

# Course Outline

- Question Answering systems (Q.A.) +lab (Q.A. fine-tune BERT model)
- Multilingual and Multimodal Q.A. +lab (Q.A. test BERT models)
- Information Retrieval (I.R.) +lab (I.R. train and test BM25 model)
- Open Domain Q.A. +assignment (Open Domain Q.A)

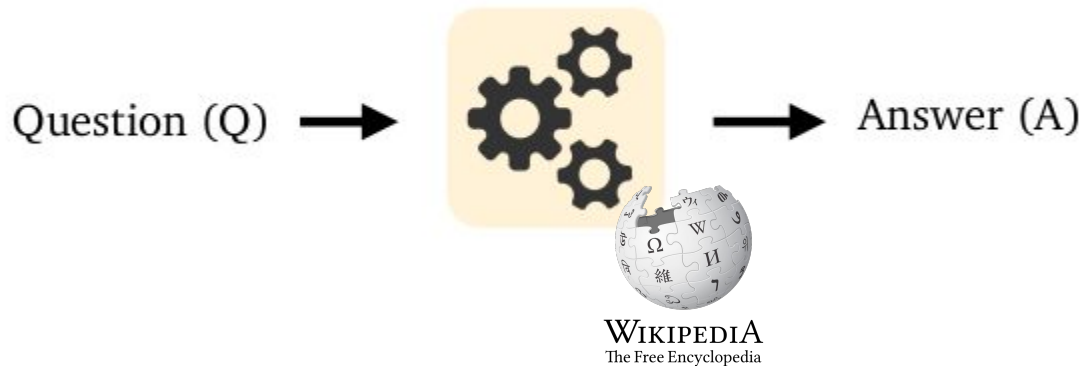
# Course Outline

- Question Answering systems (Q.A.) +lab
- Multilingual and Multimodal Q.A. +lab
- Information Retrieval (I.R.) +lab
- **Open Domain Q.A.** +assignment(copyPaste...) + final project
  - Based on Danqi Chen (Princeton University) & Christopher Manning (Stanford University) slides. Special thanks to Jon Ander Campos (UPV/EHU)

# Course Outline

- **Open Domain Question Answering systems**
  - Introduction
  - Open Domain Q.A. practical applications
  - SOA models

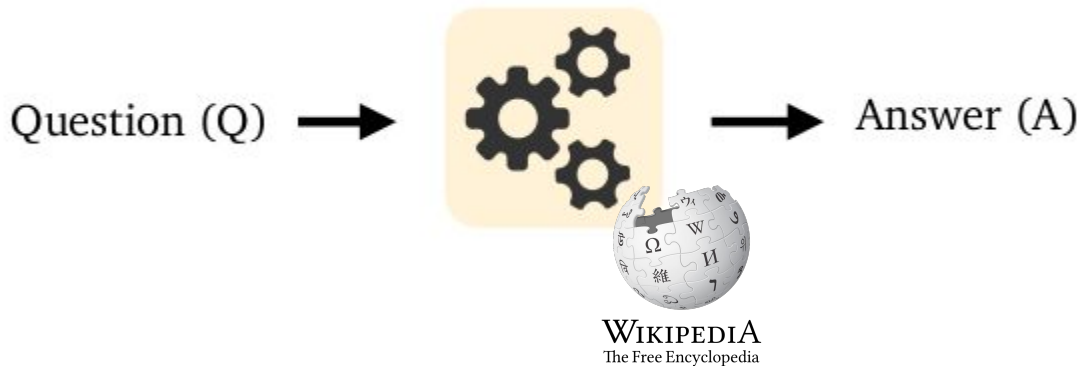
# Open Domain Q.A.



- Different from reading comprehension, we don't assume a given passage.
- Instead, we only have access to a large collection of documents (e.g., Wikipedia). We don't know where the answer is located, and the goal is to return the answer for any open-domain questions.
- Much more challenging but **a more practical problem!**

# Open Domain Q.A.

In contrast to “closed-domain” systems that deal with questions under a specific domain...



- Different from reading comprehension, we don't assume a given passage.
- Instead, we only have access to a large collection of documents (e.g., Wikipedia). We don't know where the answer is located, and the goal is to return the answer for any open-domain questions.
- Much more challenging but **a more practical problem!**

# Open Domain Q.A.

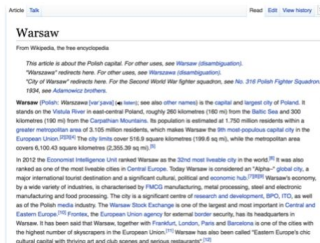
## Open-domain QA

SQuAD, TREC, WebQuestions, WikiMovies

Q: How many of Warsaw's inhabitants spoke Polish in 1933?



**Document  
Retriever**

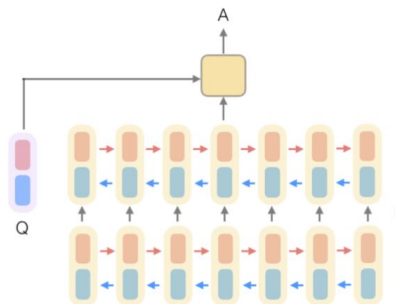


**Document  
Reader**

833,500



**Information  
Retrieval  
Model (I.R.)**



**BERT**  
fine-tuned for  
Q.A.

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# Open Domain Q.A.: Lots of practical application



Where is the deepest lake in the world?



All



Maps



Images



News



Videos

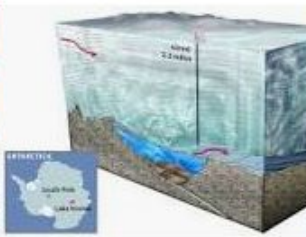


More

Settings

Tools

About 21,100,000 results (0.71 seconds)



## Siberia

Lake **Baikal**, in Siberia, holds the distinction of being both the deepest lake in the world and the largest freshwater lake, holding more than 20% of the unfrozen fresh water on the surface of Earth.



# Open Domain Q.A.: Lots of practical application

**Is the use of screening of neutralizing antibodies such as ELISAs valid for early detection of disease?**

In a study of 623 sars patients , the neutralizing - antibody levels peaked at 20 - 30 days and were sustained for over 150 days . [[Pathogenesis of severe acute respiratory syndrome, Current Opinion in Immunology, 2005-08-31](#)]

Detection of serum IgG , IgM and IgA against SARS - CoV using immunofluorescent assays and by ELISA against nucleocapsid antigen occurs around the same time with most patients seroconverted by day 14 after onset of illness [ 48 ] . IgG can be detected as early as 4 days after the onset of illness . The kinetics of neutralization antibodies nearly parallel those for IgG [ 48 ] and most of the neutralizing - antibody activity is attributed to IgG [ 49 ] . In a study of 623 SARS patients , the neutralizing - antibody levels peaked at 20 - 30 days and were sustained for over 150 days . These antibodies can neutralize the pseudotype particles bearing the S protein from different SARS - CoV strains , suggesting that these antibodies are broadly active and that the S protein is highly immunogenic [ 49 ] . Indeed the S protein , among the other structural proteins , such as M , E or N , is the only significant SARS - CoV neutralization antigen and protective antigen [ 50 ] , with amino acids 441 - 700 as the major immunodominant epitope [ 51 ] .

Early antibodies are detected in some patients within two weeks . [[Severe acute respiratory syndrome and dentistry A retrospective view, The Journal of the American Dental Association, 2004-09-30](#)]

Enzyme - linked immunosorbent assay , or ELISA , test . From about 20 days after the onset of clinical signs , ELISA tests can be used to detect immunoglobulin , or Ig , M and IgA antibodies in the serum samples of patients with SARS . Early antibodies are detected in some patients within two weeks .

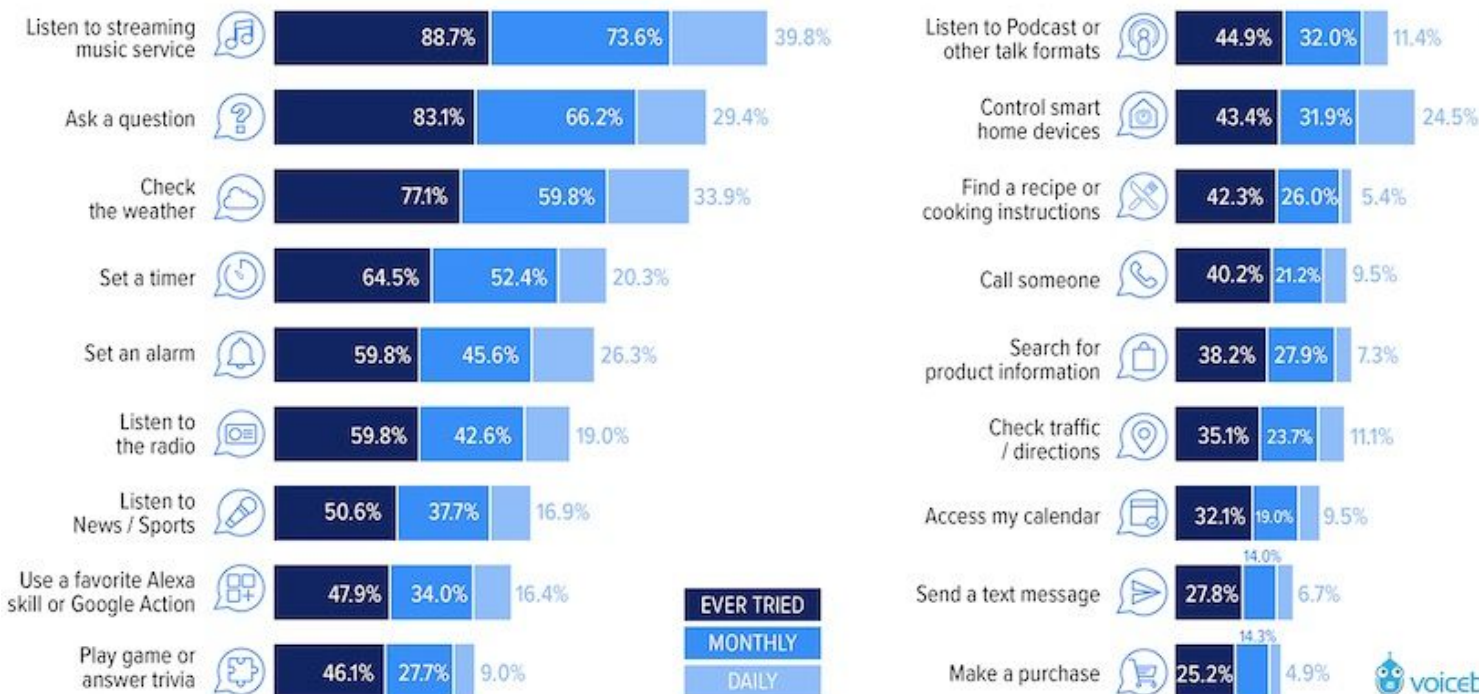
System to collect answers of Covid-related questions in scientific publications

**Winner in two competitions** (White House, NIH)

# Open Domain Q.A.: Lots of practical application



Smart Speaker Use Case Frequency January 2020



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# Open Domain Q.A. baseline:

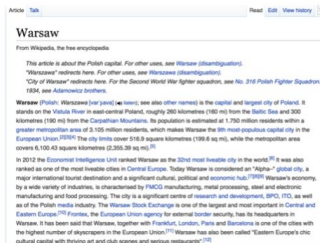
## Open-domain QA

SQuAD, TREC, WebQuestions, WikiMovies

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**Document  
Retriever**

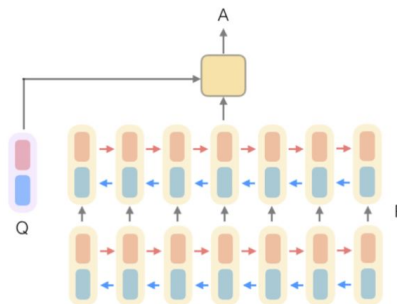


**Document  
Reader**

833,500



**Information  
Retrieval  
Model (I.R.)**



**BERT**  
fine-tuned for  
Q.A.

# Open Domain Q.A. baseline:

## Open-domain QA

SQuAD, TREC, WebQuestions, WikiMovies

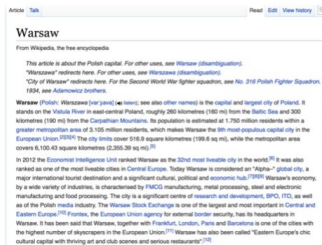
**SubOptimal!**

we can not fix the  
Retriever errors

Q: How many of Warsaw's inhabitants  
spoke Polish in 1933?



**Document  
Retriever**

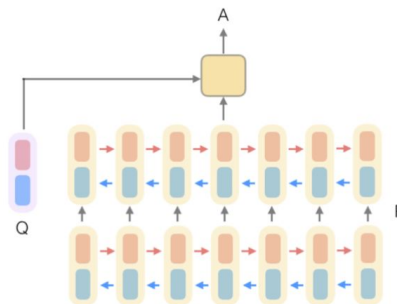


**Document  
Reader**

833,500



**Information  
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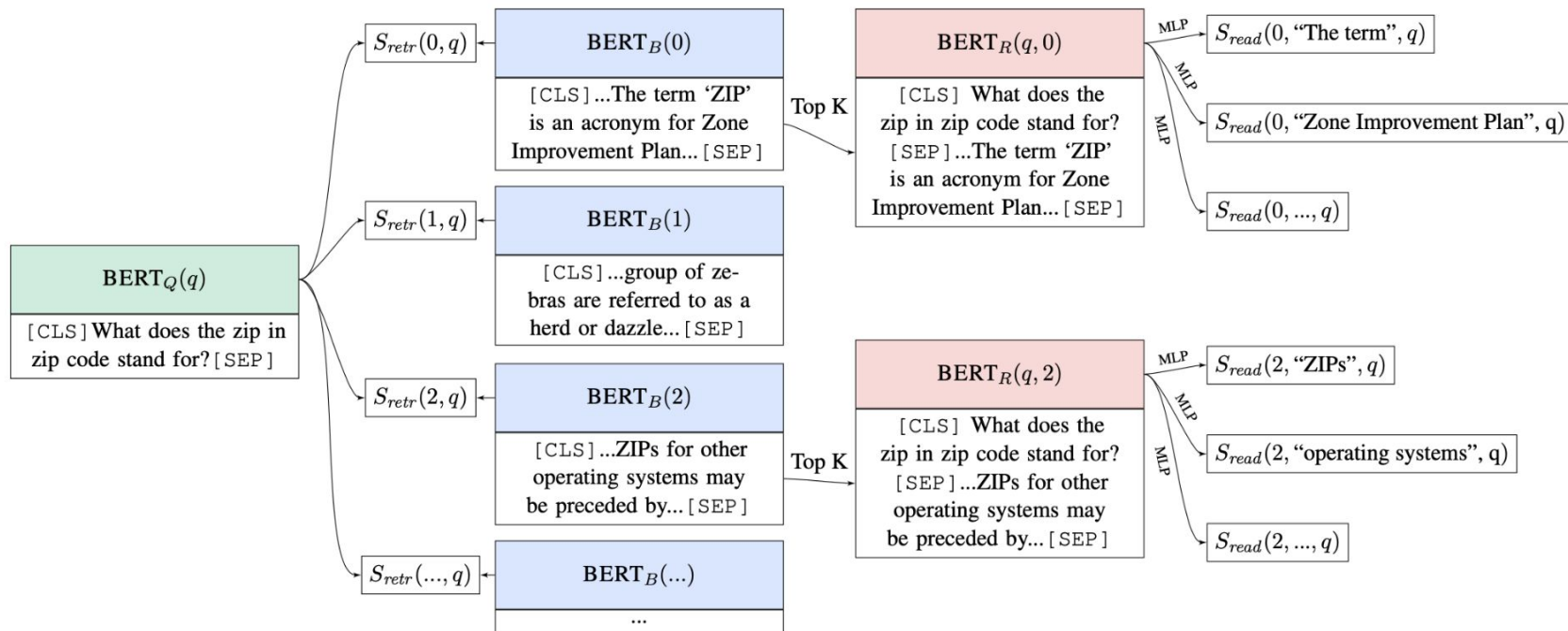
# Course Outline

- **Open Domain Question Answering systems**
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    - Lee et al., 2019. **Latent Retrieval for Weakly Supervised Open Domain Question Answering**

# Open Domain Q.A. we can train the retriever too! forget about I.R. system

- Joint training of retriever and reader
- Each text passage can be encoded as a vector using BERT and the retriever score can be measured as the dot product between the question representation and passage representation.
- However, it is not easy to model as there are a huge number of passages (e.g., 21M in English Wikipedia)

# Open Domain Q.A. we can train the retriever too



- This model outperforms BM25 by up to 19 points in exact match.



# Open Domain Q.A. we can train the retriever too

- **Retriever**

- In order for the retriever to be learnable, we define the retrieval score as the inner product of dense vector representations of the question  $q$  and the **evidence block  $b$** .

$$h_q = \mathbf{W}_q \text{BERT}_Q(q)[\text{CLS}]$$

$$h_b = \mathbf{W}_b \text{BERT}_B(b)[\text{CLS}]$$

$$S_{\text{retr}}(b, q) = h_q^\top h_b$$

# Open Domain Q.A. we can train the retriever too

- **Reader**

- The reader is a span-based variant of the reading comprehension model (finnetuned BERT):

$$\begin{aligned}h_{start} &= \text{BERT}_R(q, b)[\text{START}(s)] \\h_{end} &= \text{BERT}_R(q, b)[\text{END}(s)] \\S_{read}(b, s, q) &= \text{MLP}([h_{start}; h_{end}])\end{aligned}$$

# Open Domain Q.A. we can train the retriever too

- **Joint learn Reader & Retriever**

$$S(b, s, q) = S_{retr}(b, q) + S_{read}(b, s, q)$$

# Open Domain Q.A. we can train the retriever too

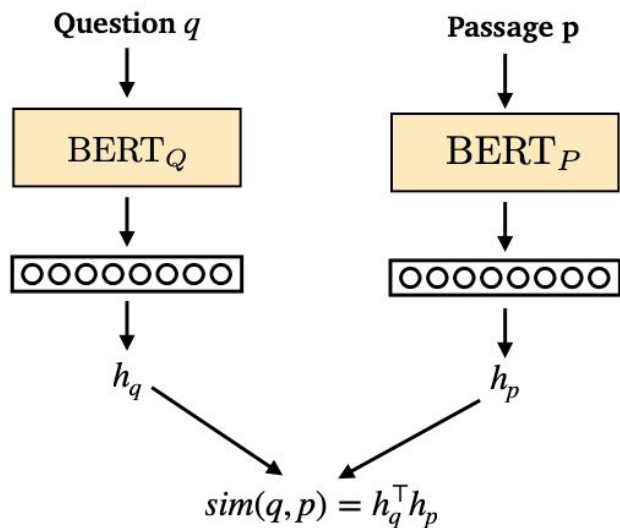
- We still have millions of evidence blocks (passages)...
- Reduce the search space by:
  - Pretraining of retrieval (Inverse Cloze Task)
  - Pre-compiled index for inference (Locality Sensitive Hashing)
  - Beam-search over top 5 candidates
  - ...

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# Open Domain Q.A. we can train the retriever too

- Dense passage retrieval (DPR) - We can also just train the retriever using question-answer pairs!
- Trainable retriever (using BERT) largely outperforms traditional IR retrieval models

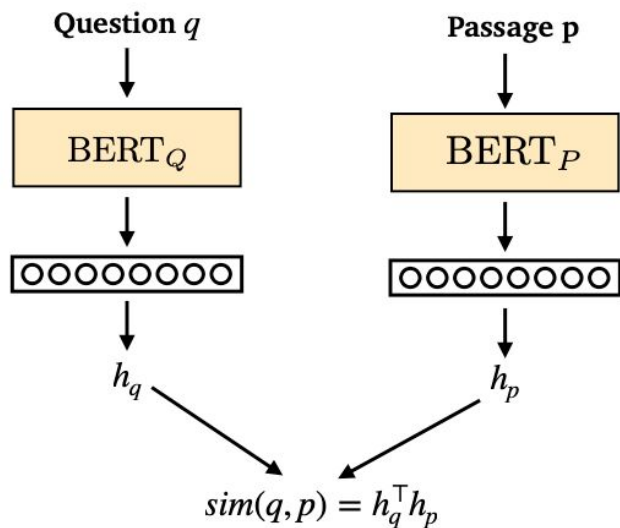


## How to train the retriever?

The goal is to create a vector space such that relevant pairs of questions and passages will have smaller distance

# Open Domain Q.A. we can train the retriever too

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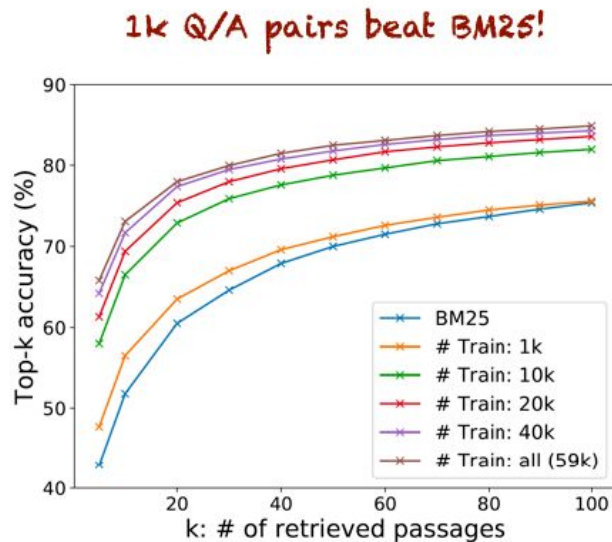
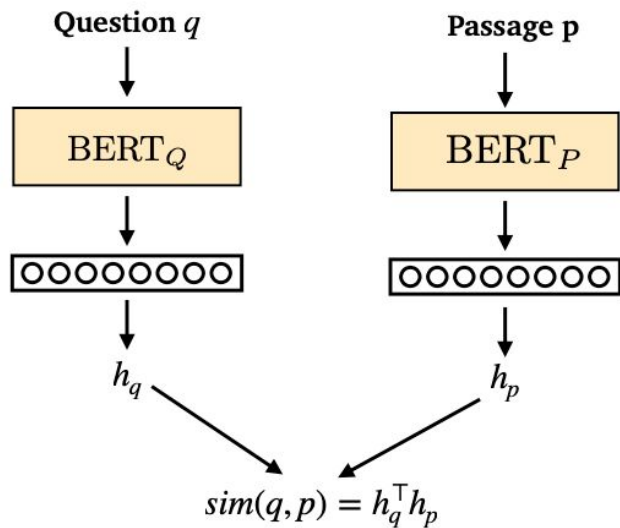


## How to train the retriever?

Using positive and negative question passage pairs!

# Open Domain Q.A. we can train the retriever too

- Dense passage retrieval (DPR) - We can also just train the retriever using question-answer pairs!
- Trainable retriever (using BERT) largely outperforms traditional IR retrieval models



Code here:  
<https://github.com/facebookresearch/DPR>



# Open Domain Q.A. we can train the retriever too

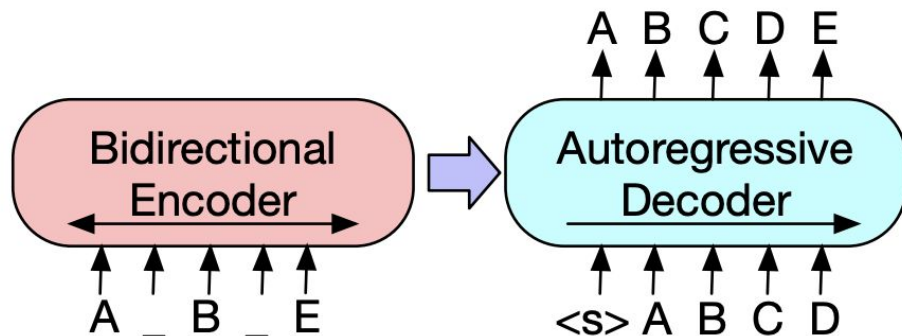
- We still have millions of evidence blocks (passages)...
- Reduce the search space by:
  - Precompute all vector representation for passages
  - **FAISS** for indexing representations
    - open-source library for similarity search and clustering of dense vectors, which can easily be applied to billions of vectors

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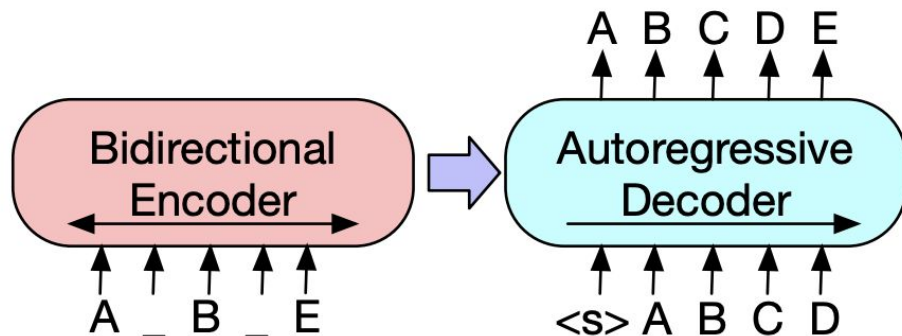
# Open Domain Q.A. Dense retrieval + Generative Models

- Recent work shows that it is beneficial to **generate answers** instead of to **extract answers**. Seq2Seq architecture to **generate** text:



# Open Domain Q.A. Dense retrieval + Generative Models

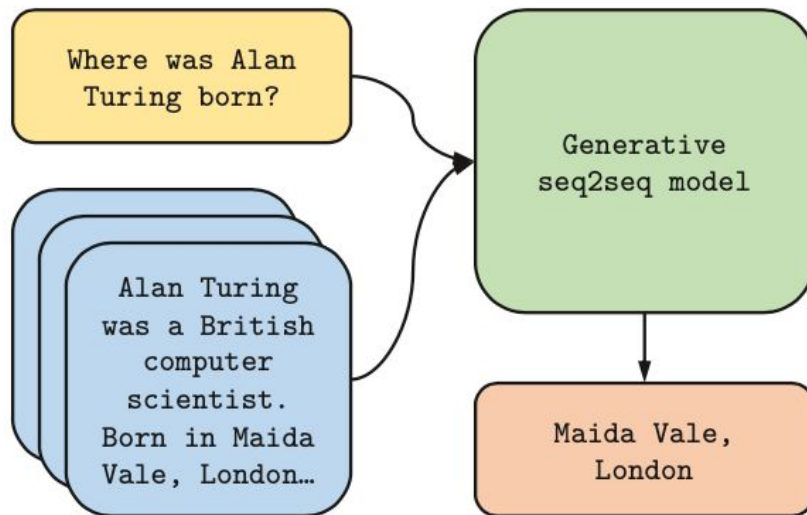
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**BERT is old fashioned!**

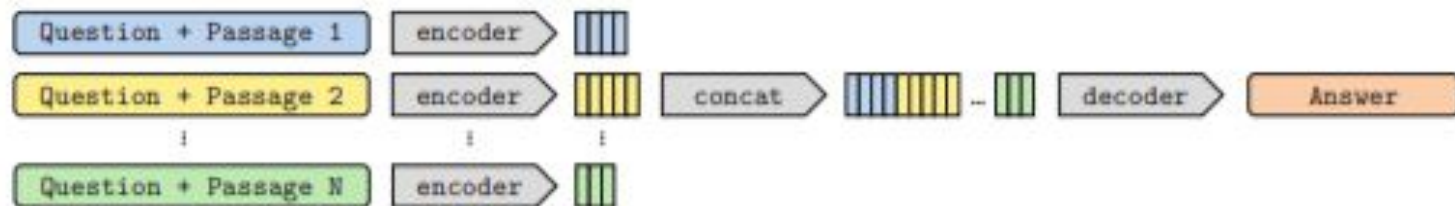
# Open Domain Q.A. Dense retrieval + Generative Models

- Recent work shows that it is beneficial to **generate answers** instead of to **extract answers**.



# Open Domain Q.A. Dense retrieval + Generative Models

- Recent work shows that it is beneficial to **generate answers** instead of to **extract answers**.



5, 10, 25, 50, 100...  
the more the better!!

# Open Domain Q.A. Dense retrieval + Generative Models

- Recent work shows that it is beneficial to **generate answers** instead of to **extract answers**.

Model		NQ EM	TriviaQA EM EM		SQuAD Open EM F1
BM25 + BERT	DrQA (Chen et al., 2017)	-	-	-	29.8 -
	Multi-Passage BERT (Wang et al., 2019)	-	-	-	53.0 60.9
	Path Retriever (Asai et al., 2020)	31.7	-	-	<b>56.5</b> <b>63.8</b>
	Graph Retriever (Min et al., 2019b)	34.7	55.8	-	- -
	Hard EM (Min et al., 2019a)	28.8	50.9	-	- -
	ORQA (Lee et al., 2019)	31.3	45.1	-	20.2 -
	REALM (Gua et al., 2020)	40.4	-	-	- -
	DPR (Karpukhin et al., 2020)	41.5	57.9	-	36.7 -
	SpanSeqGen (Min et al., 2020)	42.5	-	-	- -
	RAG (Lewis et al., 2020)	44.5	56.1	68.0	- -
	T5 (Roberts et al., 2020)	36.6	-	60.5	- -
	GPT-3 few shot (Brown et al., 2020)	29.9	-	71.2	- -
	Fusion-in-Decoder (base)	48.2	65.0	77.1	53.4 60.6
	Fusion-in-Decoder (large)	<b>51.4</b>	<b>67.6</b>	<b>80.1</b>	<b>56.7</b> 63.2

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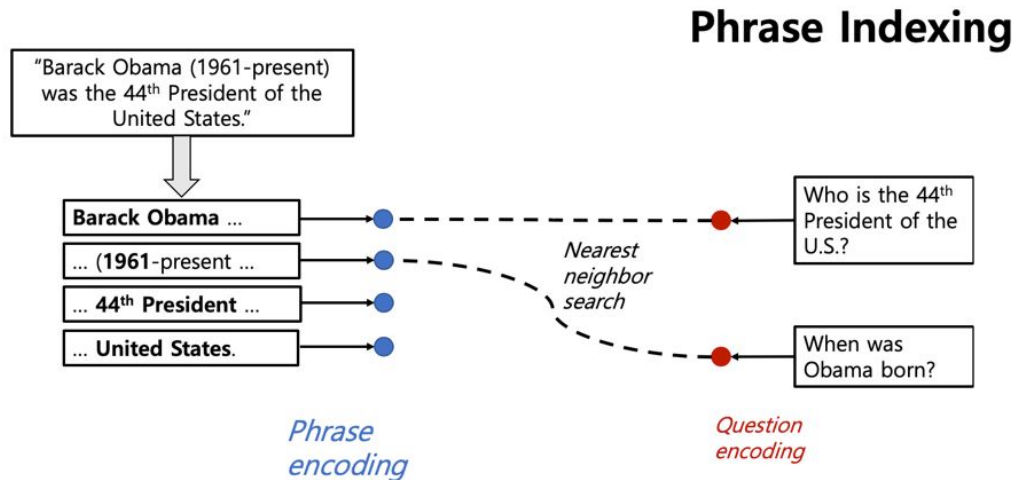
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  - Lee et al., 2020. **Learning Dense Representations of Phrases at Scale**



# Open Domain Q.A. without reader...

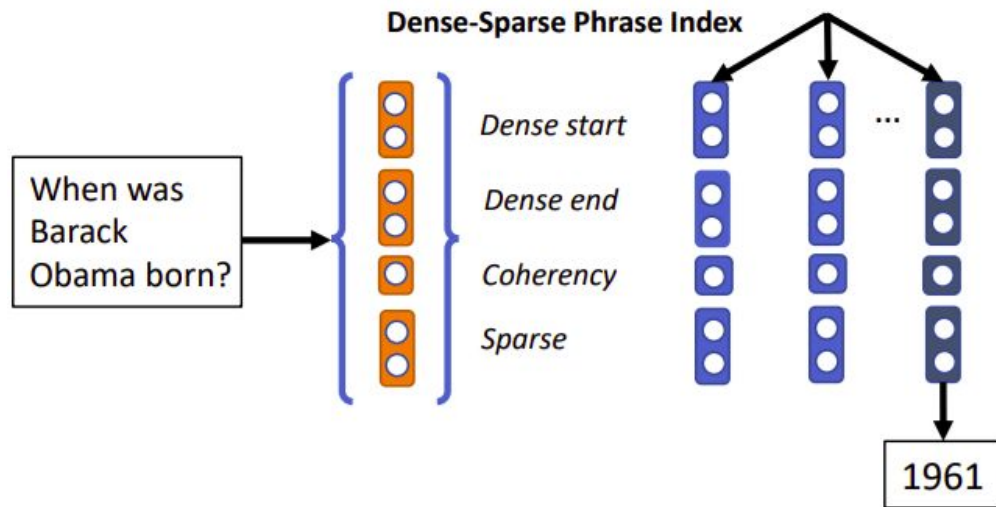
- It is possible to encode all the phrases (**60 billion phrases in Wikipedia**) using dense vectors and only do nearest neighbor search without a BERT model at inference time!



<https://github.com/princeton-nlp/DensePhrases>

# Open Domain Q.A. without reader...

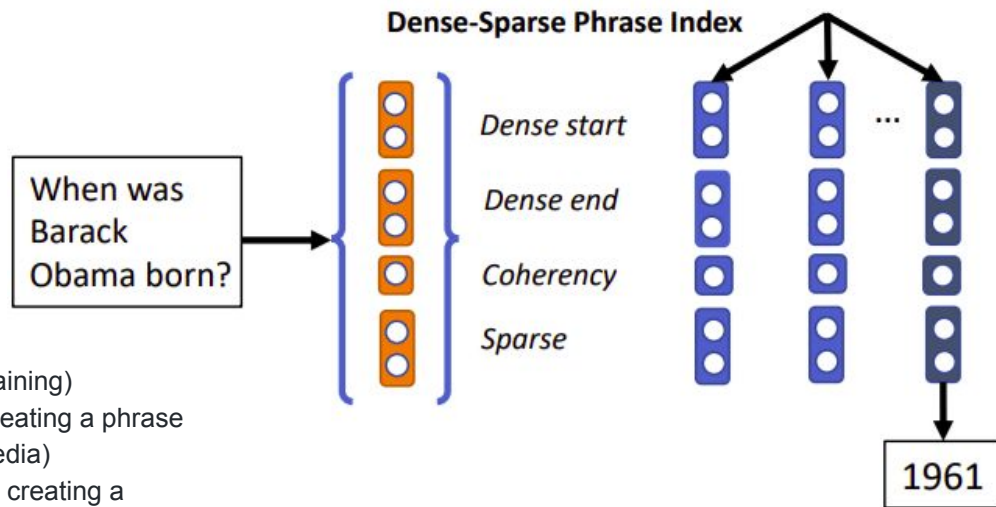
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- It is possible to encode all the phrases (**60 billion phrases in Wikipedia**) using dense vectors and only do nearest neighbor search without a BERT model at inference time!



<https://github.com/princeton-nlp/DensePhrases>

- Single 24GB GPU (for training)
- up to 150GB RAM (for creating a phrase index of the entire Wikipedia)
- up to 500GB storage (for creating a phrase dump of the entire Wikipedia)

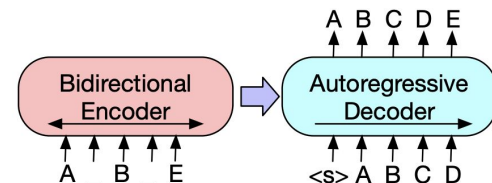
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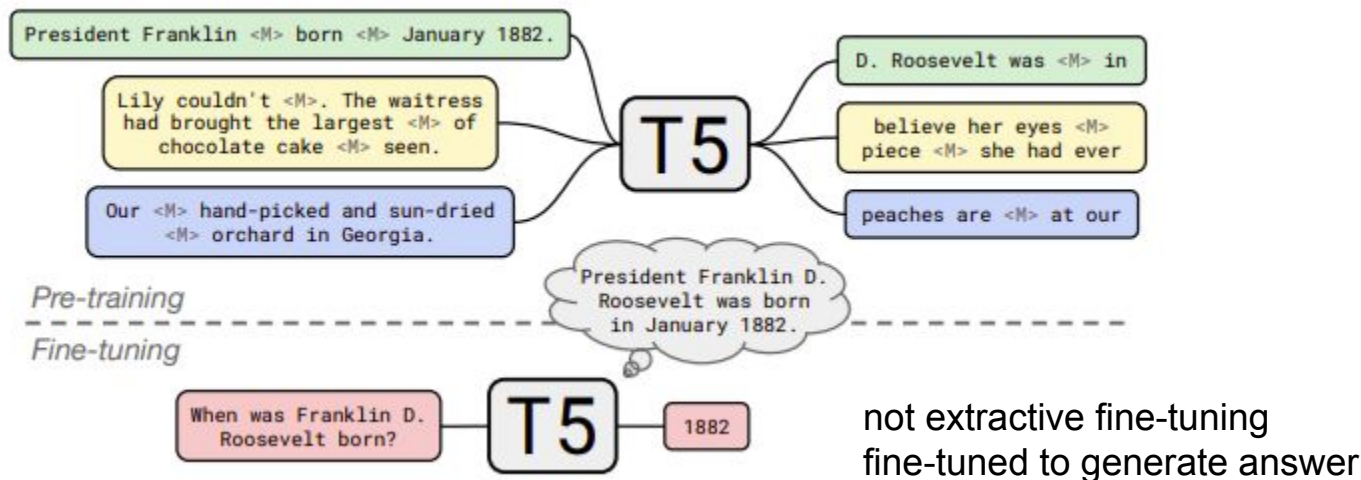
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    - Roberts et al., 2020. **How Much Knowledge Can You Pack Into the Parameters of a Language Model?**

# Open Domain Q.A. without retriever...



- How Much Knowledge Can You Pack Into the Parameters of a Language Model?

**T5** is pre-trained to fill in dropped-out spans of text



# Open Domain Q.A. without retriever...

How Much Knowledge  
Can You Pack Into the  
Parameters of a  
Language Model?

Salient span masking

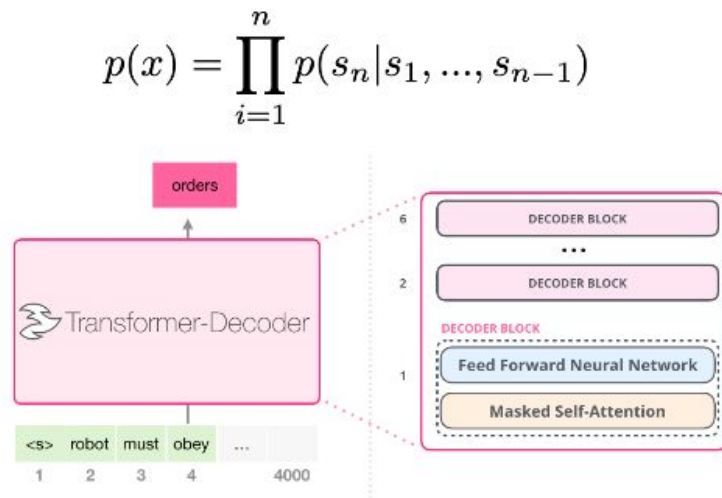
		NQ	WQ	TQA	
				dev	test
BM25 + BERT					
	Chen et al. (2017)	–	20.7	–	–
	Lee et al. (2019)	33.3	36.4	47.1	–
	Min et al. (2019a)	28.1	–	50.9	–
	Min et al. (2019b)	31.8	31.6	55.4	–
	Asai et al. (2019)	32.6	–	–	–
	Ling et al. (2020)	–	–	35.7	–
	Guu et al. (2020)	40.4	40.7	–	–
	Férvy et al. (2020)	–	–	43.2	53.4
	Karpukhin et al. (2020)	<b>41.5</b>	42.4	<b>57.9</b>	–
DPR	220M T5-Base	25.9	27.9	23.8	29.1
	770M T5-Large	28.5	30.6	28.7	35.9
	3B T5-3B	30.4	33.6	35.1	43.4
	11B T5-11B	32.6	37.2	42.3	50.1
	T5-11B + SSM	34.8	40.8	51.0	60.5
	T5.1.1-Base	25.7	28.2	24.2	30.6
	T5.1.1-Large	27.3	29.5	28.5	37.2
	T5.1.1-XL	29.5	32.4	36.0	45.1
	T5.1.1-XXL	32.8	35.6	42.9	52.5
	T5.1.1-XXL + SSM	35.2	<b>42.8</b>	51.9	<b>61.6</b>

# Open Domain Q.A. only L.M. **GPT2**

GPT2 is a very large  
transformer based  
language model trained  
on a massive dataset

48 layers, hidden size  
1600, 1.5B parameters

WebText: 8 million  
documents, excluding  
Wikipedia (!)



# Open Domain Q.A. **GPT2**

Evaluated on Natural Questions and **no training at all**

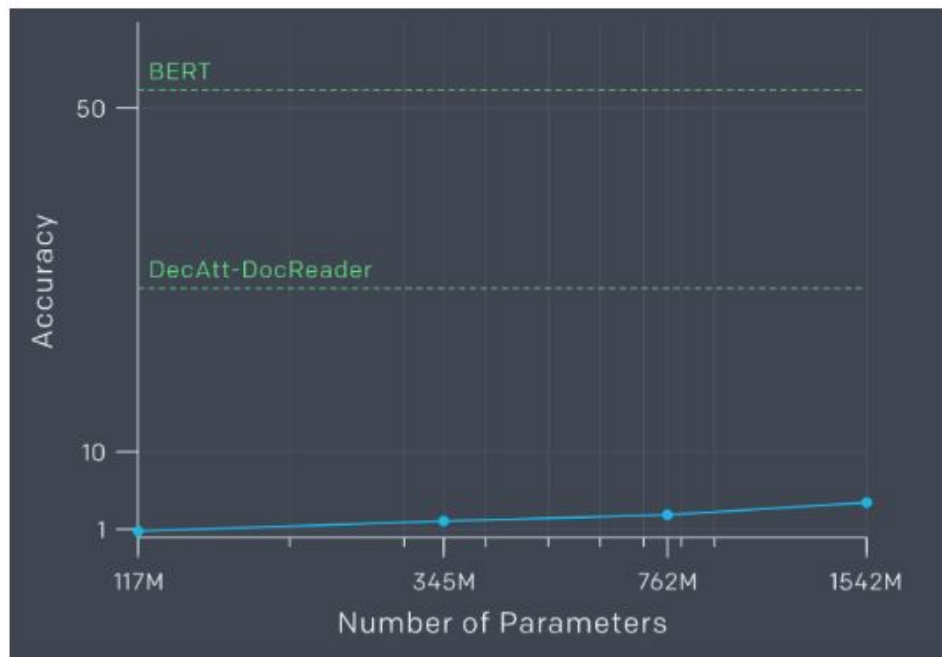
63.1% on the 1%  
of questions it is  
most confident in

Question	Generated Answer	Correct	Probability
Who wrote the book the origin of species?	Charles Darwin	✓	83.4%
Who is the founder of the ubuntu project?	Mark Shuttleworth	✓	82.0%
Who is the quarterback for the green bay packers?	Aaron Rodgers	✓	81.1%
Panda is a national animal of which country?	China	✓	76.8%
Who came up with the theory of relativity?	Albert Einstein	✓	76.4%
When was the first star wars film released?	1977	✓	71.4%
What is the most common blood type in sweden?	A	✗	70.6%
Who is regarded as the founder of psychoanalysis?	Sigmund Freud	✓	69.3%
Who took the first steps on the moon in 1969?	Neil Armstrong	✓	66.8%
Who is the largest supermarket chain in the uk?	Tesco	✓	65.3%
What is the meaning of shalom in english?	peace	✓	64.0%
Who was the author of the art of war?	Sun Tzu	✓	59.6%
Largest state in the us by land mass?	California	✗	59.2%
Green algae is an example of which type of reproduction?	parthenogenesis	✗	56.5%
Vikram samvat calender is official in which country?	India	✓	55.6%
Who is mostly responsible for writing the declaration of independence?	Thomas Jefferson	✓	53.3%
What us state forms the western boundary of montana?	Montana	✗	52.3%
Who plays ser davos in game of thrones?	Peter Dinklage	✗	52.1%
Who appoints the chair of the federal reserve system?	Janet Yellen	✗	51.5%
State the process that divides one nucleus into two genetically identical nuclei?	mitosis	✓	50.7%
Who won the most mvp awards in the nba?	Michael Jordan	✗	50.2%
What river is associated with the city of rome?	the Tiber	✓	48.6%
Who is the first president to be impeached?	Andrew Johnson	✓	48.3%
Who is the head of the department of homeland security 2017?	John Kelly	✓	47.0%
What is the name given to the common currency to the european union?	Euro	✓	46.8%
What was the emperor name in star wars?	Palpatine	✓	46.5%
Do you have to have a gun permit to shoot at a range?	No	✓	46.4%
Who proposed evolution in 1859 as the basis of biological development?	Charles Darwin	✓	45.7%
Nuclear power plant that blew up in russia?	Chernobyl	✓	45.7%
Who played john connor in the original terminator?	Arnold Schwarzenegger	✗	45.2%



# Open Domain Q.A. **GPT2**

Evaluated on Natural Questions and **no training at all**



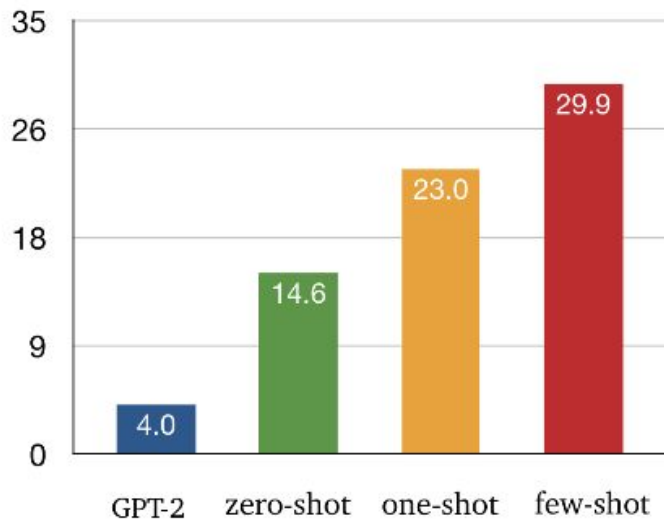
4% accuracy:  
Much much worse than  
supervised systems

# Open Domain Q.A. **GPT3** x100

96 layers, hidden  
size 12288, **175B**  
parameters

Larger corpora: Common  
Crawl + WebText + Books  
+ **English Wikipedia**

Evaluated on Natural Questions:



## Few-shot learner

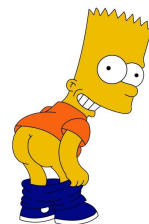
- No weight updates
- $Q_1, A_1, Q_2, A_2, \dots, Q_K, A_K, Q ?$
- One-shot setting is a special case when only **one** example is given.

# Last words

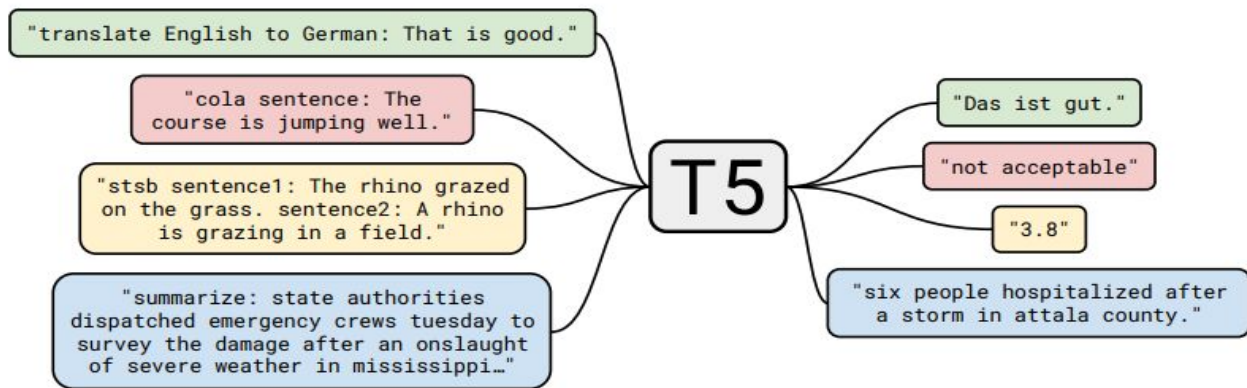
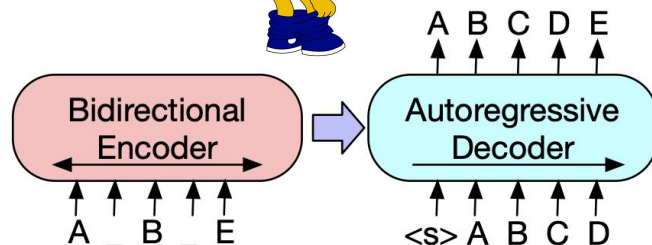
- N.L.P. -> language generation task?

- Pretrained GPT, BART or T5 models for:

- Question Answering
    - Named Entity Disambiguation
    - Information extraction
    - Machine Translation\*
    - Text Classification
    - ...



mBART

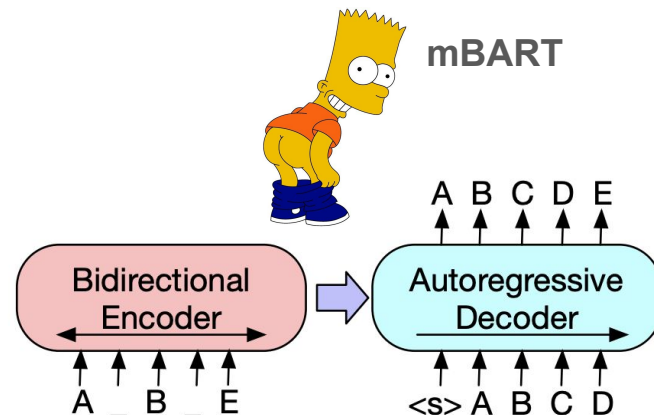


# Last words

- N.L.P. -> language generation task?

- Pretrained GPT, BART or T5 models for:

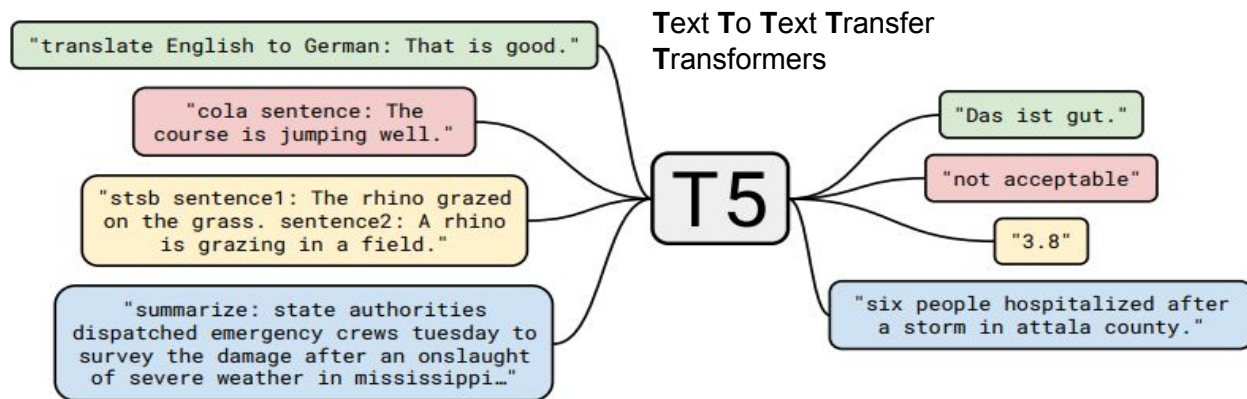
- Question Answering
- Named Entity Disambiguation
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- Machine Translation\*
- Text Classification
- ...



R.I.P.

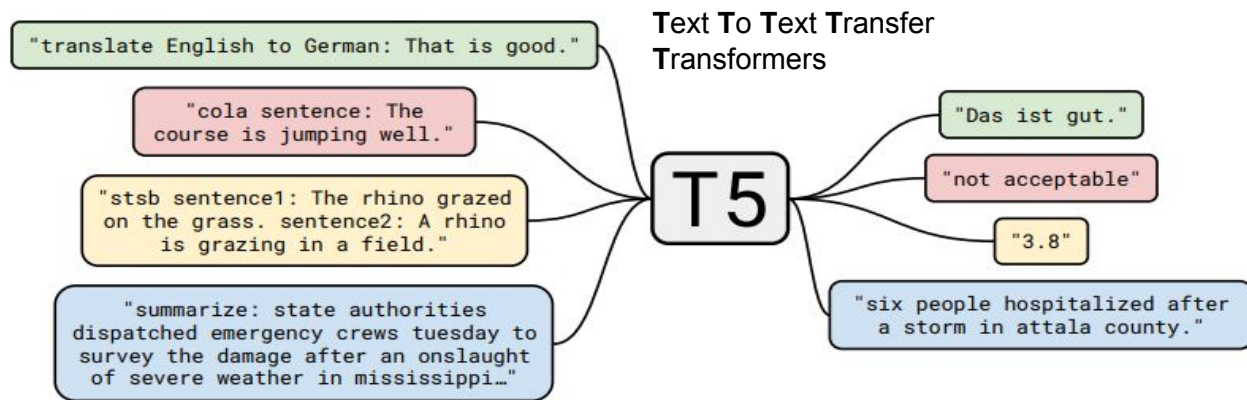
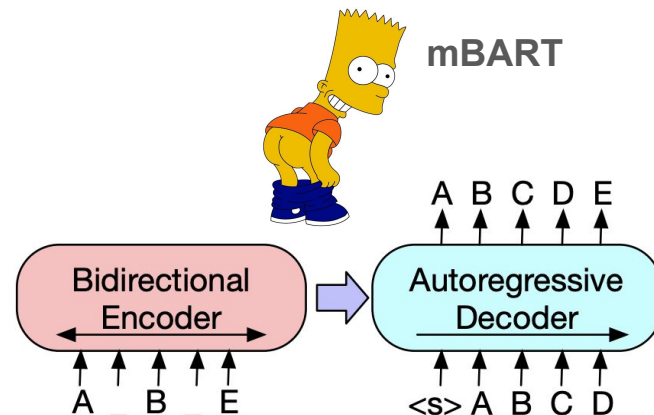


??



# Last words

- N.L.P. -> language generation task?
  - Pretrained GPT, BART or T5:
    - The performance is largely impacted by the model size.
    - A 11B T5 model = A 330M DPR BERT



# Last words

- N.L.P. -> language generation task?
  - Pretrained GPT, BART or T5:
    - The performance is largely impacted by the model size.
    - A 11B T5 model = A 330M DPR BERT
    - ...
    - GPTneox, FLAN, instructGPT

## Aligning Language Models to Follow Instructions:

Ouyang et al. 2022 *Training language models to follow instructions with human feedback*

Our labelers prefer outputs from our 1.3B **InstructGPT** model over outputs from a 175B **GPT-3** model, despite having more than 100x fewer parameters.

R.I.P.



??

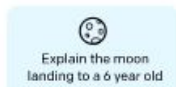
<https://openai.com/blog/instruction-following/>

# Aligning Language Models to Follow Instructions

## Step 1

**Collect demonstration data, and train a supervised policy.**

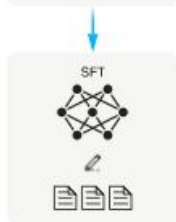
A prompt is sampled from our prompt dataset.



A labeler demonstrates the desired output behavior.



This data is used to fine-tune GPT-3 with supervised learning.



## Step 2

**Collect comparison data, and train a reward model.**

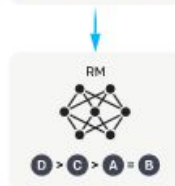
A prompt and several model outputs are sampled.



A labeler ranks the outputs from best to worst.



This data is used to train our reward model.



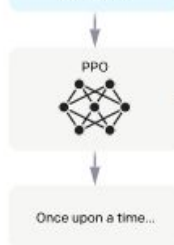
## Step 3

**Optimize a policy against the reward model using reinforcement learning.**

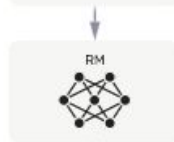
A new prompt is sampled from the dataset.



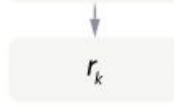
The policy generates an output.



The reward model calculates a reward for the output.



The reward is used to update the policy using PPO.



# Aligning Language Models to Follow Instructions

PROMPT    *Q: Who was president of the United States in 1955?*  
*A: Dwight D. Eisenhower was president of the United States in 1955.*

*Q: How does a telescope work?*  
*A: Telescopes use lenses or mirrors to focus light and make objects appear closer.*

*Q: Why do birds migrate south for the winter?*  
*A:*

COMPLETION    GPT-3  
Birds migrate south for the winter because the weather is colder and there is less food available.

InstructGPT  
Birds migrate south for the winter because it is warmer there.

<https://beta.openai.com/playground>  
<https://openai.com/blog/instruction-following/>

**instructGPT** -> text-davinci-002



# Aligning Language Models to Follow Instructions

Q: Who was president of the United States in 1955? **Prompt**

A: Dwight D. Eisenhower was president of the United States in 1955.

Q: How does a telescope work? **Prompt**

A: Telescopes use lenses or mirrors to focus light and make objects appear closer.

Q: Why is still BERT performing better than GPT3?

A:

Q: How many minutes does the Master of Puppets song last?

A:

Q: Who is better Goku or Vegeta?

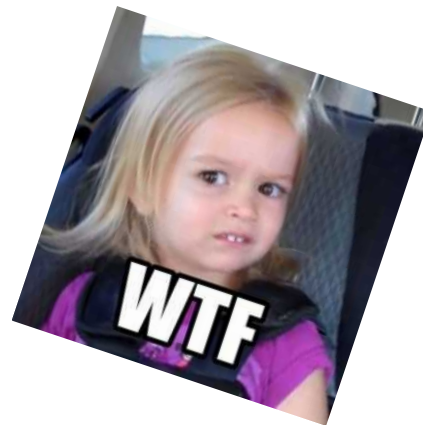
A:

Q:  $345+111$ ?

A:

Q: Who are you? Skynet?

A:



# Aligning Language Models to Follow Instructions

Q: Who was president of the United States in 1955? **Prompt**

A: Dwight D. Eisenhower was president of the United States in 1955.

Q: How does a telescope work? **Prompt**

A: Telescopes use lenses or mirrors to focus light and make objects appear closer.

Q: Why is still BERT performing better than GPT3?

A:

Q: How many minutes does the Master of Puppets song last?

A:

Q: Who is better Goku or Vegeta?

A:

Q:  $345+111$ ?

A:

Q: Who are you? Skynet?

A:

Not only Q.A.

-> read a file line by line and store it in a list in python

-> read a file line by line and print it in python

-> Spaghetti boloñesa recipe



# Thanks!

ander.barrena@ehu.eus  
@4nderB