



Stefano Fiorucci

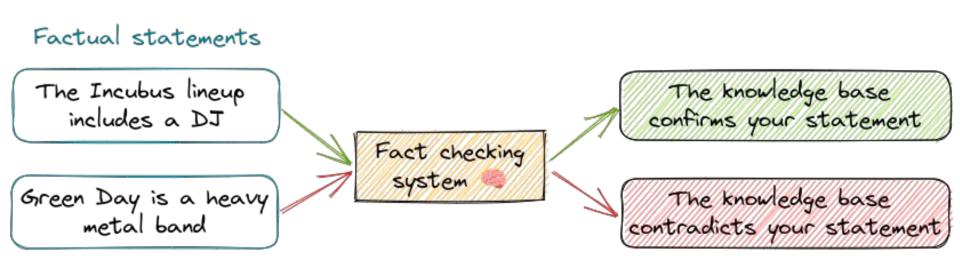
- Machine Learning Engineer (NLP)

Semantic / Neural / Vector
 Search o explorer

Find me on:

Github/HF (@anakin87) - LinkedIn

Fact Checking & Rocks!



- knowledge from Wikipedia
- modern NLP tools
- zero manual effort



Rock fact checker



Info

Fact Checking & Rocks!

Fact checking baseline combining dense retrieval and textual entailment

Github project - Based on Haystack

Data crawled from Wikipedia

Fact Checking & Rocks!

Enter a factual statement about Rock music and let the AI check it out for you...



The knowledge base seems to contradict your statement

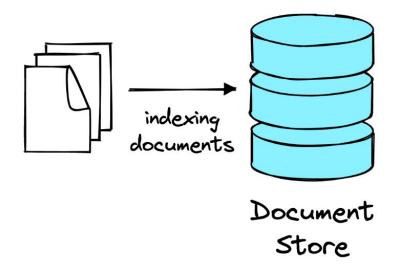
Agenda

Haystack basics Fact Checking Rocks: the idea Fact Checking Rocks: the implementation **Conclusions**



- end-to-end open-source NLP framework
- build
 - search systems
 - (Large) Language Models applications
- modular
- scalable
- customizable

Haystack: Document stores

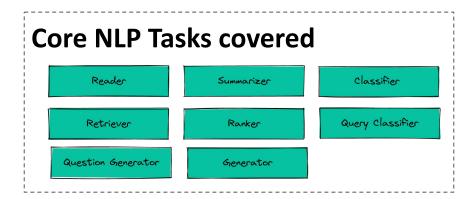


Haystack: Nodes

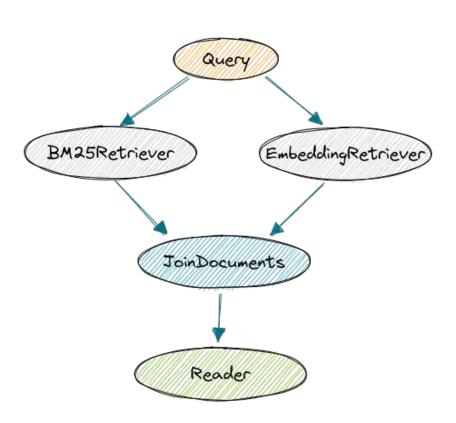
Data Connectors



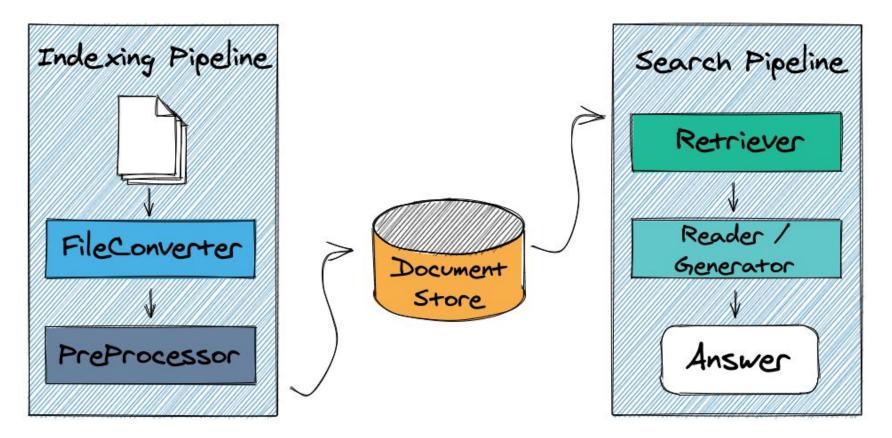
- File converters from txt, pdf, docx, markdown
- Web scraper to turn website into text
- Preprocessor to split long documents, clean text



Haystack: Pipelines



Haystack: Question Answering architecture

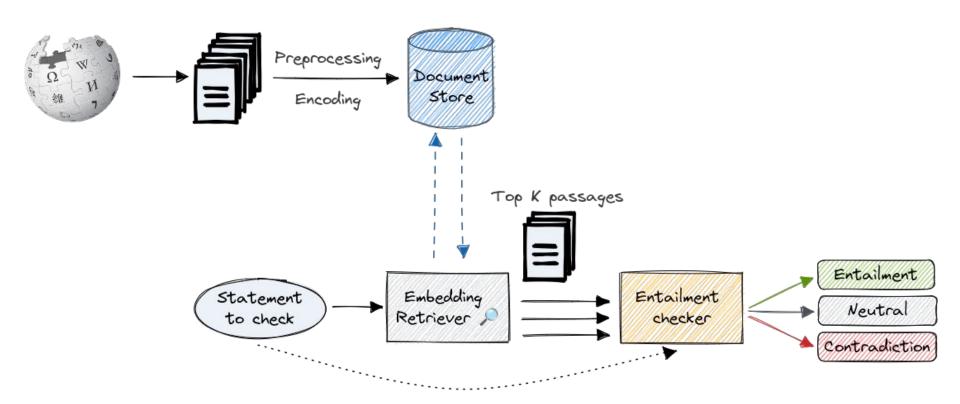


Reading Wikipedia to Answer Open-Domain Questions (2017, machine reading at scale)

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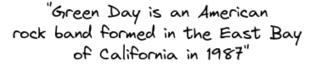
Fact Checking Rocks: the idea



Embedding Retriever

Query

"Green Day is a punk rock band"



"Green Day are a punk band, but you know, punk is the legacy of rock and roll, and they are the biggest band in the genre."

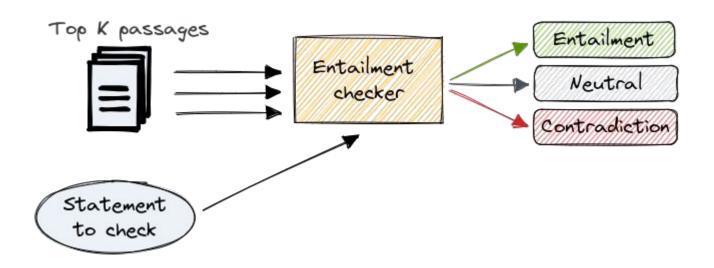
"Buckley's cover of Hallelujah was ranked No. 259 of the 500 Greatest Songs by Rolling Stone in 2004."

Natural Language Inference models

NLI: the task of determining whether a "hypothesis" is true, false, or undetermined given a "premise".

Premise	Hypothesis	Label
A man inspects the uniform of a figure in some East Asian country.	The man is sleeping.	contradiction
An older and younger man smiling.	Two men are smiling and laughing at the cats playing on the floor.	neutral
A soccer game with multiple males playing.	Some men are playing a sport.	entailment

Entailment Checker node



- compute the textual entailment for every passage
- aggregate the textual entailment scores
- empirical consideration: apply a threshold

Agenda

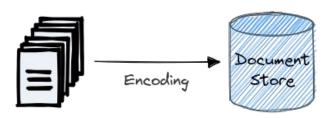
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Load and preprocess data



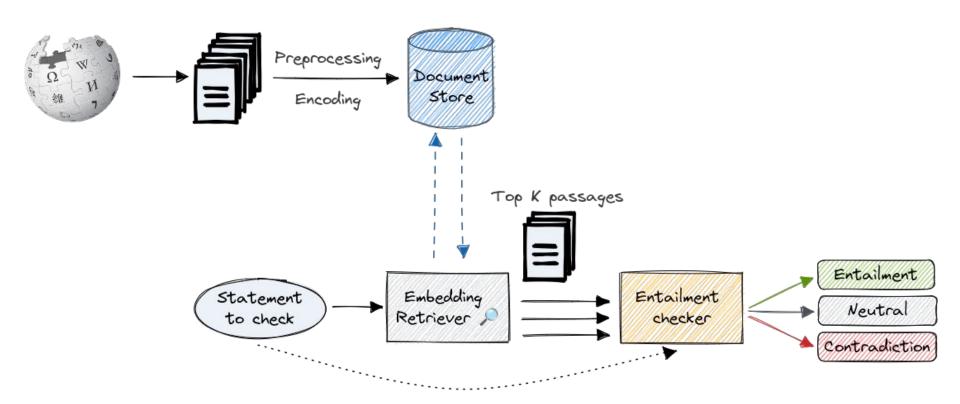
```
import glob, json
from haystack.nodes import PreProcessor
docs = []
for json_file in glob.glob("rock_wiki/*.json"):
   with open(json_file, "r") as fin:
       doc = json.load(fin)
    docs.append(doc)
# preprocess documents, splitting by chunks of 2 sentences
processor = PreProcessor(
    clean_empty_lines=True,
    clean_whitespace=True.
    clean_header_footer=True,
    split_by="sentence",
    split_length=2,
    split_respect_sentence_boundary=False,
    split_overlap=0,
    language="en",
preprocessed_docs = processor.process(docs)
# select only documents with at least 10 words.
# Otherwise, the documents are not very informative
preprocessed_docs = [doc for doc in preprocessed_docs
                     if len(doc.content.split()) >= 10]
```

2. Encode and write documents



```
from haystack.document_stores import FAISSDocumentStore
from haystack.nodes import EmbeddingRetriever
# the document store settings are those compatible with Embedding Retriever
document_store = FAISSDocumentStore(similarity="dot_product",
                                    embedding dim=768)
# write documents
document_store.write_documents(preprocessed_docs)
retriever = EmbeddingRetriever
    document store=document store,
    embedding_model="sentence-transformers/msmarco-distilbert-base-tas-b",
    model format="sentence transformers",
    embed_meta_fields=["name"],
# generate embeddings
document store.update embeddings(retriever)
```

Fact Checking Rocks: the idea



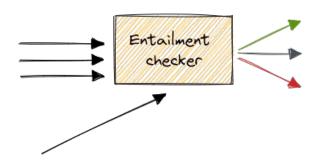
Adding custom nodes to Haystack

From Haystack docs:

- Create a new class that inherits from BaseComponent
- 2. Set outgoing_edges as a class attribute
- 3. Define a run() method, which returns a tuple
- Define a run_batch()
 method, which returns a
 tuple

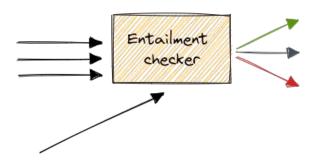
```
from haystack.nodes.base import BaseComponent
class NodeTemplate(BaseComponent):
    # If it's not a decision node, there is only one outgoing edge
    outgoing edges = 1
    def run(self, query: str, my_arg: Optional[int] = 10):
       # Insert code here to manipulate the input
       # and produce an output dictionary
        output={
            "documents": ...,
            " debug": {"anything": "you want"}
        return output, "output_1"
    def run_batch(self, queries: List[str], my_arg: Optional[int] = 10):
        # Insert code here to manipulate the input
        # and produce an output dictionary
        output={
            "documents": ...,
        return output, "output 1"
```

3. Entailment Checker node: *init*



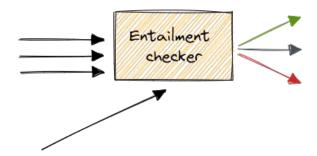
```
from typing import List, Optional
from transformers import AutoModelForSequenceClassification, AutoTokenizer, AutoConfig
import torch
from haystack.nodes.base import BaseComponent
from haystack.modeling.utils import initialize_device_settings
from haystack.schema import Document
class EntailmentChecker(BaseComponent):
    outgoing_edges = 1
    def __init__(
        self.
        model_name_or_path: str = "roberta-large-mnli",
        model_version: Optional[str] = None,
        tokenizer: Optional[str] = None,
        use apu: bool = True,
        batch size: int = 16,
        entailment contradiction threshold: float = 0.5,
    ):
        super().__init__()
        self.devices, = initialize device settings(use cuda=use gpu, multi gpu=False)
        tokenizer = tokenizer or model name or path
        self.tokenizer = AutoTokenizer.from pretrained(tokenizer)
        self.model = AutoModelForSequenceClassification.from pretrained(
            pretrained model name or path=model name or path, revision=model version
        self.batch size = batch size
        self.entailment_contradiction_threshold = entailment_contradiction_threshold
        self.model.to(str(self.devices[0]))
        id2label = AutoConfig.from_pretrained(model_name_or_path).id2label
        self.labels = [id2label[k].lower() for k in sorted(id2label)]
        if "entailment" not in self.labels:
            raise ValueError(
                "The model config must contain entailment value in the id2label dict."
```

3. Entailment Checker node: get_entailment



```
def get entailment(self, premise, hypotesis):
   with torch.inference mode():
        inputs = self.tokenizer(
            f"{premise}{self.tokenizer.sep_token}{hypotesis}",
          return_tensors="pt"
        ).to(self.devices[0])
       out = self.model(**inputs)
        logits = out.logits
        probs = (
            torch.nn.functional.softmax(logits, dim=-1)[0, :]
          .detach().cpu().numpy()
    entailment_dict = {k.lower(): v for k, v in zip(self.labels, probs)}
    return entailment dict
```

3. Entailment Checker node: run



```
def run(self, query: str, documents: List[Document]):
    scores, agg con, agg neu, agg ent = 0, 0, 0, 0
    for i, doc in enumerate(documents):
        entailment info = self.get entailment(premise=doc.content,
                                              hypotesis=query)
        doc.meta["entailment info"] = entailment info
        scores += doc.score
        con, neu, ent = (
            entailment info["contradiction"],
            entailment info["neutral"],
            entailment info["entailment"],
        agg con += con * doc.score
        agg_neu += neu * doc.score
        agg_ent += ent * doc.score
        # if in the first documents there is a strong evidence
        # of entailment/contradiction, there is no need to
        # consider less relevant documents
        if max(agg_con, agg_ent) / scores > self.entailment_contradiction_threshold:
            break
    aggregate entailment info = {
        "contradiction": round(agg_con / scores, 2),
        "neutral": round(agg_neu / scores, 2),
        "entailment": round(agg_ent / scores, 2),
    entailment_checker_result = {
        "documents": documents[: i + 1],
        "aggregate_entailment_info": aggregate_entailment_info,
    return entailment checker result, "output 1"
```

4. Build the fact-checking pipeline

```
from haystack.pipelines import Pipeline
from app_utils.entailment checker import EntailmentChecker
entailment checker = EntailmentChecker(
    model name or path="microsoft/deberta-v2-xlarge-mnli",
    use_gpu=False,
    entailment contradiction threshold=0.5,
pipe = Pipeline()
pipe.add_node(component=retriever,
              name="retriever",
              inputs=["Query"])
pipe.add_node(component=entailment_checker,
              name="ec",
              inputs=["retriever"])
```

```
results = pipe.run(
  query="The Smiths have had an influence on other bands",
params={"retriever": {"top_k": 5}})
print(results)
```

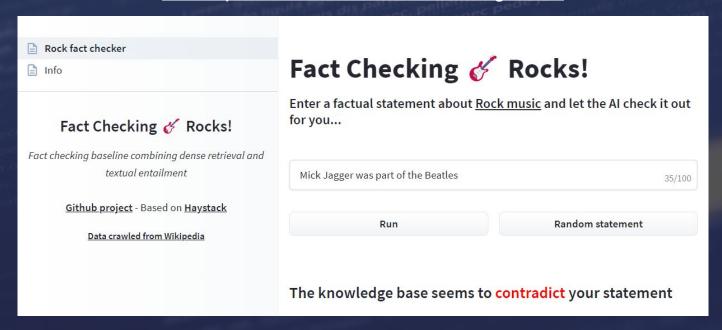
5. Let's try the pipeline!

```
{'documents':

Cocument:
 {'content': "Morrissey's songwriting was influenced by punk rock and post-punk bands such as New York Dolls, the Cramps,
the Specials and the Cult, along with 1960s girl groups and singers such as Dusty Springfield, Sandie Shaw, Marianne
Faithfull and Timi Yuro. Morrissey's lyrics, while superficially depressing, were often full of mordant humour; John Peel
remarked that the Smiths were one of the few bands capable of making him laugh out loud.",
   'score': 0.737,
   'meta': {'name': 'The Smiths',
            'entailment info': {'contradiction': 0.00, 'neutral': 0.95, 'entailment': 0.05},
           ..., ...}>,
  <Document:</pre>
 {'content': 'Singer Davey Havok of the band AFI cites the Smiths as an influence.Q magazine\'s Simon Goddard argued in
2007 that the Smiths were "the one truly vital voice of the \'80s" and "the most influential British guitar group of the
decade". He continued: "As the first indie outsiders to achieve mainstream success on their own terms (their second album
proper, 1985\'s Meat Is Murder, made Number 1 in the UK), they elevated rock\'s standard four-piece formula to new heights
of magic and poetry.',
   'score': 0.737,
   'meta': {'name': 'The Smiths',
            'entailment info': {'contradiction': 0.00, 'neutral': 0.17, 'entailment': 0.83},
           ...}, ...}>, ...],
 'aggregate_entailment_info': {'contradiction': 0.00, 'neutral': 0.48, 'entailment': 0.52}, ...}
```

Demo time!

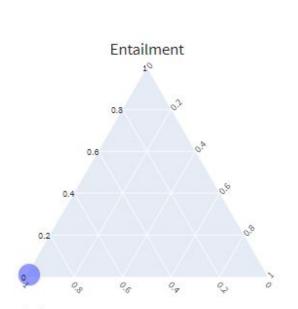
hf.co/spaces/anakin87/fact-checking-rocks





The knowledge base seems to contradict your statement

Aggregate entailment information:



```
"contradiction": 0.99
"neutral": 0.01
"entailment": 0.01
```

Most Relevant snippets:

	Title	Relevance	con	neu	ent	Content
0	Elvis Presley	0.742	0.99	0.01	0.01	Elvis Aaron Presley (January 8, 1935 – August 16, 1977), or simpl

Contradiction Neutral

Wikipedia source pages: Elvis Presley



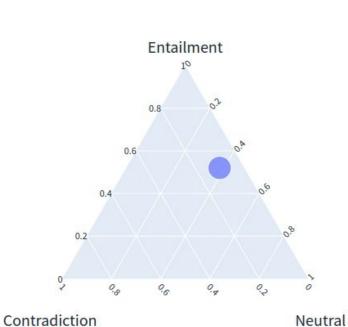




The Beach Boys were involved with the Manson family

The knowledge base seems to confirm your statement

Aggregate entailment information:



```
"contradiction": 0.1
"neutral": 0.38
"entailment": 0.52
```

Most Relevant snippets:

	Title	Relevance	con	neu	ent	Content
0	The Beach Boys	0.749	0.01	0.61	0.39	According to Jon Parks, the band's tour man
1	The Beach Boys	0.747	0.42	0.58	0.00	According to Leaf, "The entire Wilson family
2	The Beach Boys	0.740	0.00	0.08	0.92	Drawing from the Beach Boys' associations
3	The Beach Boys	0.737	0.06	0.41	0.53	Dennis then proposed that Manson be signe
4	The Beach Boys	0.736	0.01	0.23	0.77	In June 1968, Dennis befriended Charles Ma

Agenda

Haystack basics 2 Fact Checking Rocks: the idea Fact Checking Rocks: the implementation **Conclusions**

Fact Checking Rocks: limitations 1



- no statement detection
- Wikipedia is taken as a source of truth
- no guarantee that the best text passages for NLI emerge from semantic similarity
- no organic evaluation was performed

Fact Checking Rocks: how to improve \Rightarrow



expand the knowledge base

 adapt the retriever to the domain, using **Generative** Pseudo Labelling

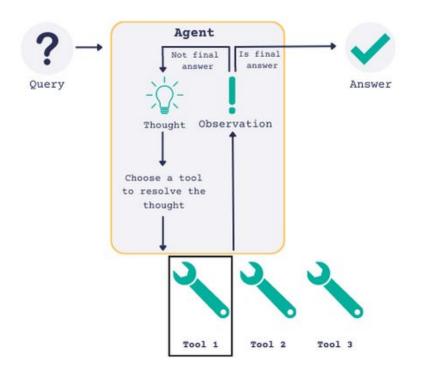
Building NLP applications

 (Large) Language Models have strong text comprehension/generation abilities

- Their knowledge is generic and not easily updated over time
- Combine them with P Retrieval systems!

LLMs in Haystack: PromptNode and Agents (2)





https://github.com/anakin87/try-agents-haystack

https://docs.haystack.deepset.ai/docs/agent



https://haystack.deepset.ai

https://discord.gg/haystack

hf.co/spaces/anakin87/fact-checking-rocks
github.com/anakin87/fact-checking-rocks



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