative passage set $\left\{p_{i, j}^{-} \in\right.$ $\mathbb{C}\left[s_{i}\right]$, s.t. $\left.p_{i, j}^{-} \neq p_{i}^{+}\right\}_{j=1}^{K}$. These negative passages are similar to $p_{i}^{+}$, as they all describe the same entity $s_{i}$.

After we collect both the positive and $K$ negative passages for all the entities, we use the passage encoder $\mathrm{Enc}_{P}$ to get a passage embedding matrix $\mathbf{P}$ with dimension $((1+K) \cdot B \times d)$. We also use question encoder $\operatorname{Enc}_{Q}$ to get question embedding matrix $\mathbf{Q}$ with dimension $(B \times d)$. We then get a similarity matrix $\mathbf{S}=\mathbf{Q} \mathbf{P}^{T}$ with dimension $(B \times(1+K) \cdot B)$, in which the diagonal entry corresponds to the similarity between question and its positive passage. We thus calculate the retrieval loss with in-batch negative samples via:

$$
\begin{equation*}
\mathcal{L}_{\text {retr }}=\frac{1}{B}\left(\sum_{i \in[1, B]}(-\log \operatorname{softmax}(\mathbf{S}))_{[i, i]}\right) . \tag{2}
\end{equation*}
$$

Masking Source Entity. As the true passage $p_{i}^{+}=$desc. $(s, t)$ might contain the name of source entity $s$. We mask out all the tokens of $s$ from the extracted passages, so that the model is required to understand the passages for correct retrieval instead of exploiting a shortcut.

### 3.2.3 Reading Comprehension Pre-Training

The goal of reading comprehension pre-training is to get a neural reader that re-ranks the top- $k$ retrieved passages and extracts an answer span from each passage as the answer. The probability of a passage contains the target answer $t$, and each token in the selected passage being the starting/ending positions of an $t$ are defined as:

$$
\begin{aligned}
& P_{\text {rank }}(t \in p)=\frac{\exp \left(\mathrm{L}_{\text {rank }}\left(\operatorname{BERT}_{\mathrm{CLS}}(q, p)\right)\right)}{\sum_{\hat{p}} \exp \left(\mathrm{~L}_{\text {rank }}\left(\operatorname{BERT}_{\mathrm{CLS}}(q, \hat{p})\right)\right)}, \\
& P_{\text {start }}(i \mid p, q)=\frac{\exp \left(\mathrm{L}_{\text {start }}\left(\operatorname{BERT}_{[i]}(q, p)\right)\right)}{\sum_{j} \exp \left(\mathrm{~L}_{\text {start }}\left(\operatorname{BERT}_{[j]}(q, p)\right)\right)}, \\
& P_{\text {end }}(i \mid p, q)=\frac{\exp \left(\mathrm{L}_{\text {end }}\left(\operatorname{BERT}_{[i]}(q, p)\right)\right)}{\sum_{j} \exp \left(\mathrm{~L}_{\text {end }}\left(\operatorname{BERT}_{[j]}(q, p)\right)\right)} .
\end{aligned}
$$

where $L_{*}$ are linear project layers with different parameters. Note that the re-ranking module adopts cross-attention over questions and passages rather than the dot product of two independently encoded embedding used in retriever. For each QA pair $d=\left\langle\langle s, r, t\rangle, q, p^{+}\right\rangle$, we select $m$ other passages in wiki-page of entity $s$ as negative passages, and maximize $P_{\text {rank }}\left(t \in p^{+}\right)$. Then, we calculate $P_{\text {start }}\left(i \mid p^{+}, q\right)$ and $P_{\text {end }}\left(i \mid p^{+}, q\right)$ and maximize the probability for the ground-truth span of target answer $t$. Combing the passage re-ranking and span extraction objectives, we get readingcomprehension loss $\mathcal{L}_{\text {read }}$.

## 4 Experiments

In this section, we evaluate RGPT-QA on three open-domain QA datasets: Natural Questions (NQ), Trivia QA and Web Questions (WQ).

### 4.1 Experiment Settings

We follow the pre-processing procedure described in DPR (Karpukhin et al., 2020) for a fair comparison. We use the English Wikipedia from Dec. 20, 2018 and split each article into passages of 100 disjoint words as the corpus. For each question in all the three datasets, we use a passage from the processed Wikipedia which contains the answer as positive passages. We evaluate the QA system by Exact Match (EM) Accuracy on the correct answer.

Our RGPT-QA could be integrated with any open-domain QA system. In this paper, we incorporate it with the recently developed QA system, Dense Passage Retriever (DPR) (Karpukhin et al., 2020) to evaluate our pre-training framework. The DPR model uses the RoBERTa-base ( $\mathrm{d}=768$, $\mathrm{l}=12$ )
model as the base encoder. We first pre-train the retriever and reader in DPR using RGPT-QA. For retriever, we use the negative passage sampling strategy (c.f. Sec. 3.2.2), with initial entity size set to be 12, batch size of 128 and the hard negative passage number of 2 . For reader, we randomly sample 64 source entities per batch to calculate the loss. For each entity, we sample 2 hard negative passages for re-ranking. We pre-train both the retriever and reader for 20 epochs using AdamW optimizer and a learning rate warm-up followed by linear decay. Pre-training is run on 8 Tesla V100 GPUs for two days. After the pre-training, we fine-tune the retriever and reader on each QA dataset following the same procedure and hyper-parameters described in DPR (Karpukhin et al., 2020).

QA Pre-Training Baselines. We compare RGPT-QA with three recently proposed pretraining methods for open-domain QA.

T5 (Raffel et al., 2020) adopts multiple generative tasks to pre-train a generative model. The fine-tuned QA models directly generate answers without needing an additional retrieval step.

ORQA (Lee et al., 2019) adopts a Inverse Cloze Task (ICT) to pre-train retriever, which forces each sentence's embedding close to context sentences.

REALM (Guu et al., 2020a) incorporates a retriever as a module into language model and trains the whole model over masked entity spans.

We directly report the results listed in their papers as they follow the same experiment settings.

We also add two knowledge-guided language models as baselines. Though not targeted at QA problem, these two methods are both designed to capture structured knowledge.

KnowBERT (Peters et al., 2019) adds entity embedding to each entity mention in text, and adopts the entity linking objective to pre-train the model.

KEPLER (Wang et al., 2019) uses Knowledge Embedding objective, i.e., TransE, to guide embedding encoded over entity description.

We initialize DPR base encoders by the released pre-trained models of these two work, and then finetune on each QA dataset with the same procedure.

We also add a Unsupervised Question Answering (Unsup.QA) (Lewis et al., 2019) as a baseline. For each entity as the answer, Unsup.QA selects a passage containing the entity as context passage and a cloze question. The cloze question is later rewritten by a machine translator to natural language. We use the generated QA dataset to pre-train both

|  | QA System Name | Pre-Training Task for QA | $\begin{gathered} \text { NQ } \\ (58.9 \mathrm{k} / 3.6 \mathrm{k}) \end{gathered}$ | $\begin{gathered} \text { Trivia QA } \\ (60.4 \mathrm{k} / 11.3 \mathrm{k}) \end{gathered}$ | $\begin{gathered} \mathbf{W Q} \\ (2.5 \mathrm{k} / 2 \mathrm{k}) \end{gathered}$ |
| :---: | :---: | :---: | :---: | :---: | :---: |
| $\begin{aligned} & \ddot{0} \\ & \stackrel{N}{y} \\ & \stackrel{0}{0} \\ & \overrightarrow{0} \end{aligned}$ | BM25+BERT (Lee et al., 2019) | - | 26.5 | 47.1 | 17.7 |
|  | HardEM (Min et al., 2019a) | - | 28.1 | 50.9 | - |
|  | GraphRetriever (Min et al., 2019b) | - | 34.5 | 56.0 | 36.4 |
|  | PathRetriever (Asai et al., 2020) | - | 32.6 | - | - |
|  | DPR (Karpukhin et al., 2020) | - | 41.5 | 56.8 | 34.6 |
|  | T5 (large) (Raffel et al., 2020) | T5 (Multitask) | 29.8 | - | 32.2 |
|  | ORQA (Lee et al., 2019) | ICT | 33.3 | 45.0 | 36.4 |
|  | REALM ${ }_{\text {Wiki }}$ (Guu et al., 2020a) | REALM | 39.2 | - | 40.2 |
|  | REALM $_{\text {News }}$ (Guu et al., 2020a) | REALM | 40.4 | - | 40.7 |
|  | DPR (KnowBERT (Peters et al., 2019)) | Entity Linking | 39.1 | 56.4 | 34.8 |
|  | DPR (KEPLER (Wang et al., 2019)) | TransE | 40.9 | 57.1 | 35.2 |
|  | DPR (Unsup.QA (Lewis et al., 2019)) | Cloze Translation | 41.9 | 57.3 | 36.5 |
|  | Ours, DPR (RGPT-QA) | RGPT-QA | 43.7 | 59.2 | 40.9 |

Table 2: End-to-end QA Exact Match Accuracy (\%) on test sets of three Open-Domain QA datasets, with the number of train/test examples shown in paretheses below. All the results except the last four rows are copied from the original papers. "-" denotes no results are available. Models in the first block are initialized by BERT/RoBERTa and then directly fine-tuned on the supervised QA datasets. While models in the second block are initialized by RoBERTa and then tuned on some QA pre-training tasks first, and then fine-tuned on the supervised QA datasets.
the retriever and reader of the DPR framework.

### 4.2 Experimental Results

| Pre-Train Model | NQ | Trivia QA | WQ |
| :--- | :---: | :---: | :---: |
| RoBERTa | $78.4 / 63.3$ | $79.4 / 72.6$ | $73.2 / 58.1$ |
| KnowBERT | $76.7 / 62.6$ | $78.9 / 72.2$ | $73.4 / 58.3$ |
| KEPLER | $77.9 / 62.8$ | $79.7 / 72.9$ | $74.5 / 58.6$ |
| Unsup.QA | $78.6 / 63.7$ | $79.9 / 73.0$ | $74.5 / 59.1$ |
| RGPT-QA | $\mathbf{8 0 . 1} / \mathbf{6 4 . 8}$ | $\mathbf{8 1 . 2} / \mathbf{7 3 . 7}$ | $\mathbf{7 6 . 7} / \mathbf{6 1 . 0}$ |

Table 3: Retrieval (left) accuracy over Top-20 results and Reader (right) Exact Match over Golden-Passages on validation sets of three Open-Domain QA datasets.

| Mask | NPS | $\mathcal{L}_{\text {distill }}$ | $\mathcal{L}_{\text {rel }}$ | NQ | Trivia QA | WQ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| $\boldsymbol{\checkmark}$ | $\boldsymbol{J}$ | $\boldsymbol{J}$ | $\boldsymbol{\checkmark}$ | 44.3 | 59.8 | 41.4 |
| $\boldsymbol{X}$ | $\boldsymbol{\checkmark}$ | $\boldsymbol{\checkmark}$ | $\boldsymbol{\checkmark}$ | 39.7 | 56.3 | 34.2 |
| $\boldsymbol{\checkmark}$ | $\boldsymbol{X}$ | $\boldsymbol{\checkmark}$ | $\boldsymbol{\checkmark}$ | 43.5 | 58.1 | 39.8 |
| $\boldsymbol{\checkmark}$ | $\boldsymbol{\checkmark}$ | $\boldsymbol{X}$ | $\boldsymbol{\checkmark}$ | 43.8 | 59.3 | 40.8 |
| $\boldsymbol{\checkmark}$ | $\boldsymbol{\checkmark}$ | $\boldsymbol{X}$ | $\boldsymbol{X}$ | 43.1 | 58.5 | 40.0 |

Table 4: Ablation of RGPT-QA components on validation sets of three Open-Domain QA datasets. Mask: Mask target entity from question and source entity from passage; NPS: Two-level Negative Passage Sampling.

Table 2 summarizes the overall EM accuracy of the QA systems on the three datasets. The DPR framework pre-trained by RGPT-QA outperforms all other open-domain QA systems. Comparing with DPR without pre-training, RGPT-QA achieves $2.2 \%, 2.4 \%$ and $6.3 \%$ enhancement in EM accuracy on the three datasets.

| B | K | NQ | Trivia QA | WQ |
| :---: | :---: | :---: | :---: | :---: |
| 128 | 2 | 80.1 | 81.2 | 76.6 |
| 128 | 1 | 79.7 | 80.8 | 76.1 |
| 64 | 2 | 79.6 | 80.6 | 75.8 |
| 64 | 1 | 79.2 | 80.1 | 75.3 |

Table 5: Ablation of batch size and negative sampling for retrieval pre-training. B: Batch Size; K: Number of other passages as negative sample.

Comparing with other pre-training tasks for QA, RGPT-QA outperforms ORQA by $10.4 \%, 14.2 \%$ and $4.5 \%$ on the three datasets, and outperforms REALM $_{\text {News }}$ by $3.3 \%$ and $0.2 \%$ on NQ and WQ. This demonstrates that the model performance can be enhanced by leveraging relational QA dataset guided by Grounded Relational Wiki-Graph. We provide a detailed analysis in Sec. 4.3.

KnowBERT and KEPER encode structural knowledge into pre-trained language models. Both models focus on generating meaningful entity embedding, and are not designed to infer relations between entities for question answering. From the table, KEPLER trained via TransE performs slightly better than KnowBERT trained via entity linking, and RGPT-QA outperforms KEPLER by $2.8 \%, 2.1 \%, 5.7 \%$ on the three datasets.

Similar to RGPT-QA, Unsup.QA (Lewis et al., 2019) also generates QA data from Wikipedia. This baseline slightly improves DPR by $0.4 \%, 0.5 \%$, $1.9 \%$ on the three datasets, while our RGPT-QA outperforms it by $1.8 \%, 1.9 \%, 4.4 \%$. As discussed in Sec 3.1, one of the main reasons that
our graph-based QA generation strategy performs better is that we adopt grounded description passages desc. $(t, s)$ and desc. $(s, t)$ from different documents as questions and contexts. This avoids the lexical overlap problem in Unsup.QA and help model to capture relational facts.

We also show the retrieval and reader performance separately on validation sets in Table 3. Compared with DPR without pre-training, RGPTQA improves top-20 accuracy of Retriever by $1.7 \%$, $1.8 \%$, and $3.5 \%$, and improves EM accuracy of Reader by $1.5 \%, 1.1 \%$, and $2.9 \%$. Also, RGPTQA outperforms all the other pre-training baselines. This shows that RGPT-QA improves both the retrieval and reader steps of open-domain QA.

Ablation Studies. We then analyze the importance of each model component in RGPT-QA. One key strategy is to mask out the target answer from questions and mask out source entities from passages during retrieval training. This can avoid the model using the entity surface to find the correct passage and answer. Without using masking strategy, the average EM performance drops $5.1 \%$. This shows that it is essential to apply the mask strategy to avoid shortcut in QA pre-training. Next, we replace the hard negative passage sampling during retrieval pre-training with random batch sampling. The average EM performance drops $1.4 \%$, showing the importance of hard negative samples. Finally, we study the unsupervised relation loss $\mathcal{L}_{\text {distill }}$ and the supervised $\mathcal{L}_{\text {rel }}$. Removing them leads to $0.5 \%$ and $1.3 \%$ performance drop, which shows the benefit of training the model to explicitly infer the relation from questions.

Another key component is the negative passage sampling for dense retrieval pre-training. We study how the batch size and number of negative sample influence the performance of trained retrieval. As is shown in Table 5, increasing batch size and negative sample size can improve the performance of retriever. Even with a small batch size and negative sample, our pre-training framework could still achieves better performance against non-pretrain baseline, showing that our approach is not sensitive to these two hyperparameters.

Few-Shot QA Performance. We analyze the improvement of RGPT-QA when only a few labelled training samples are available. We fine-tune DPR initialized by RGPT-QA on subset of Natural Questions with different percentages. As is shown in


Figure 4: Few-shot QA experiment. Figure shows EM accuracy in validation set of DPR model with and without RGPT-QA pre-training, fine-tuned with different percentage of data on Natural Questions.


Figure 5: Long-tail relation experiment. EM accuracy of questions in validation set with different relation frequency in training set.

Figure 4, RGPT-QA consistently outperforms DPR without pre-training, and the improvement is more significant with small data. Specifically, when only $0.5 \%$ (594) labelled QA pairs are provided, the DPR pre-trained by RGPT-QA can still achieve $26.0 \%$ Val EM accuracy, significantly higher than $9.4 \%$ achieved by the DPR without pre-training. The results show that RGPT-QA provides a good initialization for QA systems and reduce the requirement of large human-annotated QA dataset.

### 4.3 Generalization for long-tail relations.

As pointed out in Section 2, existing QA datasets suffer high relation bias, and thus a QA model trained on these datasets cannot generalize well to questions with long-tail relations. We thus analyze whether our RGPT-QA can remedy this issue. As is shown in Figure 5, the performance improvement of RGPT-QA against the supervised baseline is much more significant for the questions with infrequent relations. Specifically, for all relations appear less than 5 times in training set, the average EM accuracy of RGPT-QA is $33.3 \%$, significantly higher than $22.4 \%$ achieved by DPR without pre-
training. This indicates that our relation QA generation method could indeed improve the performance on QA pairs with long-tail relations. Detailed prediction results are shown in Table 11 in Appendix.

## 5 Related Works

## Unsupervised QA via Question Generation

To train a QA system without human annotation of QA pairs, Unsupervised QA has been proposed by Lewis et al. (2019) to generate synthetic <context, question, answer〉 data for training QA models. Lewis et al. (2019) synthesize the QA data by: 1) run NER or noun chunkers over randomly sampled English Wikipedia paragraphs to extract answers; 2) Treat the paragraphs surrounding the answer as context; 3) Treat the context as clozestyle question and feed into a unsupervised machine translator to generate natural questions. Some follow-up works also utilize template (Fabbri et al., 2020) and pre-trained language model (Puri et al., 2020) over masked cloze-style questions for more human-readable questions. These cloze-style unsupervised QA methods achieve promising performance than previous heuristic QA baselines but underperform supervised ones. The main limitation is that the question is generated with the masked context as input, resulting in severe overlap of lexicon and word surface with the context. Consequently, the QA model might utilize the lexical pattern as a shortcut to find the answer. To address the problem of context-question lexical overlap, Dhingra et al. (2018) assume each article has an introductory paragraph, and use this paragraph to generate answer. Li et al. (2020) retrieve the Wikipedia cited document as context, Pan et al. (2020) leverage structured tables to extract key information from context, with which to synthesize questions.

To tackle the challenges in previous studies, our framework propose to leverage the Wikipedia hyperlinks and Wikidata relations as the bridge to connect two entities with linked descriptions. With one description as question and the other as context, the question and context are semantically relevant and lexical different, which naturally solve the problem without involving any additional module.

Knowledge-Guided Pre-Training Recently, researchers investigated to inject structured knowledge into pre-trained language models. Zhang et al. (2019) and Peters et al. (2019) propose to add entity embedding to each entity mentions in text, and
add entity linking objective to guide model capture structured knowledge. Wang et al. (2019) encode entity text description as entity embeddings and train them via TransE objective. Though these work show improvements over several natural language understanding tasks, they are not dedicated to open-domain question answering tasks.

There are also several pre-training studies for QA. For retrieval, Lee et al. (2019) propose an inverse cloze task, which treats a random sentence as query and the surrounding contexts as groundtruth evidence to train a QA retrieval model. Guu et al. (2020b) propose to explicitly add a retriever module in the language model to train the retriever via language modelling pre-training. For reader, Xiong et al. (2020) propose to a weakly supervised pre-training objective. They construct some fake sentences by replacing the entities in a sentence with the other entities of the same type, and train the model to discriminate original sentence from the fake ones. Verga et al. (2020) incorporate the knowledge graph triplets into language model, so the model could utilize the triplets to predict correct entity. Sun et al. (2021) extend this work by learning a virtual knowledge base by inferring the relation between two co-occurring entity pairs.

Compared with these works, our RGPT-QA mainly differs in: 1) We do not change the base QA model, so the pre-training framework could be applied to any QA systems. 2) We explicitly model the relations between entities, which proves to benefit QA pairs with less frequent relation patterns.

## 6 Conclusion

In this paper, we propose a simple yet effective pre-training framework RGPT-QA. We leverage both the Wikipedia hyperlinks and Wikidata relation triplets to construct Grounded Relational WikiGraph, based on which we generate relational QA dataset. We then pre-train a QA model to infer the latent relation from the question, and then conduct extractive QA to get the target answer entity. RGPT-QA improves the performance of the state-of-the-art QA frameworks, especially for questions with long-tail relations.

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| Relation | Frequency | Question | True Answer |
| :---: | :---: | :---: | :---: |
| P161 (cast member) | 1915 | what was the geeks name in 16 candles | anthony michael hall |
| P175 (performer) | 1844 | who sang the original blinded by the light | bruce springsteen |
| P676 (lyrics by) | 519 | who sings the song i can see clearly now the rain is gone | johnny nash |
| P86 (composer) | 442 | who made the beavis and butthead theme song | mike judge |
| P725 (voice actor) | 334 | who plays the voice of tiana in princess and the frog | anika noni rose |
| P1346 (winner) | 283 | who has won the 2017 womens singles wimbledon tennis tournament | garbiñe muguruza |
| P50 (author) | 263 | where does the saying standing on the shoulders of giants come from | bernard of chartres |
| P17 (country) | 257 | where did the black panther party take place | united states |
| P527 (has part) | 198 | the unit of area in mks system is | metre |
| P162 (producer) | 134 | who is in the video do n 't worry be happy | bobby mcferrin |
| P276 (location) | 117 | where will the summer olympics be held in 2020 | tokyo |
| P840 (narrative location) | 103 | what state is a christmas story based in | indiana |
| P915 (filming location) | 98 | where was the movie the english patient filmed | tunisia |
| P710 (participant) | 88 | who died at the gunfight at okay corral | billy clanton |
| P170 (creator) | 87 | who came up with britain 's got talent | simon cowell |
| P1308 (officeholder) | 87 | who is the first lady of the usa | melania trump |
| P361 (part of) | 74 | who sings if you want to destroy my sweater | weezer |
| P39 ${ }_{r}$ (R: position held) | 64 | who is the attorney general for new jersey | gurbir grewal |
| P138 (named after) | 64 | who proved that mar 's orbit is elliptical not circular | nicolaus copernicus |
| P112 (founded by) | 61 | who created a settlement house with the help of other social reformers | ellen gates starr |
| $\mathrm{P} 161{ }_{r}$ (R: cast member) | 60 | who is miss sue in the blind side | kathy bates |
| $\mathrm{P} 311_{r}$ (R: instance of) | 57 | the world 's oldest epic tale told in poetry is called the epic of | epic of gilgamesh |
| P58 (screenwriter) | 57 | who wrote the story for the shape of water | vanessa taylor |
| P61 (discoverer or inventor) | 55 | who developed the analytical engine which had features of present day computers | charles babbage |
| P26 (spouse) | 53 | who does young catherine marry in wuthering heights | hareton earnshaw |
| P1923 (participating team) | 52 | who did the bengals play in the super bowl | san francisco 49ers |
| P166 (R: award received) | 51 | which indian actor has won the most national awards | amitabh bachchan |
| P674 (characters) | 50 | who said better to reign in hell than serve in heaven | satan |
| P279 (subclass of) | 49 | when does dna replication occur during the eukaryotic cell cycle | mitosis |
| P361r (R: part of) | 49 | where does the transmission of electrical impulses in the heart begin | sinoatrial node |
| P131 (is located in) | 46 | where is saba university school of medicine located |  |
| P279 ${ }_{r}$ (R: subclass of) | 45 | what are the names of the three pedals on a piano | soft pedal |
| P54 (member of sports team) | 41 | what team does steph curry brother play for | dallas mavericks |
| P1344 ${ }_{r}$ (R: participant in) | 38 | who won rupauls drag race all stars three | trixie mattel |
| P495 (country of origin) | 37 | where was the movie snow white and the huntsman filmed | united kingdom |
| P39 (position held) | 34 | who is the present speaker of lok sabha 2018 | sumitra mahajan |
| P127 (owned by) | 33 | who owns the independent newspaper in the uk | alexander lebedev |
| P607 ${ }_{r}$ (R: conflict) | 32 | in the civil war who had more soldiers | union army |
| P31 (instance of) | 32 | what kind of bridge is the mackinac bridge | suspension bridge |
| P1441 ${ }_{r}$ (R: present in work) | 29 | what is the dads name in the adams family | gomez addams |
| $\mathrm{P} 175{ }_{r}$ (R: performer) | 28 | who does sean astin play in lord of the rings | samwise gamgee |
| P36 (capital) | 28 | what is the capital of dadra and nagar haveli | silvassa |
| P921 (main subject) | 24 | what disease did susannah have in brain on fir | anti-nmda receptor encephalitis |
| P186 (material used) | 22 | what is the liquid in a magic 8 ball | alcohol |
| $\mathrm{P} 179{ }_{r}$ (R: part of the series) | 22 | what is the second book in the mortal instruments series | city of ashes |
| $\mathrm{P} 793_{r}$ (R: significant event) | 21 | which territories did the us gain in the spanish-american war | puerto rico |
| P115 (home venue) | 21 | where does portland 's nba basketball team the portland trailblazers play | moda center |
| P371 (presenter) | 21 | who won beat bobby flay shrimp and grits | bobby flay |
| P180 (depicts) | 19 | who r the 4 presidents on mt . rushmore | abraham lincoln |
| P800 ${ }_{r}$ (R: notable work) | 19 | the explorer accurately mapped the coasts of europe and north africa | piri reis |
| P136 (genre) | 17 | scott joplin is best known as a composer of what kind of music | ragtime |
| P1431 (executive producer) | 16 | who hosted the daily show before trevor noah | jon stewart |
| P47 (shares border with) | 16 | which indian states share a border with delhi | uttar pradesh |
| $\mathrm{P} 54{ }_{r}$ (R: member of sports team) | 16 | who scored the first goal in dallas stars history | neal broten |
| P144 (based on) | 16 | the tribute money depicts a scene from the | gospel of matthew |
| P57 (director) | 15 | who is the director of welcome to new york | chakri toleti |
| P488 (chairperson) | 15 | who is the leader of the democratic party now | tom perez |
| P403 (watercourse outflow) | 15 | what sea does the nile river flow into | mediterranean sea |
| P1889 (different from) | 14 | how to do alt codes on a mac | option key |
| P1441 (present in work) | 14 | when does luke skywalker find out leia is his sister | return of the jedi |
| P734 (family name) | 14 | who threw the first brick in the stonewall riots | johnson |
| P1269 (facet of) | 14 | which supreme court case established the separate but equal doctrine | plessy v. ferguson |
| P706 (takes place in) | 13 | what region of the world is greece in | southern europe |
| P176 (manufacturer) | 13 | who built the gerald r ford aircraft carrier | newport news shipbuilding |
| P84 (architect) | 12 | scottish architect who developed st martins in the field | james gibbs |
| P150 (contains) | 12 | what is the name of capital of argentina | buenos aires |
| P1532 (country for sport) | 12 | cristiano ronaldo what country does he play for | portugal |
| P800 (notable work) | 12 | what was the first book that charles dickens published | the pickwick papers |
| P641 (sport) | 11 | what is the number 1 sport in the usa | american football |
| P1001 (applies to jurisdiction) | 11 | who won the schenck v . united states case | united states |
| P206 (on lake) | 11 | where is ellis island located in new york | upper new york bay |
| P178 (developer) | 11 | ms office 2000 was developed by which company | microsoft |
| P166 (award received) | 11 | who won best actor in the academy awards this year | gary oldman |
| P102 ${ }_{r}$ (R: party) | 11 | who was known as the father of indian national congress | mahatma gandhi |
| P449 (original broadcaster) | 10 | what cbs channel is the late late show on |  |
| P2438 (narrator) | 10 | whos the main character in the great gatsby | nick carraway |
| P264 (record label) | 10 | who did the soundtrack for beverly hills cop | mca records |
| P674 ${ }_{r}$ (R:characters) | 10 | where is the story of joseph in the bible found | book of genesis |
| P1891 (signatory) | 10 | who has started reducing emissions from deforestation and forest degradation | brazil |
| P138 ${ }_{r}$ (R: named after) | 10 | roman god of underworld also called orcus and pluto | pluto |
| P69 (educated at) | 10 | where did jaren jackson senior play college basketball | georgetown university |
| P1877 (after a work by) | 10 | the movie catch me if you can is based on who | frank abagnale |

Table 6: Relation with grounded QA pairs of Natural Questions Training Set (Top 1-82 by frequency).

| Relation | Frequency | Question | Answer |
| :---: | :---: | :---: | :---: |
| P155 (follows) | 10 | what is the latest george rr martin book | a dance with dragons |
| P1029 (crew member) | 10 | who was the first to step on moon | neil armstrong |
| P3342 (significant person) | 10 | who was picked over kevin durant in the draft | greg oden |
| P749 (parent organization) | 9 | what does chi mean in chi st lukes | catholic health initiatives |
| P735 (given name) | 9 | who won in the war of alexander and porus | alexander |
| $\mathrm{P} 463_{r}$ (R: member of) | 9 | countries in the warsaw pact during the cold war | soviet union |
| P1376 (capital of) | 8 | cape town is the capital of what country | south africa |
| P156 (followed by) | 8 | the things we do for love song artist | 10cc |
| P451 (unmarried partner) | 8 | who does elena date in the vampire diaries | stefan salvatore |
| P40 (child) | 8 | howard stark is the father of what superhero | iron man |
| P159 (headquarters location) | 8 | where is the head office of rbi located | mumbai |
| P287 (designed by) | 8 | who built the world first binary digit computer z1 | konrad zuse |
| P551 (residence) | 8 | where did dorothy live in the wizard of oz | kansas |
| P647 (drafted by) | 8 | who does dwyane wade play for in the nba | miami heat |
| P30 (continent) | 8 | on what continents was the roman empire located at the height of its expansion | asia |
| P634 (captain) | 7 | who is the captain of kolkata knight riders | dinesh karthik |
| P828 (has cause) | 7 | what is the most common manifestation of portal hypertension - induced splenomegaly | cirrhosis |
| P123 (publisher) | 7 | who made all the call of duty games | activision |
| P2408 (set in period) | 7 | when did hunchback of notre dame take place | 1482 |
| $\mathrm{P} 27_{r}$ (R: country of citizenship) | 7 | who was the last ruler of the tang dynasty | emperor ai of tang |
| $\mathrm{P} 135_{r}$ (R: movement) | 7 | who wanted the catholic church to reform and address | martin luther |
| P101 ${ }_{r}$ (R: field of work) | 7 | who invented the steam engine in the 1800s | james watt |
| P941 (inspired by) | 6 | who does squealer in animal farm represent in the russian revolution | vyacheslav molotov |
| P136 ${ }_{r}$ (R: genre) | 6 | who are the founding fathers of hip hop | grandmaster flash |
| $\mathrm{P} 466^{\text {r }}$ (R: occupant) | 6 | where did the patriots play before gillette stadium | foxboro stadium |
| P119 ${ }_{r}$ (R: place of burial) | 6 | who is buried in the great mausoleum at forest lawn glendale | michael jackson |
| P88 (commissioned by) | 5 | who built the castle in just one day | toyotomi hideyoshi |
| P110 (illustrator) | 5 | scary stories to tell in the dark artist | stephen gammell |
| P1366 (replaced by) | 5 | the old greek city-state of byzantium was rebuilt and became known as | constantinople |
| P169 (chief executive officer) | 5 | who become the ceo indian it company wipro in 2016 | abidali neemuchwala |
| P3279 (statistical leader) | 5 | who is the captain of argentina national team fifa world cup 2018 | lionel messi |
| P2388 (leader's office) | 5 | who does the us department of justice report to | united states attorney general |
| $\mathrm{P} 53_{r}$ (R: family) | 5 | who began the first dynasty of egyptian rulers | narmer |
| $\mathrm{P} 2522_{r}$ (R: victory) | 5 | who won season 2 of food network star | guy fieri |
| P823 (speaker) | 5 | who wrote we shall fight on the beaches | winston churchill |
| P748 (appointed by) | 5 | who can appoint comptroller and auditor general of india | president of india |
| P1363 (points/goal scored by) | 5 | who scored the winning goal for england in the 1966 world cup final | geoff hurst |
| P22 (father) | 5 | who was the king after david in the bible | solomon |
| P1027 (conferred by) | 5 | who presents national film award traditionally in india | directorate of film festivals |
| P750 (distributed by) | 5 | who own the rights to the black panther movie | walt disney studios motion pictures |
| P825 (dedicated to) | 5 | who was the song candle in the wind written about | marilyn monroe |
| P974 (tributary) | 5 | a tributary flowing into the mississippi from the east is the | ohio river |
| P8031 (perpetrator) | 4 | who was the guy who shoot in las vegas | stephen paddock |
| P885 (river source) | 4 | what is the starting point of the mississippi river | lake itasca |
| P631 (structural engineer) | 4 | who designed the first tunnel under the river thames | marc isambard brunel |
| $\mathrm{P} 17_{r}$ (R: country) | 4 | what are the countries of the united arab emirates | sharjah |
| P98 (editor) | 4 | who was an abolitionist who published and autobiography and anti-slavery newspaper | frederick douglass |
| P737 (influenced by) | 4 | qbasic is the extension of which programming language | quickbasic |
| $\mathrm{P} 206_{r}$ (R: on lake) | 4 | where does the river mekong start and end | mekong delta |
| P2789 (connects with) | 4 | a ship traveling through the panama canal could be crossing from the | atlantic ocean |
| P740 (location of formation) | 4 | where did the beatles started their career as a band | liverpool |
| P4743 (animal breed) | 4 | what kind of dog is bo and sunny | portuguese water dog |
| P466 (occupant) | 4 | who used to play in the alamo dome | utsa roadrunners |
| P2868 ${ }_{\text {( }}$ (R: subject has role) | 4 | who is the commander in chief of military | president of the united states |
| P5053 (fastest lap) | 4 | who won the 2018 chinese formula 1 grand prix | daniel ricciardo |
| $\mathrm{P} 106_{r}$ (R: occupation) | 4 | who is the griot that sings the epic , frest | balla fasséké |
| $\mathrm{P} 50_{r}$ (R: author) | 4 | what is the title of langston hughes 's first book of poetry | the weary blues |
| P118 (league) | 4 | what conference is ohio state in for football | big ten conference |
| $\mathrm{P} 2416_{r}$ (R: sport discipline) | 4 | who has the world record for the long jump | galina chistyakova |
| P1552 (has quality) | 4 | which metal does the word ' ferrous ' refer to answer in words not symbols | iron |
| P8111 (unit) | 4 | unit of measure for area of a triangle | square metre |
| P179 (part of the series) | 4 | which games are in crash bandicoot n sane trilogy | crash bandicoot |
| $\mathrm{P} 131_{r}(\mathrm{R}$ : is located in) | 4 | what is the name of capital of andhra pradesh | amaravati |
| P7047 (enemy of) | 4 | who sent doomsday to the end of time | superman |
| $\mathrm{P} 725_{r}$ (R: voice actor) | 3 | who does the voice of the cat in the hat | martin short |
| $\mathrm{P} 61{ }_{r}$ (R: discoverer or inventor) | 3 | who discover the simple microscope first time and when | zacharias janssen |
| P6 (head of government) | 3 | who was the founder of the mauryan empire | chandragupta maurya |
| $\mathrm{P} 264{ }_{r}$ (R: record label) | 3 | this artist was signed in 1952 by atlantic and brought a string of hits | ray charles |
| P462 (color) | 3 | what color was the white house when it was built | white |
| P533 (target) | 3 | who was killed in the ides of march | julius caesar |
| P972 (catalog) | 3 | who is on the top ten most wanted | alexis flores |
| P1344 (participant in) | 3 | india 's first olympic medal win as a free nation | 1948 summer olympics |
| P106 (occupation) | 3 | what did pete best play in the beatles | drummer |
| $\mathrm{P} 1366_{r}$ (R: replaced by) | 3 | what is the old name for south africa | union of south africa |
| P171 ${ }_{r}$ (R: parent taxon) | 3 | what type of organism is made up of prokaryotic cells | archaea |
| P1411 (nominated for) | 3 | who won best director at the academy awards | guillermo del toro |
| P8345 (media franchise) | 3 | what is the first star wars movie in the series | star wars |
| P1433 (published in) | 3 | the story of seven ages by william shakespeare | as you like it |
| $\mathrm{P} 20_{r}$ (R: place of death) | 3 | who was the explorer that reached the cape of good hope at the southern tip of africa | bartolomeu dias |
| P87 (librettist) | 3 | who wrote the libretto for dido and aeneas | nahum tate |
| P3764 (pole position) | 3 | who won the abu dhabi grand prix 2017 | valtteri bottas |
| P559 (terminus) | 3 | what is the southern end of the appalachian trail | springer mountain |

Table 7: Relation with grounded QA pairs of Natural Questions Training Set (Top 83-164 by frequency).

| Relation | Frequency | Question | Answer |
| :---: | :---: | :---: | :---: |
| P366 (use) | 3 | what did the chinese use oracle bones for | pyromancy |
| P706 ${ }_{r}$ (R: takes place in) | 3 | what seven countries make up the subcontinent of south asia | sri lanka |
| P610 (highest point) | 3 | what is the highest point in the pyrenees mountains in france | aneto |
| P461 (opposite of) | 3 | the results of dehydration reactions can be reversed by | hydration reaction |
| P467 (legislated by) | 3 | which group is responsible for adopting the declaration of independence | second continental congress |
| P272 (production company) | 3 | what is the tv show riverdale based off of | archie comics |
| $\mathrm{P} 140{ }_{r}$ (R: religion) | 3 | who is the leader of the baptist denomination | thomas helwys |
| P1419 (shape) | 3 | what is the shape of the earth 's orbit around the sun | ellipse |
| P942 ${ }_{r}$ (R: theme music) | 3 | clubs who sing you 'll never walk alone | liverpool f.c. |
| P376 (planet) | 3 | this planet is home to the great red spot | jupiter |
| P5202 (adapted by) | 3 | who wrote the lyrics for the song my way | paul anka |
| P171 (parent taxon) | 3 | trees of the betel nut genus of palms | areca |
| P509 (cause of death) | 3 | what was the cause of the tollund man 's death | hanging |
| $\mathrm{P} 527{ }_{r}$ (R: has part) | 3 | the bronchi are considered to be part of the | respiratory system |
| P2849 (produced by) | 3 | where does red blood cell formation occur in adults | bone marrow |
| P460 (said to be the same as) | 3 | the word zion is an ancient biblical term that referred to what city | jerusalem |
| $\mathrm{P} 1346{ }_{r}$ (R: winner) | 3 | when did the philadelphia eagles last win the super bowl | super bowl lii |
| P2341 (indigenous to) | 3 | dogri language is spoken in which state of india | himachal pradesh |
| P355 (subsidiary) | 3 | the main agency under the department of homeland security that is responsible for border security is | u.s. customs and border protection |
| P457 (foundational text) | 3 | where does one look to find the powers of a corporation | articles of incorporation |
| P108 ${ }_{\text {( }}$ (R: employer) | 3 | who is the current ceo of mcdonald 's corporation | steve easterbrook |
| P1923 ${ }_{r}$ (R: participating team) | 3 | last time houston astros have been to the world series | 2017 world series |
| P8345 $r$ (R: media franchise) | 2 | what star wars movie came out before the last jedi | the empire strikes back |
| P112 ${ }_{r}$ (R: founded by) | 2 | real name of raj chandra in rani rashmoni | babughat |
| $\mathrm{P} 113_{r}$ (R: airline hub) | 2 | what airline has its hub in charlotte nc | american airlines |
| P156 ${ }_{r}$ (R: followed by) | 2 | what is the origin of the coptic language | egyptian language |
| P137 (operator) | 2 | who owns the white house in washington dc | national park service |
| $\mathrm{P} 1552_{r}$ (R: has quality) | 2 | what physical quantity is a measure of the amount of inertia and object has |  |
| P2175 (disease treated) | 2 | topiramate ( topamax trokendi ) is used to treat which of the following diseases | epilepsy |
| P25 (mother) | 2 | who is carries mother on days of our lives | anna dimera |
| $\mathrm{P} 170^{\text {( }}$ (R: creator) | 2 | when was beverly cleary 's first book published | henry huggins |
| P641 $r_{r}$ (R: sport) | 2 | where do the rocks from curling come from | ailsa craig |
| $\mathrm{P} 451{ }_{r}$ (R: unmarried partner) | 2 | who does raven end up with in the comics | beast boy |
| P4584 (first appearance) | 2 | what was the first game waluigi was in | mario tennis |
| P2670 (has parts of the class) | 2 | what do you rest a golf ball on | tee |
| P1040 (film editor) | 2 | who is the director of the film avatar | james cameron |
| P1056 $r$ (R: material produced) | 2 | who introduced the first micro processor in 1971 | intel |
| P1192 ${ }_{r}$ (R: connecting service) | 2 | where does the eurostar leave from in paris | gare du nord |
| P1830 (owner of) | 2 | where do the carolina panthers play home games | bank of america stadium |
| $\mathrm{P} 241{ }_{r}$ (R: military branch) | 2 | who served as the general of confederate forces during the civil war | robert e. lee |
| P111 (measure of) | 2 | joule is unit of. in mks system | energy |
| P19 ${ }_{r}$ (R: place of birth) | 2 | who was the last person to live in versaille | louis xvi |
| P291 (place of publication) | 2 | where was the institutes of the christian religion published | basel |
| P1056 (material produced) | 2 | by product of saponification of fats and oils | soap |
| P140 (religion) | 2 | of which religion is the avesta a sacred book | zoroastrianism |
| $\mathrm{P} 137{ }_{r}$ (R: operator) | 2 | where do the fisher cats play in nh | northeast delta dental stadium |
| $\mathrm{P} 162_{r}$ (R: producer) | 2 | producer and director of silence of the lambs | edward saxon |
| P1582 (fruit of (taxon)) | 2 | a plant that produces a type of bean | fabaceae |
| P286 (head coach) | 2 | 2 ) who is the current manager of liverpool fc | jürgen klopp |
| $\mathrm{P} 118{ }_{r}$ (R: league) | 2 | which nrl teams have never won a premiership | new zealand warriors |
| P413 (fielding position) | 2 | what position did ryan tannehill play in college | quarterback |
| P35 (head of state) | 2 | the longest serving samma ruler in sindh was | jam nizamuddin ii |
| P3173 (offers view on) | 2 | where is the leaning tower of pisa in italy located | pisa |
| P7959 (historic county) | 2 | archipelago that includes neolithic settlement of skara brae | orkney |
| P598r (R: commands) | 2 | union generals civil war army of the potomac | ambrose burnside |
| P306 (operating system) | 2 | what operating system does the macbook pro have | macos |
| P101 (field of work) | 2 | what did robert moog contribute to the music industry in the 1960s | electronic music |
| P27 (country of citizenship) | 2 | where is the actress that played wonder woman from | israel |
| P463 (member of) | 2 | what band is the girl from the grinch in | the pretty reckless |
| P4969 (derivative work) | 2 | what is the first book of pretty little liars | pretty little liars |
| P19 (place of birth) | 2 | where did anakin live before he met qui-gon | tatooine |
| P3938 (named by) |  | who developed the concept of an iron law of wages | ferdinand lassalle |
| P157 ${ }^{\text {(R: killed by) }}$ | 2 | who does sansa end up with in game of thrones | ramsay bolton |
| P607 (conflict) | 2 | what battle did the tuskegee airmen help win | world war ii |
| P366 ${ }_{r}$ (R: use) | 2 | what kind of wax are crayons made from | paraffin wax |
| P551r ${ }_{r}$ (R: residence) | 2 | who lived in the land of nod east of eden | cain |
| P113 (airline hub) |  | where does porter airlines fly from in toronto | billy bishop toronto city airport |
| $\mathrm{P} 927_{r}$ (R: anatomical location) | 2 | where do the ilium the ischium and the pubis meet | acetabulum |
| P1000 (record held) | 1 | who holds the world record for 100 meters | usain bolt |
| P2541 (operating area) | 1 | what states does the i pass work in | illinois |
| P4647 (place of first performan) | 1 | where does medea go at the end of the play | athens |
| P483 (studio) | 1 | where was the dark side of the moon recorded | abbey road studios |
| P197 ${ }_{r}$ (R: adjacent station) | 1 | where does the rocky mountaineer leave from in vancouver | pacific central station |
| P36 ${ }_{r}$ (R:capital) | 1 | what country is in between poland and lithuania | kaliningrad oblast |
| P1589 ${ }_{\text {r }}$ (R: lowest point) | 1 | which state is bordered to the north by the artic ocean | alaska |
| P669 (located on street) | 1 | what area of paris is the eiffel tower | champ de mars |
| P1478 (has immediate cause) | 1 | the united states ' war on terror began in the wake of which of the following events | september 11 attacks |
| P1269 ${ }_{r}$ (R: facet of) | 1 | the enlightenment idea of separation of powers included which branches of government | legislature |
| P2679 (author of foreword) | 1 | who wrote the current edition of the catechism | pope john paul ii |
| P669 ${ }_{r}$ (R: located on street) | 1 | where did the beatles take the abbey road picture | abbey road studios |
| P837 ${ }_{r}$ (R: day in year) | 1 | what are three other names for makar sankranti | magh bihu |
| P3113 (does not have part) | 1 | which element in group 1 is not an alkaline metal | hydrogen |

Table 8: Relation with grounded QA pairs of Natural Questions Training Set (Top 165-246 by frequency).

| Relation | Frequency | Question | Answer |
| :---: | :---: | :---: | :---: |
| P7047 ${ }_{r}$ (R: enemy of) | 1 | who took out the governor 's eye on walking dead | michonne |
| P59 ${ }_{r}$ (R: constellation) | 1 | brightest star in the constellation lyra dan word | vega |
| P3092 (film crew member) | 1 | who pioneered animated movies with his short feature steamboat willie in 1928 | walt disney |
| $\mathrm{P} 2348{ }_{r}$ (R: time period) | 1 | the main port of axum was the red sea city of | adulis |
| P736 (cover art by) | 1 | who wrote all quite on the western front | erich maria remarque |
| P469 (lakes on river) | 1 | where does the water from the nile come from | lake victoria |
| P205 (basin country) | 1 | in what country would you find the yellow river | china |
| $\mathrm{P921}{ }_{r}$ (R: main subject) | 1 | who began the systematic study of political science | american political science review |
| P4934 (calculated from) | 1 | a quantity $15 \mathrm{~m} / \mathrm{s}$ to the north is a measure of | velocity |
| P1411 ${ }_{r}$ (R: nominated for) | 1 | who won the first oscar for best actress | janet gaynor |
| P4147 (conjugate acid) | 1 | give the name and formula for the acid derived from the following anion chlorite | chlorous acid |
| $\mathrm{P} 276{ }_{r}$ (R: location) | 1 | the area between the tigris and euphrates rivers | mesopotamia |
| $\mathrm{P} 413_{r}$ (R: fielding position) | 1 | who has the most clean sheets in the world | iker casillas |
| $\mathrm{P} 710_{r}$ (R: participant) | 1 | in the second punic war between carthage and rome carthage formed an alliance with | massylii |
| $\mathrm{P} 2563_{r}$ (R: superpower) | 1 | who taught defence against the dark arts in book number 5 | dolores umbridge |
| P2596 (culture) | 1 | which american civilization was located in a rain forest | maya civilization |
| P1071 ${ }_{r}$ (R: location of creation) | 1 | where does the young ones develop in humans | uterus |
| $\mathrm{P} 1535{ }_{r}$ (R: used by) | 1 | what programming language is used in microsoft access | visual basic for applications |
| P400 (platform) | 1 | what consoles can you play star wars battlefront on | xbox one |
| P4913 (dialect of) | 1 | what type of arabic is spoken in palestine | south levantine arabic |
| P1066 (student of) | 1 | who is the minister during the regime of chandragupta | chanakya |
| $\mathrm{P} 3342{ }_{r}$ (R: significant person) | 1 | who went before michael jordan in the draft | hakeem olajuwon |
| P86 ${ }_{r}$ (R: composer) | 1 | who wrote the power of love celine dion | candy derouge |
| P1427 (start point) | 1 | where did the tour de france start in 1954 | amsterdam |
| P3373 (sibling) | 1 | who is the older brother mario or luigi | mario |
| P2512 ${ }_{r}$ (R: series spin-off) | 1 | which came first family guy or american dad | family guy |
| $\mathrm{P} 2505_{r} \text { (R: carries) }$ | 1 | where does the appalachian trail cross the hudson river | bear mountain bridge |
| P5009 (complies with) | 1 | what type of port is used by flash drives | usb mass storage device class |
| P2094 (competition class) | 1 | what weight class did muhammad ali fight in | heavyweight |
| P1889 ${ }_{r}$ (R: different from) | 1 | what name is given to fats that are liquid at room temperature | oil |
| P7937 (form of creative work) | 1 | wagner 's tristan und isolde is an example of | opera |
| P522 (type of orbit) | 1 | what 's the orbit of the international space station | low earth orbit |
| P1303 (instrument) | 1 | what kind of bass does john cooper play | bass guitar |
| $\mathrm{P} 737{ }_{r}$ (R: influenced by) | 1 | who are the members of 36 mafia | juicy j |
| P263 (official residence) | 1 | where did zeus spend most of his time | mount olympus |
| P201 (lake outflow) | 1 | where does the water from lake okeechobee drain | caloosahatchee river |
| $\mathrm{P} 178{ }_{r}$ (R: developer) | 1 | operating system developed in 1969 at at\&t 's bell laboratories | unix |
| P1312 (has facet polytope) | 1 | what is the opposite side of a right angle triangle | hypotenuse |
| P20 (place of death) | 1 | where did omri build his new political capital | samaria |
| P2936 (language used) | 1 | what is the national language of saudi arabia | arabic |
| $\mathrm{P} 460{ }_{r}$ (R: said to be the same as) | 1 | what color is a school bus yellow or orange | chrome yellow |
| $\mathrm{P} 682_{r}$ (R: biological process) | 1 | which protein is responsible for the breakdown of a fibrin clot | plasmin |
| P3300 (musical conductor) | 1 | who did the music for ready player one | alan silvestri |
| P547 (commemorates) | 1 | name of ship that landed at plymouth rock | mayflower |
| P2079 (fabrication method) | 1 | the medium of the artwork that decorates the sistine chapel ceiling is | fresco |
| P1037 (director / manager) | 1 | who led the red shirts to victory in sicily | giuseppe garibaldi |
| $\mathrm{P} 972_{r} \text { (R: catalog) }$ | 1 | who is number one on america 's most wanted | jason derek brown |
| $\mathrm{P} 263_{r}$ (R: official residence) | 1 | which greek god ruled over a gloomy kingdom | hades |
| P2152 (antiparticle) | 1 | a packet or unit of light energy is called a | photon |
| P1462 (standards body) | 1 | who is responsible for creating the standards used on the internet | internet engineering task force |
| $\mathrm{P} 664_{r}$ (R: organizer) | 1 | when did they start using gloves in ufc | ufc 14 |
| P937 (work location) | 1 | where did beethoven live most of his life | vienna |
| P4675 ${ }_{r}$ (R: appears in the form of) | 1 | what was robin 's name in batman and robin | dick grayson |
| $\mathrm{P} 2596{ }_{r}$ (R: culture) | 1 | a ruined city on crete centre of the minoan bronze age civilisation | knossos |
| P2554 (production designer) | 1 | who made the movie all dogs go to heaven | don bluth |
| P1038 (relative) | 1 | what is the first name of huey 's dewey 's and louie 's uncle | donald duck |
| P3301 (broadcast by) | 1 | who is broadcasting the super bowl on sunday | nbc |
| P943 (programmer) | 1 | who wrote the first computer virus called elk cloner | rich skrenta |
| $\mathrm{P} 30_{r}$ (R: continent) | 1 | is puerto rico in north or central america | puerto rico |
| P135 (movement) | 1 | what kind of art did claude monet paint | impressionism |
| P5051 (towards) | 1 | which part of the cerebral hemisphere is supplied by the middle cerebral artery | cerebrum |
| P676 (R: lyrics by) | 1 | what beatles songs does paul play drums on | dear prudence |
| P364 (original language) | 1 | what language do they speak in kite runner | dari |
| P1071 (location of creation) | 1 | a town in the netherlands known for the production of a tin glazed earthenware | delft |
| P400 ${ }_{r}$ (R: platform) | 1 | name of the windows phone 8.1 virtual assistant | cortana |
| P452 (industry) | 1 | what did the hudson bay company do for canada | retail |
| P598 (commands) | 1 | who controlled or ordered the viet cong in combat | hoàng văn thái |
| $\mathrm{Pl303}_{r}$ (R: instrument) | 1 | who introduced the bass clarinet as a solo instrument in jazz | herbie mann |
| P3491 (muscle insertion) | 1 | what is the origin and insertion of the semimembranosus | medial condyle of tibia |
| P530 (diplomatic relation) | 1 | which two countries are on the western border of bolivia | chile |
| P1542 (has effect) | 1 | what disease is caused by bacterium treponema pallidum | syphilis |
| P1336 (territory claimed by) | 1 | the falkland islands are off the coast of what south american country | argentina |
| P747 (editions) | 1 | what is the latest ms office for mac | microsoft office 2016 |
| P7153 (significant place) | 1 | on which island is the uss arizona memorial | honolulu |
| $\mathrm{P} 610^{\text {( }}$ (R: highest point) | 1 | the highest peak in north america mt . mckinley ( or denali ) is located in the state of | alaska |
| P1809 (choreographer) | 1 | who danced the lead role in appalachian spring | martha graham |
| P81 (connecting line) | 1 | what line is parsons green on tube map | district line |
| P122 (type of government) | 1 | what type of government did european settlers create in south africa in 1909 | constitutional monarchy |
| $\mathrm{P} 97_{r}$ (R: noble title) | 1 | who was crowned the first holy roman emperor | charlemagne |
| P4552 (mountain range) | 1 | what mountain range is the blue mountains part of | great dividing range |
| P658 (tracklist) | 1 | what was u 2 's lead single from' the joshua tree' | with or without you |
| P195 (collection) | 1 | where is the original star spangled banner located | national museum of american history |
| P609 (terminus location) | 1 | where does route 66 start on the east coast | chicago |

Table 9: Relation with grounded QA pairs of Natural Questions Training Set (Top 247-329 by frequency).

| Triplet <br> Question True Passage Pred. Relation Pred. Answer | ```<'edward heath',?, "admiral's cup"> <mask> of edward heath which 1971 the british prime minister, edward heath, captained one of the winning boats. recent history. ...he captained britain's winning team for the admiral's cup in 1971 - while prime minister - and also captained the team in the 1979 fastnet race... participant in: 0.33, winner: 0.19, participant: 0.04, victory: 0.03, sport: 0.03 admiral's cup (\checkmark)``` |
| :---: | :---: |
| Triplet <br> Question <br> True Passage <br> Pred. Relation <br> Pred. Answer | <'scary stories to tell in the dark', 'P110 (illustrated by)', 'stephen gammell'> <br> <mask> of scary stories to tell in the dark which evocative, nightmarish illustrations for alvin schwartz's "scary stories to tell in the dark" trilogy, he has illustrated nearly seventy scary stories to tell in the dark is a series of three collections of short horror stories for children, written by alvin schwartz and originally illustrated by stephen gammell... <br> illustrator: 0.13 , creator: 0.11 , author: 0.07 , editor: 0.02 , notable work: 0.02 <br> stephen gammell $(\checkmark)$ |
| Triplet <br> Question <br> True Passage <br> Pred. Relation <br> Pred. Answer | <'heeley', ?, 'sheffield tramway'> <br> <mask> of heeley which first routes, to attercliffe and carbrook, brightside, heeley, nether edge and owlerton opened between 1873 <br> ...sheffield's old tramway stretched from sheffield city centre to woodseats and heeley was at a time the terminus... <br> located in the administrative territorial entity: 0.21 , located in the administrative territorial entity: 0.14 , location: 0.07 , shares border with: 0.04 , terminus: 0.03 old tramway $(\boldsymbol{X})$ |
| Triplet <br> Question True Passage Pred. Relation Pred. Answer | <'pablo goncálvez', ?, 'patricia miller (tennis)'> <br> <mask> of pablo goncálvez which luisa, was the first victim of uruguayan serial killer pablo goncálvez, who suffocated the 26-year old to ...the victim was 26 years old, had a degree in history and a practicing teacher, and was the sister of the well-known tennis player patricia miller... sibling: 0.29 , relative: 0.13 , spouse: 0.04 , relative: 0.04 , place of burial: 0.02 patricia miller $(\checkmark)$ |
| Triplet <br> Question <br> True Passage <br> Pred. Relation <br> Pred. Answer | <'chai prakan district', 'P131 $(\mathrm{R}:$ located in the administrative territorial entity)', 'chai prakan', > <br> <mask> of chai prakan district which, is home to the district headquarters of<mask><mask><mask><mask> district in the far north of<mask>iang m<mask> province .chai prakan is divided into four sub-districts ("tambons"), which are further subdivided into 44 administrative villages ("muban")... <br> located in the administrative territorial entity: 0.43 , capital: 0.42 , contains administrative territorial entity: 0.04 , different from: 0.02 , contains settlement: 0.01 chai prakarn $(\checkmark)$ |
| Triplet <br> Question True Passage Pred. Relation Pred. Answer | <'gothic western', ?, 'lorin morgan-richards'> <br> <mask> of gothic western which lifestyle and his series "the goodbye family" has been categorized as gothic western. in addition to his work, rich ...in the young adult series, "the goodbye family" by lorin morgan-richards has been considered gothic western with an element of humor... <br> genre: 0.92 , movement: 0.02 , field of work: 0.0 , genre: 0.0 , occupation: 0.0 <br> lorin morgan-richards $(\checkmark)$ |
| Triplet <br> Question True Passage Pred. Relation Pred. Answer | <'quentin bell', 'P40 (child)', 'virginia nicholson'> <br> has kid of quentin bell which her father was the writer and art historian quentin bell, nephew of ...they had three children: julian bell, an artist and muralist; cressida bell, a notable textile designer; and virginia nicholson, the writer of "charleston: a bloomsbury house... child: 0.98 , father: 0.0 , student: 0.0 , sibling: 0.0 , relative: 0.0 virginia nicholson $(\checkmark)$ |
| Triplet <br> Question True Passage Pred. Relation Pred. Answer | <'take me back to london', 'P361 (part of)', 'no. 6 collaborations project'> <br> <mask> of take me back to london which the border" featuring cabello and cardi $b$, and "take me back to london" featuring stormzy ...it was released as the eighth single from sheeran's fourth studio album 'no.6 collaborations project" (2019)... <br> part of: 0.88 , performer: 0.07 , lyrics by: 0.01 , producer: 0.0 , followed by: 0.0 <br> "no. 6 collaborations project ( $\checkmark$ ) |
| Triplet <br> Question True Passage Pred. Relation Pred. Answer | <'u.s. route 441 in georgia', ? , 'lakemont, georgia'> <br> <mask> of u.s. route 441 in georgia which area between u.s. route $23 / 441$ and<mask> rabun.<mask><mask> has a post office with zip code ...from there it passes through the blue ridge mountain communities of wiley, lakemont, and tiger, the latter of which includes... terminus location: 0.15 , terminus: 0.11 , located in the administrative territorial entity: 0.08 , terminus: 0.05 , connects with: 0.03 wiley (X) |
| Triplet <br> Question True Passage Pred. Relation Pred. Answer | <'anjelica huston', 'P57r (R: directed by)', 'agnes browne'> <br> <mask> of anjelica huston which irish romantic comedy-drama film directed, produced by, and starring anjelica huston, based on the book "the mammy" by brendan o ...her next directorial effort, the irish dramedy "agnes browne" (1999) - in which she also starred as the title character- was released to mixed reviews... <br> terminus location: 0.15 , terminus: 0.11 , located in the administrative territorial entity: 0.08 , terminus: 0.05 , connects with: 0.03 <br> "agnes browne ( $\checkmark$ ) |
| Triplet <br> Question True Passage Pred. Relation Pred. Answer | <'cadillac eldorado', ?, 'oldsmobile toronado'> <br> <mask> of cadillac eldorado which 1967, cadillac adopted its own version of the upp for the cadillac eldor<mask>, using the cadillac v8 engine. <br> ...by 2000, the eldorado was the last of a dying breed: its buick riviera and oldsmobile toronado stablemates had been discontinued, as had its perennial rival the lincoln mark... <br> follows: 0.38 , followed by: 0.05 , brand: 0.04 , based on: 0.02 , subclass of: 0.02 <br> oldsmobile toronado ( $\checkmark$ ) |
| Triplet <br> Question <br> True Passage <br> Pred. Relation <br> Pred. Answer | <'corsican nuthatch', 'P138 (named after)', 'john whitehead (explorer)'> <br> eponym of corsican nuthatch which 82 where he discovered a bird new to science, the corsican nuthatch. white<mask> travelled in malacca, north borneo, ...the corsican nuthatch was discovered by the english collector john whitehead in june 1883 when he shot a specimen while on a trip in the corsican mountains... named after: 0.97 , discoverer or inventor: 0.01 , named after: 0.0 , different from: 0.0 , place served by transport hub: 0.0 john whitehead $(\checkmark)$ |
| Triplet <br> Question <br> True Passage <br> Pred. Relation <br> Pred. Answer | <'mutual information', ?, 'information content'> <br> <mask> of mutual information which formula_13 is also often used for the related quantity of mutual<mask>, many authors use a lowercase formula_14 for ...it quantifies the "amount of information" (in units such as shannons (bits), nats or hartleys) obtained about one random variable through observing the other random variable... subclass of: 0.48 , different from: 0.08 , opposite of: 0.06 , subclass of: 0.05 , said to be the same as: 0.03 information theory $(\boldsymbol{X})$ |
| Triplet <br> Question True Passage Pred. Relation Pred. Answer | ```<'oculus (film)', 'P272 (production company)', 'intrepid pictures'> <mask> of oculus (film) which". in may 2012 filmdistrict acquired the film rights to what would become "oculus". soon after, the film released on april 11 ...eventually, intrepid pictures expressed interest in producing the film "as long as you don't do it found footage".... production company: 0.29, producer: 0.19, distributed by: 0.07, screenwriter: 0.04, director: 0.03 intrepid pictures (\checkmark)``` |

Table 10: Examples of generated Relational QA datapoints and the predicted relation and answer by DPR pre-trained via
RGPT-QA.

| Relation Name | Freq | Question | True Answer | RGPT-QA Predictio | Supervised DPR Prediction |
| :---: | :---: | :---: | :---: | :---: | :---: |
|  | 0 | theme song to bridge on the river kwai <br> phenothiazines such as chlorpromazine were the first type of who does the call to prayer in islam what product or market does netflix deal with mohair is made from the fleece of what animal where is the leaning tower of pisa built who is the founder of ramoji film city who bore abraham first son in the bible who is the youngest judge currently sitting on the u.s. supreme court corpora cavernosa and corpus spongiosum are anatomic structures of urdu is the official language of which state what is the source of the colorado river river that joins the severn near chepstow crossword who was the second ruler of the davidic monarchy who is considered by many to be the father of soul who brought surf music to a national audience who owns flying j and pilot truck stops who plays the mom in cheaper by the dozen where did the gabbie show go to college who did the movie i can only imagine who stars in the movie the quiet place pitt players in the nfl hall of fame who was the captain when india played its first-ever odi what state is directly west of north dakota who raised the american flag on iwo jima faces of the presidents on mt. rushmore who won the 2015 great british baking show who presented gardeners world from 2008 to 2010 who are the new hosts of british bake off what kind of meat is on a t-bone new york times co v sullivan held that there must be proof of what kind of money do they use in russia how does a plane wing create lift which physics concept applies who used the springfield rifle in the civil war waste water that contain solid and liquid excreta refers to when a blood vessel is injured the first phase in hemostasis to occur is what part of new york is coney island what regions of south asia have the highest population densities what led to the downfall of the incan empire the settlement of the israelites in canaan is the theme of which book most number of national awards for best actress who did melbourne beat in the 1964 grand final who does jackson end up with in sons of anarchy which of the following is the si unit for length what is the most abundant neurotransmitter in the nervous system pricing tactics lower the price of a product below cost who was saint patrick's day named after arabian sea is the part of which ocean who produces the most tires in the world what is the name of the governor of new jersey what two groups were fighting in the chinese civil war who played the superbowl halftime show last year who came second in the overall ranked of the tour de france last year what city does the terminator take place in where was back to the future three filmed | the river kwai march antipsychotic <br> muezzin <br> streaming media <br> angora goat <br> pisa <br> ramoji rao <br> hagar <br> neil gorsuch <br> penis <br> pakistan <br> la poudre pass <br> river wye <br> solomon <br> james brown <br> the beach boys <br> pilot corporation <br> bonnie hunt <br> university of pittsburgh <br> erwin brothers <br> john krasinski <br> mike ditka <br> ajit wadekar <br> montana <br> michael strank <br> abraham lincoln <br> nadiya hussain <br> toby buckland <br> noel fielding <br> beef <br> actual malice <br> kopeks <br> force <br> united states <br> sewage <br> coagulation <br> brooklyn <br> philippines <br> battle of cajamarca <br> joshua <br> shabana azmi <br> collingwood football club <br> tara knowles <br> metre <br> serotonin <br> loss leader <br> saint patrick <br> indian ocean <br> lego <br> phil murphy <br> communist party of china <br> bruno mars <br> rigoberto urán <br> los angeles <br> monument valley |  |  |
| subject has rol | 0 |  |  | psychiatry | edi |
| practice | 0 |  |  |  | osqu |
| industry | 1 |  |  | ng med | etflix |
| made from | 2 |  |  | ngora goat | goa |
| offers view on | 2 |  |  |  | pisa |
| R : residenc | 2 |  |  | ram | telugu film producer ramoji |
| mothe | 2 |  |  | sarah | yishma'el |
| R : employer | 3 |  |  | neil gorsuc | leonard i. gart |
| R : has part | 3 |  |  | penis | corpus caverno |
| indigenous | 3 |  |  | jh | jammu and kashmir |
| river source | 4 |  |  | la poudre | colorado begins at la poudre pa |
| tributary | 5 |  |  | river lugg | lugg |
| R : family | 5 |  |  | jeroboa | solomon's son, rehoboa |
| R : genre | 6 |  |  | james bro | sam cook |
| R : genre | 6 |  |  | the beach boy | dean |
| parent organization | 8 |  |  | pilot corporatio | berkshire hathaway |
| narrator | 10 |  |  | bonnie hun | kate |
| educated | 10 |  |  | university of pittsburg | the university of pittsburgh |
| director | 15 |  |  | the erwin brothers | bart milla |
| executive producer | 16 |  |  | john krasi | emily blu |
| R : player of | 16 |  |  | ruben brow | tony dorse |
| R : player of | 16 |  |  | srinivasaraghavan | s |
| shares border with | 16 |  |  | montana | manito |
| depicts | 19 |  |  | ira | rene gagnon, ira hay |
| depicts | 19 |  |  | thomas jeffer | theodore roosevelt |
| R : notable work | 19 |  |  | joanne wheatley | edd kimber |
| presenter | 21 |  |  | joe swift | carol klein and joe swift |
| presenter | 21 |  |  | noel fieldin | sandi toksvig |
| material used | 22 |  |  | cut from the short loin | tenderloin |
| main subject | 24 |  |  | malic | truth |
| instance of | 29 |  |  | ruble or | the russian ruble or roubl |
| instance of | 29 |  |  | newton's second law | reaction force |
| country of origin | 37 |  |  | marine corp | army |
| R : subclass of | 44 |  |  | sewage | pathogens |
| R : subclass o | 44 |  |  | wound healin | endothelial injury |
| located in | 46 |  |  | brooklyn | borough of brooklyn |
| R : part of | 49 |  |  | philippine | indonesia |
| R: part of | 49 |  |  | captured | victory |
| characters | 50 |  |  | the book of joshu | book of joshua |
| R : award received | 51 |  |  | five | three |
| participating team | 52 |  |  | collingw | melbourne football clu |
| spouse | 53 |  |  | tara knowl | opie winston |
| R : instance of | 56 |  |  | m | litre |
| R : instance of | 56 |  |  | seroton | glutamate |
| R : instance of | 56 |  |  | loss leade | increase in profi |
| named af | 64 |  |  | saint patrick | saint patrick |
| part of | 70 |  |  | northern indian ocean | the northern indian ocea |
| part | 70 |  |  | lego tire: lego tire a lego | lego blocks. lego |
| officeholde | 87 |  |  | phil murphy | democrat phil murphy |
| participan | 88 |  |  | communist party of china | republic of china |
| participan | 88 |  |  | beyoncé | coldplay |
| participant | 88 |  |  | rigoberto ur | chris froome |
| filming locatio | 98 |  |  | los angeles | hemdale |
| filming location | 98 |  |  |  | jamestown, california |

Table 11: Comparison of the prediction of DPR initialized by RGPT-QA with DPR without pre-training. These are all samples that two models made different predictions, and the relation frequency in the training set is less than 100 .

| Question | Predicted Answer | True Answer | Match? | Predicted Relation |
| :---: | :---: | :---: | :---: | :---: |
| Who played mr darling on andy griffith show | Denver Pyle | Denver Pyle | $\checkmark$ | P175 (performer) |
| Who voices flik in a bug's life | Dave Foley | Dave Foley | $\checkmark$ | P725 (voice actor) |
| Who's the dad of blair waldorf's baby | Chuck | Chuck | $\checkmark$ | P26 (spouse) |
| Where do you think glaciers can be found today | rocky mountains | mountain ranges on every continent | $x$ | $\mathrm{P} 31{ }_{r}$ (R: instance of) |
| When did ginny weasley join the quidditch team | half-blood prince | half-blood prince | $\checkmark$ | P674 ${ }_{r}$ (R: characters) |
| When does far cry 5 for ps 4 come out | 2018 | march 27, 2018 | $x$ | P400 (game platform) |
| When do millennials end and gen z start | mid-1990s to mid-2000s | mid-1990s | $x$ | P155 (preceded by) |
| Who killed hotchner's wife in criminal minds | George Foyet | George Foyet | $\checkmark$ | P7047 (enemy of) |
| Who said walk tall and carry a big stick | u.s. president theodore roosevelt | theodore roosevelt | $x$ | P170 (creator) |
| Who does the voice of sheen from jimmy neutron | Jeffrey Garcia | Jeffrey Garcia | $\checkmark$ | P734 (family name) |
| How many seasons of gossip girl are there | 6 | 6 | $\checkmark$ | P527 (has part) |
| What can be used to detect the charge of particles | ionization detectors | particle detector | $x$ | P279 ${ }_{r}$ (R: subclass of) |
| Who was robin in the original batman series | Burt Ward | Burt Ward | $\checkmark$ | P161 (cast member) |
| What is the song funky cold medina about | a love potion | a fictional aphrodisiac | $x$ | P138 (named after) |
| What do you call a quarter pounder in france | royal cheese | royal cheese | $\checkmark$ | P1889 (different from) |
| Who developed the first alternating current electric system | Galileo Ferraris | Nikola Tesla | $x$ | P61 (inventor) |
| Who won s5 of rupaul's drag race | Jinkx Monsoon | Jinkx Monsoon | $\checkmark$ | P1346 (winner) |
| When was the svalbard global seed vault built | 2006 | 2006 | $\checkmark$ | P88 (built for) |
| Who was the mother of dragons married to | Khal Drogo | Dothraki Horselord Khal Drogo | $x$ | P26 (spouse) |
| Which organization sets monetary policy for the united states | the federal reserve | the federal reserve | $\checkmark$ | P1001 (jurisdiction) |
| What season of the voice was miley cyrus on | eleventh | season 11 | $x$ | P179 (part of series) |
| Upon which document in american history is the language of the declaration of sentiments based | united states declaration of independence | united states declaration of independence | $\checkmark$ | P144 (based on) |
| What kind of car does dale earnhardt jr drive | chevrolet camaro | chevrolet | $x$ | P54 (played for) |
| How many times did brazil win the fifa world cup | five | five | $\checkmark$ | P1344 (participant in) |
| Second life is an example of a | massively multiplayer online role-playing games | an online virtual world | $x$ | P31 (instance of) |
| What percentage of the world's population lives in east asia | 22\% | 22\% | $\checkmark$ | $\mathrm{P} 276^{\text {( }}$ (R: located in) |
| From which body part shurpnakha drive her name | fingernails | fingernails | $\checkmark$ | P186 (ingredient) |
| How many chapters does the gospel of john have | four | 21 | $x$ | P527 (has part) |
| Who sang the original always on my mind | b.j. thomas | gwen mccrae | $x$ | P175 (performer) |
| Where does the amazon river start and finish | atlantic ocean | atlantic ocean | $\checkmark$ | P403 (watercourse outflow) |
| Who did dwayne wade play for last year | the miami heat | miami heat | $x$ | P647 (drafted by) |
| Who owns the rights to the power rangers | Hasbro | Hasbro | $\checkmark$ | P127 (owned by) |

Table 12: Predicted relations for those QA pairs in Natural Questions Valid Set that cannot be aligned to WikiData.


[^0]:    ${ }^{1}$ Dataset and code are released at https://github. com/acbull/RGPT-QA.

