Project 2 - Insurance

Load Data

```
dfs = []
data_folder_path = os.path.join('assurance')
for file_name in os.listdir(data_folder_path):
    path = os.path.join(data_folder_path, file_name)
    new_df = pd.read_excel(path)
    dfs.append(new_df)
df = pd.concat(dfs, ignore_index=True)
```

assurance avis_1_traduit.xlsx avis_2_traduit.xlsx avis_3_traduit.xlsx avis_4_traduit.xlsx avis_5_traduit.xlsx avis_6_traduit.xlsx avis_7_traduit.xlsx avis_8_traduit.xlsx avis_9_traduit.xlsx avis_10_traduit.xlsx avis_11_traduit.xlsx avis_12_traduit.xlsx avis_13_traduit.xlsx avis_14

Preprocess

```
# Select
drop colmun = [
  'auteur', 'produit', 'date_publication', 'date_exp',
  'avis cor', 'avis cor en']
df.drop(drop_colmun, axis=1, inplace=True)
df.dropna(subset=['avis_en'], inplace=True)
df['note'] = df['note'].fillna(0).astype(int)
# Rename columns
df.rename(columns={'avis': 'avis_fr', 'assureur': 'insurer'}, inplace=True)
df = df.reindex(columns=['insurer', 'avis_fr', 'avis_en','note', 'type']) # no agr inplace
# Split df
df_train = df[df['type'] == 'train'].drop(columns=['type'])
                                                              # 24 103 lines
df_test = df[df['type'] == 'test'].drop(columns=['type', 'note']) # 10 330lines
```

Sentiment Analysis

Choose tabularisai/multilingual-sentiment-analysis because

- multilingual so it's the same model for the review in french and english
- return an integer score [0, 4] => translation to [1, 5] scale like stars

```
from transformers import AutoTokenizer, AutoModelForSequenceClassification
senti_model_name = "tabularisai/multilingual-sentiment-analysis"
senti_tokenizer = AutoTokenizer.from_pretrained(senti_model_name)
senti_model = AutoModelForSequenceClassification.from_pretrained(
  senti_model_name).to(device)
senti_model = senti_model.eval()
def sentiment_pipeline(texts):
  inputs = senti_tokenizer(
    texts, return_tensors="pt", truncation=True, padding=True, max_length=512
  ).to(device)
  with torch.no_grad():
    outputs = senti_model(**inputs)
    probabilities = torch.nn.functional.softmax(outputs.logits, dim=-1)
    sentiment = torch.argmax(probabilities, dim=-1).tolist()
  senti_rescale = [int(senti+1) for senti in sentiment] # convert to 1-5 scale like stars
  return senti rescale
```

Zero-shot

Average distance with ground-truth

• Language fr: 0.79

• Language en: 0.79

Fine-tune on df_train

Didn't succeed

Subject Classifier

Choose cross-encoder/nli-deberta-v3-base because

- Finetune of the model microsoft/deberta-v3-base on **SNLI** and **MultiNLI** datasets, so it's also multilingual
- Return a score of probability devide between the different possible label

Pipeline

split_long_reviews

- 1. Split review in smaller chunks
- 2. Store the chunks with the other reviews
- 3. Keep a log of the original index of each review/subreview on id_map

merdge_splited_review

- Merdge chunks with the same index in id_map
- 2. Compute average of each label if there is a merdge
- 3. Return the best label with it score
- 4. If the score is under a threshold, return 'Other'

```
from transformers import pipeline
from numpy import argmax
classifier = pipeline("zero-shot-classification",
  model='cross-encoder/nli-deberta-v3-base',
  use fast = False, device=device
def subject_pipeline(reviews, lang='fr', threshold=0.5):
  # Pre-process
  labels = labels fr if lang == 'fr' else labels en
  split reviews, id map = split long reviews(reviews, 500)
  # Run model (exclude 'Others' label)
  resp = classifier(split reviews, labels[:-1])
  # Post process
  subjects, scores = merdge splited review(resp, id map)
  # Classify as 'Others' under thresholds
  for i in range(len(scores)):
    if scores[i] <= threshold:</pre>
      subjects[i] = labels[-1]
  return subjects
```

Summarize reviews

```
from transformers import pipeline
summarizer = pipeline("summarization",
  model="Falconsai/text summarization",
  device=device
def summarize reviews(summary text, lang='fr', input length max=512,
output length max=300):
  nb loop = 0
  print(f'Original length : {len(summary text)} lines')
  print('Nb loop \t Summary length')
  # Summarize and merge until we have only one summary
  while len(summary text) > 1:
    summary merdge = join reviews(summary text, input length max)
    summary resp = summarizer(summary merdge,
      max length=output length max,
      min length=50,
      do_sample=False
    summary_text = [r['summary_text'] for r in summary_resp]
  return summary_text[0]
```

join_reviews

Merge summaries till get "super review" of length just below input_length_max

If some reviews are already taller, apply a recursion on it

summarizer

Summarize the reviews till length [output_length_max, 50]

Do it while we have one summary

Preprocess data test

For both language:

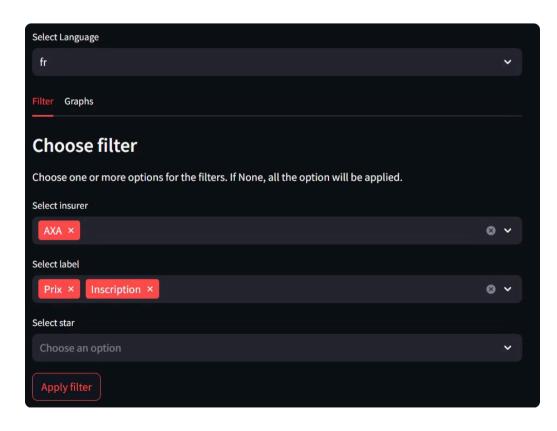
- Apply both sentiment_pipeline and subject_pipeline to the whole dataset
- Remove the reviews for the other language
- Save it as csv



Streamlit Filter

```
def dataframe_filter(df_input, insurer=None, label=None, star=None):
    df_filter = df_input.copy()
    filter_dict = {'insurer': insurer, 'label': label, 'star': star }

for filter_name, filter in filter_dict.items():
    if filter not in [None, []]:  # if str or [str]
        filter = [filter] if type(filter) == str else filter  # convert str to [str]
        df_filter = df_filter[df_filter[filter_name].isin(filter)]
return df_filter
```





Print the result

Ask for the filter

Streamlit Graph

