#### Install !pip install pingouin

!pip install pingouin Collecting pingouin Downloading pingouin-0.5.3-py3-none-any.whl (198 kB) - 198.6/198.6 kB 2.4 MB/s eta 0:00:00 Requirement already satisfied: numpy>=1.19 in /usr/local/lib/python3.10/dist-packages (from pingouin) (1.23.5) Requirement already satisfied: scipy>=1.7 in /usr/local/lib/python3.10/dist-packages (from pingouin) (1.11.4) Requirement already satisfied: pandas>=1.0 in /usr/local/lib/python3.10/dist-packages (from pingouin) (1.5.3) Requirement already satisfied: matplotlib>=3.0.2 in /usr/local/lib/python3.10/dist-packages (from pingouin) (3.7.1) Requirement already satisfied: seaborn>=0.11 in /usr/local/lib/python3.10/dist-packages (from pingouin) (0.12.2) Requirement already satisfied: statsmodels>=0.13 in /usr/local/lib/python3.10/dist-packages (from pingouin) (0.14.0) Requirement already satisfied: scikit-learn in /usr/local/lib/python3.10/dist-packages (from pingouin) (1.2.2) Collecting pandas-flavor>=0.2.0 (from pingouin) Downloading pandas flavor-0.6.0-pv3-none-anv.whl (7.2 kB) Collecting outdated (from pingouin) Downloading outdated-0.2.2-py2.py3-none-any.whl (7.5 kB) Requirement already satisfied: tabulate in /usr/local/lib/python3.10/dist-packages (from pingouin) (0.9.0) Requirement already satisfied: contourpy>=1.0.1 in /usr/local/lib/python3.10/dist-packages (from matplotlib>=3.0.2->pingouin) (1.2.0 Requirement already satisfied: cycler>=0.10 in /usr/local/lib/python3.10/dist-packages (from matplotlib>=3.0.2->pingouin) (0.12.1) Requirement already satisfied: fonttools>=4.22.0 in /usr/local/lib/python3.10/dist-packages (from matplotlib>=3.0.2->pingouin) (4.45 Requirement already satisfied: kiwisolver>=1.0.1 in /usr/local/lib/python3.10/dist-packages (from matplotlib>=3.0.2->pingouin) (1.4. Requirement already satisfied: packaging>=20.0 in /usr/local/lib/python3.10/dist-packages (from matplotlib>=3.0.2->pingouin) (23.2) Requirement already satisfied: pillow>=6.2.0 in /usr/local/lib/python3.10/dist-packages (from matplotlib>=3.0.2->pingouin) (9.4.0) Requirement already satisfied: pyparsing>=2.3.1 in /usr/local/lib/python3.10/dist-packages (from matplotlib>=3.0.2->pingouin) (3.1.1 Requirement already satisfied: python-dateutil>=2.7 in /usr/local/lib/python3.10/dist-packages (from matplotlib>=3.0.2->pingouin) (2 Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.10/dist-packages (from pandas>=1.0->pingouin) (2023.3.post1) Requirement already satisfied: xarray in /usr/local/lib/python3.10/dist-packages (from pandas-flavor>=0.2.0->pingouin) (2023.7.0) Requirement already satisfied: patsy>=0.5.2 in /usr/local/lib/python3.10/dist-packages (from statsmodels>=0.13->pingouin) (0.5.3) Requirement already satisfied: setuptools>=44 in /usr/local/lib/python3.10/dist-packages (from outdated->pingouin) (67.7.2) Collecting littleutils (from outdated->pingouin) Downloading littleutils-0.2.2.tar.gz (6.6 kB) Preparing metadata (setup.py) ... done Requirement already satisfied: requests in /usr/local/lib/python3.10/dist-packages (from outdated->pingouin) (2.31.0) Requirement already satisfied: joblib>=1.1.1 in /usr/local/lib/python3.10/dist-packages (from scikit-learn->pingouin) (1.3.2) Requirement already satisfied: threadpoolctl>=2.0.0 in /usr/local/lib/python3.10/dist-packages (from scikit-learn->pingouin) (3.2.0) Requirement already satisfied: six in /usr/local/lib/python3.10/dist-packages (from patsy>=0.5.2->statsmodels>=0.13->pingouin) (1.16 Requirement already satisfied: charset-normalizer<4.>=2 in /usr/local/lib/python3.10/dist-packages (from requests->outdated->pingoui Requirement already satisfied: idna<4,>=2.5 in /usr/local/lib/python3.10/dist-packages (from requests->outdated->pingouin) (3.6) Requirement already satisfied: urllib3<3.>=1.21.1 in /usr/local/lib/python3.10/dist-packages (from requests->outdated->pingouin) (2. Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.10/dist-packages (from requests->outdated->pingouin) (20 Building wheels for collected packages: littleutils

```
Building wheel for littleutils (setup.py) ... done
      Created wheel for littleutils: filename=littleutils-0.2.2-py3-none-any.whl size=7026 sha256=a5fc2c20675a7105873cc542eb624c6734081e
      Stored in directory: /root/.cache/pip/wheels/3d/fe/b0/27a9892da57472e538c7452a721a9cf463cc03cf7379889266
    Successfully built littleutils
    Installing collected packages: littleutils, outdated, pandas-flavor, pingouin
    Successfully installed littleutils-0.2.2 outdated-0.2.2 pandas-flavor-0.6.0 pingouin-0.5.3
import pandas as pd
import pingouin as pg
#Annotater Shubham
df 1 = pd.read excel('/content/CounselChat-EDA (1).xlsx',sheet name='Training Sheet Shubham')
Double-click (or enter) to edit
#Annotater Mustafa
df 2 = pd.read excel('/content/CounselChat-EDA (1).xlsx',sheet name='Training Sheet Mustafa')
from google.colab import drive
drive.mount('/content/drive')
    Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force remount=True).
```

## Annotation for Actionability

Based on 3 dimensions Clarity, Specificity and Explainability

```
#Annotater Gopika
df_3 = pd.read_excel('/content/CounselChat-EDA (1).xlsx',sheet_name='Gopika - Training Sheet Main')
#Annotater Ali
df_4 = pd.read_excel('/content/CounselChat-EDA (1).xlsx',sheet_name='Ali - Training Sheet Main')
```

#### Clean Data

```
df_1 = df_1[['id','Warmth','Concern','Acknowledgement']]
df_2 = df_2[['id','Warmth','Concern','Acknowledgement']]
df_3 = df_3[['id','Clarity','Specificity','Explainability']]
df_4 = df_4[['id','Clarity','Specificity','Explainability']]
```

## Analysis

#### Is empathy score higher for lengthier texts?

Annotator Shubham

```
#Annotater Shubham
df_1_full = pd.read_excel('/content/CounselChat-EDA (1).xlsx',sheet_name='Training Sheet_Shubham')
import numpy as np
df_1_full["length"]=df_1_full["answerText"].str.len()

df_1_full=df_1_full[['id','Warmth','Concern','Acknowledgement',"length"]]

df_1_full['Mean Empathy'] = df_1_full[['Warmth', 'Concern', 'Acknowledgement']].mean(axis=1)

correlation = df_1_full['length'].corr(df_1_full['Mean Empathy'])

print("Correlation between Length of reply and Overall Empathy for annotator Shubham:", correlation)

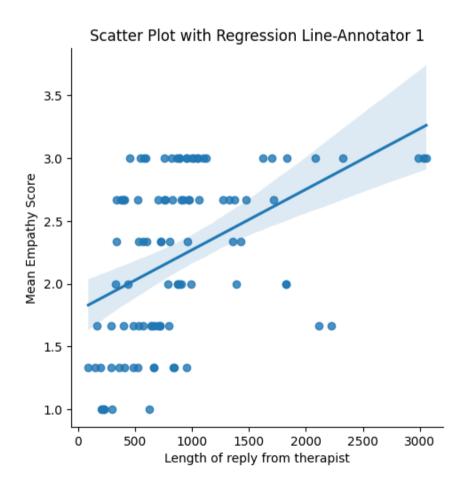
Correlation between Length of reply and Overall Empathy for annotator Shubham: 0.443948408532576

import scipy.stats as stats
corr, p_value = stats.pearsonr(df_1_full['length'], df_1_full['Mean Empathy'])
p_value
```

#### 3.7184019000076554e-06

```
import matplotlib.pyplot as plt
import seaborn as sns

# Create a scatter plot with a regression line
sns.lmplot(x='length', y='Mean Empathy', data=df_1_full)
plt.title('Scatter Plot with Regression Line-Annotator 1')
plt.xlabel('Length of reply from therapist')
plt.ylabel('Mean Empathy Score')
plt.show()
```



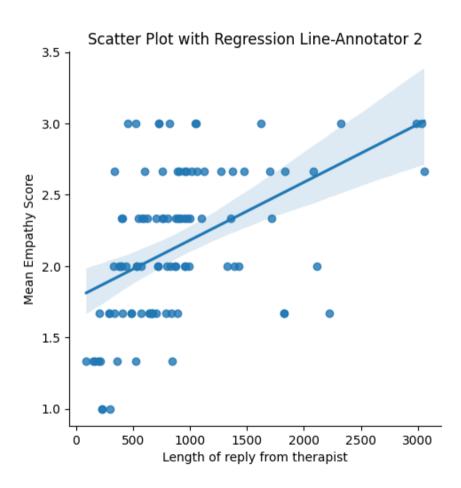
### Is empathy score higher for lengthier texts?

Annotator Mustafa

```
#Annotater Mustafa
df 2 full = pd.read excel('/content/CounselChat-EDA (1).xlsx',sheet name='Training Sheet Mustafa')
import numpy as np
df 2 full["length"]=df 2 full["answerText"].str.len()
df_2_full=df_2_full[['id','Warmth','Concern','Acknowledgement',"length"]]
df 2 full['Mean Empathy'] = df 2 full[['Warmth', 'Concern', 'Acknowledgement']].mean(axis=1)
correlation = df_2_full['length'].corr(df_2_full['Mean Empathy'])
print("Correlation between Length of reply and Overall Empathy for annotator Mustafa:", correlation)
    Correlation between Length of reply and Overall Empathy for annotator Mustafa: 0.4702036310903028
corr, p_value = stats.pearsonr(df_2_full['length'], df_2_full['Mean Empathy'])
p_value
    7.972729680658673e-07
```

```
import matplotlib.pyplot as plt
import seaborn as sns
```

```
# Create a scatter plot with a regression line
sns.lmplot(x='length', y='Mean Empathy', data=df_2_full)
plt.title('Scatter Plot with Regression Line-Annotator 2')
plt.xlabel('Length of reply from therapist')
plt.ylabel('Mean Empathy Score')
plt.show()
```



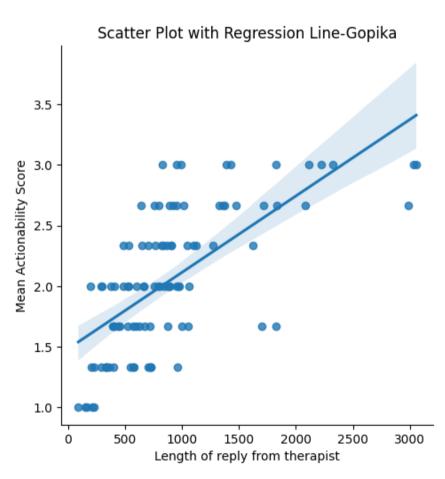
### Is actionability score higher for lengthier texts?

**Annotator Gopika** 

```
#Annotater Gopika
df_3_full= pd.read_excel('/content/CounselChat-EDA (1).xlsx',sheet_name='Gopika - Training Sheet Main')
df_3_full.rename(columns={'Specificty':'Specificity'},inplace=True)
import numpy as np
df 3 full["length"]=df_3_full["answerText"].str.len()
df_3_full=df_3_full[['id','Clarity','Specificity','Explainability',"length"]]
df_3_full['Mean Actionability'] = df_3_full[['Clarity', 'Specificity', 'Explainability']].mean(axis=1)
correlation = df_3_full['length'].corr(df_3_full['Mean Actionability'])
print("Correlation between Length of reply and Overall Actionability for annotator Gopika:", correlation)
    Correlation between Length of reply and Overall Actionability for annotator Gopika: 0.6684060540235198
corr, p_value = stats.pearsonr(df_3_full['length'], df_3_full['Mean Actionability'])
p_value
    3.0023672147530743e-14
```

import matplotlib.pyplot as plt
import seaborn as sns

# Create a scatter plot with a regression line
sns.lmplot(x='length', y='Mean Actionability', data=df\_3\_full)
plt.title('Scatter Plot with Regression Line-Gopika')
plt.xlabel('Length of reply from therapist')
plt.ylabel('Mean Actionability Score')
plt.show()



Annotator Ali

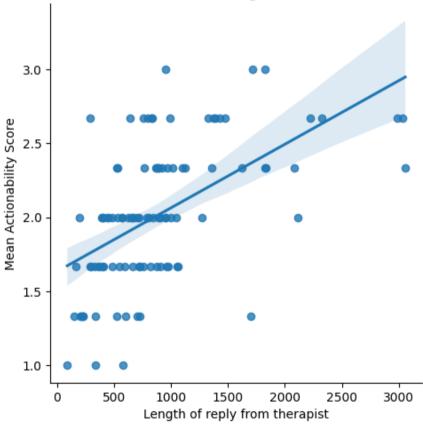
## Is actionability score higher for lengthier texts?

#Annotater Ali df 4 full= pd.read excel('/content/CounselChat-EDA (1).xlsx',sheet name='Ali - Training Sheet Main') import numpy as np df 4 full["length"]=df 4 full["answerText"].str.len() df\_4\_full=df\_4\_full[['id','Clarity','Specificity','Explainability',"length","topic"]] df\_4\_full['Mean Actionability'] = df\_4\_full[['Clarity', 'Specificity', 'Explainability']].mean(axis=1) <ipython-input-42-850229ef8b6c>:4: SettingWithCopyWarning: A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row indexer,col indexer] = value instead See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user guide/indexing.html#returning-a-view-versusdf 4 full['Mean Actionability'] = df 4 full[['Clarity', 'Specificity', 'Explainability']].mean(axis=1) import scipy.stats as stats correlation = df\_4\_full['length'].corr(df\_4\_full['Mean Actionability']) print("Correlation between Length of reply and Overall Actionability for annotator Ali:", correlation) Correlation between Length of reply and Overall Actionability for annotator Ali: 0.5521024980834964 corr, p\_value = stats.pearsonr(df\_4\_full['length'], df\_4\_full['Mean Actionability']) p value 2.6100405102308717e-09

```
import matplotlib.pyplot as plt
import seaborn as sns
```

```
# Create a scatter plot with a regression line
sns.lmplot(x='length', y='Mean Actionability', data=df_4_full)
plt.title('Scatter Plot with Regression Line-Ali')
plt.xlabel('Length of reply from therapist')
plt.ylabel('Mean Actionability Score')
plt.show()
```





# Cohen's Kappa Analysis for Empathy

```
from sklearn.metrics import cohen kappa score
kappa warmth = cohen kappa score(df 1 full["Warmth"], df 2 full["Warmth"], weights='quadratic')
print("Quadratic Weighted Kappa:", kappa warmth)
    Quadratic Weighted Kappa: 0.6532156368221942
from sklearn.metrics import cohen kappa score
kappa_acknowledgement = cohen_kappa_score(df_1_full["Acknowledgement"], df_2_full["Acknowledgement"], weights='quadratic')
print("Quadratic Weighted Kappa:", kappa acknowledgement)
    Quadratic Weighted Kappa: 0.5792833309154214
from sklearn.metrics import cohen_kappa_score
kappa_concern = cohen_kappa_score(df_1_full["Concern"], df_2_full["Concern"], weights='quadratic')
print("Quadratic Weighted Kappa:", kappa concern)
    Quadratic Weighted Kappa: 0.49141767323585506
import statistics
data empathy=[kappa warmth,kappa acknowledgement,kappa concern]
# Calculate the mean
mean value = statistics.mean(data empathy)
# Calculate the standard deviation
std_value = statistics.stdev(data_empathy)
print("Mean:", mean value)
print("Standard Deviation:", std value)
    Mean: 0.5746388803244902
    Standard Deviation: 0.08099891015402788
```

## Cohen's Kappa Analysis for Actionability

```
from sklearn.metrics import cohen kappa score
kappa_specificity = cohen_kappa_score(df_3_full["Specificity"], df_4_full["Specificity"])
print("Quadratic Weighted Kappa:", kappa specificity)
    Quadratic Weighted Kappa: 0.37339331619537275
from sklearn.metrics import cohen kappa score
kappa clarity = cohen kappa score(df 3 full["Clarity"], df 4 full["Clarity"], weights='quadratic')
print("Quadratic Weighted Kappa:", kappa clarity)
    Quadratic Weighted Kappa: 0.751727115716753
from sklearn.metrics import cohen kappa score
kappa explainability = cohen kappa score(df 3 full["Explainability"], df 4 full["Explainability"], weights='quadratic')
print("Quadratic Weighted Kappa:", kappa_explainability)
    Quadratic Weighted Kappa: 0.6366939146230699
import statistics
data_actionability=[kappa_specificity,kappa_clarity,kappa_explainability]
# Calculate the mean
mean value = statistics.mean(data actionability)
# Calculate the standard deviation
std_value = statistics.stdev(data_actionability)
print("Mean:", mean value)
print("Standard Deviation:", std value)
```

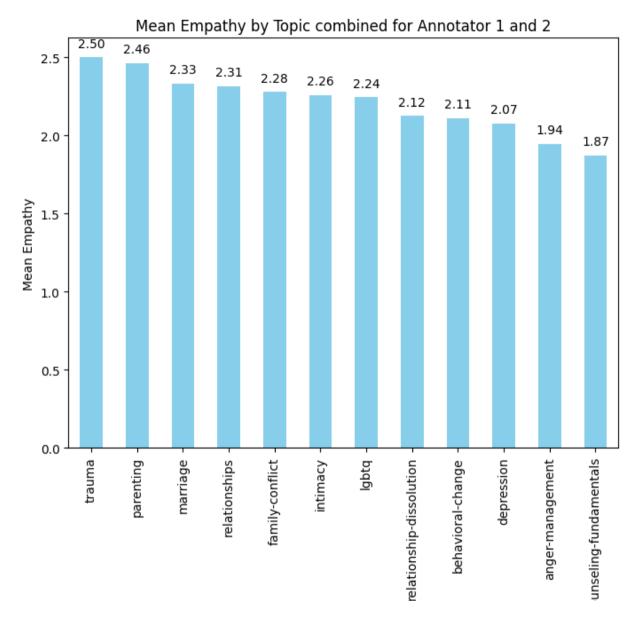
Mean: 0.5872714488450652

Standard Deviation: 0.19394857838549506

## Mean Empathy across topics

```
df 1 full=pd.read excel('/content/CounselChat-EDA (1).xlsx',sheet_name='Training Sheet_Shubham')
df 2 full = pd.read excel('/content/CounselChat-EDA (1).xlsx',sheet name='Training Sheet Mustafa')
df combined 1= pd.merge(df 1 full, df 2 full, on='id', how='left')
df combined 1.shape
     (100, 42)
df combined 1.columns
    Index(['id', 'og_id_x', 'questionID_x', 'questionTitle_x', 'questionText_x',
            'questionLink_x', 'topic_x', 'therapistInfo_x', 'therapistURL_x',
            'answerText_x', 'Warmth_x', 'Concern_x', 'Acknowledgement x',
            'Specificity_x', 'Clarity_x', 'Practicality (feasible)', 'Unnamed: 16',
            'Unnamed: 17', 'Unnamed: 18', 'Concern.1 x',
            'The therapist conveys concern by seeming to show regard for, and interest in, the patient. The therapist uses vocabulary
    and syntax which give the impression that they are involved with the patient and attentive to what the patient has said. x',
            'og_id_y', 'questionID_y', 'questionTitle_y', 'questionText_y',
            'questionLink_y', 'topic_y', 'therapistInfo_y', 'therapistURL_y',
            'answerText_y', 'upvotes', 'views', 'split', 'Warmth_y', 'Concern_y',
            'Acknowledgement_y', 'Specificity_y', 'Clarity_y', 'Practicality',
            'Unnamed: 19', 'Concern.1 v',
            'The therapist conveys concern by seeming to show regard for, and interest in, the patient. The therapist uses vocabulary
    and syntax which give the impression that they are involved with the patient and attentive to what the patient has said. y'],
           dtype='object')
df combined 1=df combined 1[['id','topic x', 'answerText x','Warmth x', 'Concern x', 'Acknowledgement x','Warmth y', 'Concern y',
       'Acknowledgement v']]
```

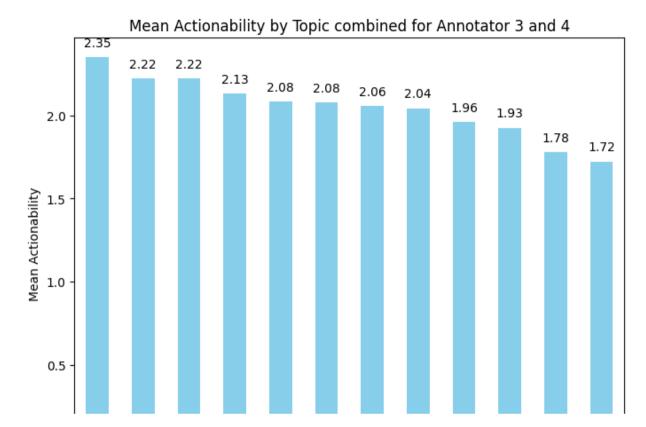
```
df combined 1["Mean Empathy 1 2"]=df combined 1[['Warmth x', 'Concern x', 'Acknowledgement x', 'Warmth v', 'Concern v',
       'Acknowledgement v'll.mean(axis=1)
df combined 1["Mean Empathy_1"]=df_combined_1[['Warmth_x', 'Concern_x', 'Acknowledgement_x']].mean(axis=1)
df combined 1["Mean Empathy 2"]=df combined 1[['Warmth y', 'Concern y', 'Acknowledgement y']].mean(axis=1)
mean empathy by topic = df combined 1.groupby('topic x')['Mean Empathy 1 2'].mean()
mean_empathy_by_topic
    topic x
    anger-management
                                 1.944444
    behavioral-change
                                 2.111111
    counseling-fundamentals
                                 1.871795
    depression
                                 2.074074
                                 2,277778
    family-conflict
    intimacy
                                 2.257576
     labta
                                 2.244444
                                 2.333333
    marriage
                                 2,462963
    parenting
    relationship-dissolution
                                 2.125000
     relationships
                                 2.314815
    trauma
                                 2.500000
    Name: Mean Empathy 1 2, dtype: float64
mean empathy by topic = mean empathy by topic.sort values(ascending=False)
plt.figure(figsize=(8, 6))
ax=mean empathy by topic.plot(kind='bar', color='skyblue')
for i, v in enumerate(mean empathy by topic):
    ax.text(i, v + 0.05, f'{v:.2f}', ha='center', va='bottom')
# Add labels and a title
plt.xlabel('Topic')
plt.ylabel('Mean Empathy')
plt.title('Mean Empathy by Topic combined for Annotator 1 and 2')
# Display the chart
plt.show()
```



Mean Actionability across topics

```
df 3 full= pd.read excel('/content/CounselChat-EDA (1).xlsx',sheet name='Gopika - Training Sheet Main')
df 4 full= pd.read excel('/content/CounselChat-EDA (1).xlsx',sheet name='Ali - Training Sheet Main')
df combined 2= pd.merge(df 3 full, df 4 full, on='id', how='left')
df combined 2.columns
    Index(['id', 'og_id_x', 'questionID_x', 'questionTitle_x', 'questionText_x',
            'questionLink_x', 'topic_x', 'therapistInfo_x', 'therapistURL_x',
            'answerText x', 'upvotes x', 'views x', 'split x', 'Warmth x',
            'Concern x', 'Acknowledgement x', 'Clarity x', 'Specificity x',
            'Explainability_x', 'Actionability_Avg_x', 'Unnamed: 20_x',
            'Unnamed: 21 x', 'Unnamed: 22 x', 'og id y', 'questionID y',
            'questionTitle_y', 'questionText_y', 'questionLink_y', 'topic_y',
            'therapistInfo_y', 'therapistURL_y', 'answerText_y', 'upvotes_y',
            'views_y', 'split_y', 'Warmth_y', 'Concern_y', 'Acknowledgement_y',
            'Clarity_y', 'Specificity_y', 'Explainability_y', 'Actionability_Avg_y',
            'Unnamed: 20 y', 'Unnamed: 21 y', 'Unnamed: 22 y'],
           dtype='object')
df combined 2=df combined 2[['id','topic x', 'answerText x','Clarity x', 'Specificity x',
       'Explainability x','Clarity y', 'Specificity y', 'Explainability y']]
df combined 2["Mean Actionability 3 4"]=df combined 2[['Clarity x', 'Specificity x',
       'Explainability_x','Clarity_y', 'Specificity_y', 'Explainability_y']].mean(axis=1)
df combined 2["Mean Actionability 3"]=df combined 2[['Clarity x', 'Specificity x',
       'Explainability x']].mean(axis=1)
df combined 2["Mean Actionability 4"]=df combined 2[['Clarity y', 'Specificity y', 'Explainability y']].mean(axis=1)
    <ipython-input-79-550ba755b28c>:1: SettingWithCopyWarning:
    A value is trying to be set on a copy of a slice from a DataFrame.
    Try using .loc[row indexer,col indexer] = value instead
    See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user quide/indexing.html#returning-a-view-versus-
      df combined 2["Mean Actionability 3 4"]=df combined 2[['Clarity x', 'Specificity x',
    <ipython-input-79-550ba755b28c>:3: SettingWithCopyWarning:
    A value is trying to be set on a copy of a slice from a DataFrame.
    Try using .loc[row indexer,col indexer] = value instead
```

```
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user guide/indexing.html#returning-a-view-versus-
      df combined 2["Mean Actionability 3"]=df combined 2[['Clarity x'. 'Specificity x'.
    <ipvthon-input-79-550ba755b28c>:5: SettingWithCopvWarning:
    A value is trying to be set on a copy of a slice from a DataFrame.
    Try using .loc[row indexer,col indexer] = value instead
    See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user guide/indexing.html#returning-a-view-versus-
      df combined 2["Mean Actionability 4"]=df combined 2[['Clarity y', 'Specificity y', 'Explainability y']].mean(axis=1)
mean actionability by topic = df combined 2.groupby('topic x')['Mean Actionability 3 4'].mean()
mean actionability by topic = mean actionability by topic.sort values(ascending=False)
plt.figure(figsize=(8, 6))
ax=mean actionability by topic.plot(kind='bar', color='skyblue')
for i, v in enumerate(mean actionability by topic):
    ax.text(i, v + 0.05, f'\{v:.2f\}', ha='center', va='bottom')
# Add labels and a title
plt.xlabel('Topic')
plt.ylabel('Mean Actionability')
plt.title('Mean Actionability by Topic combined for Annotator 3 and 4')
# Display the chart
plt.show()
```



## Regression Analysis- Empathy

```
Requirement already satisfied: pandas in /usr/local/lib/python3.10/dist-packages (1.5.3)
Requirement already satisfied: scikit-learn in /usr/local/lib/python3.10/dist-packages (1.2.2)
Requirement already satisfied: python-dateutil>=2.8.1 in /usr/local/lib/python3.10/dist-packages (from pandas) (2.8.2)
Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.10/dist-packages (from pandas) (2023.3.post1)
Requirement already satisfied: numpy>=1.21.0 in /usr/local/lib/python3.10/dist-packages (from scikit-learn) (1.23.5)
Requirement already satisfied: joblib>=1.1.1 in /usr/local/lib/python3.10/dist-packages (from scikit-learn) (1.3.2)
Requirement already satisfied: threadpoolctl>=2.0.0 in /usr/local/lib/python3.10/dist-packages (from scikit-learn) (3.2.0)
Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.10/dist-packages (from python-dateutil>=2.8.1->pandas) (1.16.0)
```

```
df_empathy=df_combined_1[['id','answerText_x','Mean Empathy_1_2']]
df_empathy["length"]=df_empathy['answerText_x'].str.len()
df empathy
```

<ipython-input-83-73109c4fb138>:2: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row\_indexer,col\_indexer] = value instead

See the caveats in the documentation: <a href="https://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returnin">https://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returnin</a> df\_empathy["length"]=df\_empathy['answerText\_x'].str.len()

	id	answerText_x	Mean Empathy_1_2	length	$\blacksquare$
0	1	If everyone thinks you're worthless, then mayb	2.166667	961	th
1	2	Hello, and thank you for your question and see	2.833333	2082	
2	3	First thing I'd suggest is getting the sleep y	1.000000	299	
3	4	Therapy is essential for those that are feelin	1.333333	844	
4	5	I first want to let you know that you are not	2.500000	337	
95	96	There is some great advice here that can reall	2.333333	375	
96	97	It's a good question man, and it must be terri	2.666667	576	
97	98	Make sure that you continue to treat your $\operatorname{Mom}\nolimits \ldots$	1.333333	208	
98	99	You have the answer already. It is not your mo	1.833333	401	
99	100	It is very difficult to move from being the ch	2.500000	905	

100 rows x 4 columns

```
import pandas as pd
from sklearn.model selection import train test split
from sklearn.feature extraction.text import TfidfVectorizer
from sklearn.linear model import LinearRegression
from sklearn.compose import ColumnTransformer
from sklearn.pipeline import Pipeline
from sklearn.metrics import mean squared error
from sklearn.preprocessing import StandardScaler
from sklearn.feature extraction.text import TfidfVectorizer
# Split the data into training and testing sets
X train, X test, y train, y test = train test split(df empathy[['answerText x','length']], df empathy['Mean Empathy 1 2'], test size=0.2
# Create a pipeline with TF-IDF for text and StandardScaler for text length, followed by Linear Regression
model = Pipeline([
    ('features', ColumnTransformer([
       ('answerText x', TfidfVectorizer(), 'answerText x'),
                                                                    # Convert text to TF-IDF features
        ('length', StandardScaler(), ['length']) # Standardize text length
    1)).
    ('regressor', LinearRegression()) # Linear Regression model
1)
# Train the model
model.fit(X train, y train)
# Make predictions on the test set
predictions = model.predict(X test)
# Evaluate the model
mse = mean squared error(y test, predictions)
print(f'Mean Squared Error: {mse}')
    Mean Squared Error: 0.21670950539886774
```

```
from sklearn.metrics import r2_score
# Calculate R-squared
r2 = r2_score(y_test, predictions)
print(f'R-squared: {r2}')
    R-squared: 0.3184938026329558
```

# Regression Analysis- Actionability

```
Using TFIDF

df_actionability=df_combined_2[['id', 'answerText_x', 'Mean Actionability_3_4']]

df_actionability["length"]=df_actionability['answerText_x'].str.len()

<ipython-input-89-18c505b0de23>:2: SettingWithCopyWarning:
    A value is trying to be set on a copy of a slice from a DataFrame.
    Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: <a href="https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-df_actionability["length"]=df_actionability['answerText_x'].str.len()

X_train, X_test, y_train, y_test = train_test_split(df_actionability[['answerText_x', 'length']], df_actionability['Mean Actionability_3_</a>
```

```
# Create a pipeline with TF-IDF for text and StandardScaler for text length, followed by Linear Regression
model = Pipeline([
    ('features', ColumnTransformer([
        ('answerText x', TfidfVectorizer(), 'answerText x'),
                                                                    # Convert text to TF-IDF features
        ('length', StandardScaler(), ['length']) # Standardize text length
    1)).
    ('regressor', LinearRegression()) # Linear Regression model
1)
# Train the model
model.fit(X_train, y_train)
# Make predictions on the test set
predictions = model.predict(X test)
# Evaluate the model
mse = mean squared error(y test, predictions)
print(f'Mean Squared Error: {mse}')
    Mean Squared Error: 0.1725684592320577
from sklearn.metrics import r2 score
# Calculate R-squared
r2 = r2_score(y_test, predictions)
print(f'R-squared: {r2}')
    R-squared: 0.28158837440253504
```

# BERT based analysis- Empathy

```
df_empathy=df_combined_1[['id','answerText_x','Mean Empathy_1_2']]
df_empathy["length"]=df_empathy['answerText_x'].str.len()
df empathy
```

<ipython-input-99-73109c4fb138>:2: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row\_indexer,col\_indexer] = value instead

See the caveats in the documentation: <a href="https://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returnin">https://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returnin</a> df\_empathy["length"]=df\_empathy['answerText\_x'].str.len()

	id	answerText_x	Mean Empathy_1_2	length
0	1	If everyone thinks you're worthless, then mayb	2.166667	961
1	2	Hello, and thank you for your question and see	2.833333	2082
2	3	First thing I'd suggest is getting the sleep y	1.000000	299
3	4	Therapy is essential for those that are feelin	1.333333	844
4	5	I first want to let you know that you are not	2.500000	337
95	96	There is some great advice here that can reall	2.333333	375
96	97	It's a good question man, and it must be terri	2.666667	576
97	98	Make sure that you continue to treat your Mom	1.333333	208
98	99	You have the answer already. It is not your mo	1.833333	401
		the second of th	2 52222	225

<sup>!</sup>pip install torch

```
Requirement already satisfied: torch in /usr/local/lib/python3.10/dist-packages (2.1.0+cu118)
Requirement already satisfied: filelock in /usr/local/lib/python3.10/dist-packages (from torch) (3.13.1)
Requirement already satisfied: typing-extensions in /usr/local/lib/python3.10/dist-packages (from torch) (4.5.0)
Requirement already satisfied: sympy in /usr/local/lib/python3.10/dist-packages (from torch) (1.12)
Requirement already satisfied: networkx in /usr/local/lib/python3.10/dist-packages (from torch) (3.2.1)
Requirement already satisfied: fsspec in /usr/local/lib/python3.10/dist-packages (from torch) (3.1.2)
Requirement already satisfied: fsspec in /usr/local/lib/python3.10/dist-packages (from torch) (2023.6.0)
Requirement already satisfied: MarkupSafe>=2.0 in /usr/local/lib/python3.10/dist-packages (from jinja2->torch) (2.1.3)
Requirement already satisfied: mpmath>=0.19 in /usr/local/lib/python3.10/dist-packages (from sympy->torch) (1.3.0)
Requirement already satisfied: transformers in /usr/local/lib/python3.10/dist-packages (from transformers) (3.13.1)
Requirement already satisfied: huggingface-hub<1.0,>=0.16.4 in /usr/local/lib/python3.10/dist-packages (from transformers) (0.19.4)
```

<sup>!</sup>pip install transformers

<sup>!</sup>pip install scikit-learn

```
Requirement already satisfied: numpy>=1.17 in /usr/local/lib/python3.10/dist-packages (from transformers) (1.23.5)
Requirement already satisfied: packaging>=20.0 in /usr/local/lib/python3.10/dist-packages (from transformers) (23.2)
Requirement already satisfied: pyvaml>=5.1 in /usr/local/lib/python3.10/dist-packages (from transformers) (6.0.1)
Requirement already satisfied: regex!=2019.12.17 in /usr/local/lib/python3.10/dist-packages (from transformers) (2023.6.3)
Requirement already satisfied: requests in /usr/local/lib/python3.10/dist-packages (from transformers) (2.31.0)
Requirement already satisfied: tokenizers<0.19.>=0.14 in /usr/local/lib/python3.10/dist-packages (from transformers) (0.15.0)
Requirement already satisfied: safetensors>=0.3.1 in /usr/local/lib/python3.10/dist-packages (from transformers) (0.4.1)
Requirement already satisfied: tgdm>=4.27 in /usr/local/lib/python3.10/dist-packages (from transformers) (4.66.1)
Requirement already satisfied: fsspec>=2023.5.0 in /usr/local/lib/python3.10/dist-packages (from huggingface-hub<1.0.>=0.16.4->trans
Requirement already satisfied: typing-extensions>=3.7.4.3 in /usr/local/lib/python3.10/dist-packages (from huggingface-hub<1.0,>=0.1
Requirement already satisfied: charset-normalizer<4.>=2 in /usr/local/lib/python3.10/dist-packages (from requests->transformers) (3.
Requirement already satisfied: idna<4.>=2.5 in /usr/local/lib/python3.10/dist-packages (from requests->transformers) (3.6)
Requirement already satisfied: urllib3<3,>=1.21.1 in /usr/local/lib/python3.10/dist-packages (from requests->transformers) (2.0.7)
Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.10/dist-packages (from requests->transformers) (2023.11.
Requirement already satisfied: scikit-learn in /usr/local/lib/python3.10/dist-packages (1.2.2)
Requirement already satisfied: numpy>=1.17.3 in /usr/local/lib/python3.10/dist-packages (from scikit-learn) (1.23.5)
Requirement already satisfied: scipy>=1.3.2 in /usr/local/lib/python3.10/dist-packages (from scikit-learn) (1.11.4)
Requirement already satisfied: ioblib>=1.1.1 in /usr/local/lib/python3.10/dist-packages (from scikit-learn) (1.3.2)
Requirement already satisfied: threadpoolctl>=2.0.0 in /usr/local/lib/python3.10/dist-packages (from scikit-learn) (3.2.0)
```

```
import torch
from transformers import BertTokenizer, BertModel
from sklearn.model selection import train test split
from sklearn.linear model import LinearRegression
from sklearn.metrics import mean squared error
import pandas as pd
# Split the data
train data, test data = train test split(df empathy, test size=0.2, random state=42)
# Tokenization using BERT
tokenizer = BertTokenizer.from pretrained('bert-base-uncased')
train tokens = tokenizer(train data['answerText x'].tolist(), padding=True, truncation=True, max length=128, return tensors='pt')
test_tokens = tokenizer(test_data['answerText_x'].tolist(), padding=True, truncation=True, max_length=128, return_tensors='pt')
# Use BERT embeddings for regression
bert model = BertModel.from pretrained('bert-base-uncased')
with torch.no_grad():
    train embeddings = bert model(**train tokens).last hidden state.mean(dim=1)
    test embeddings = bert model(**test tokens).last hidden state.mean(dim=1)
# Include the length of the reply as another parameter
train lengths = torch.tensor(train data['length'].tolist(), dtype=torch.float32).unsqueeze(1)
test lengths = torch.tensor(test data['length'].tolist(), dtype=torch.float32).unsqueeze(1)
```

## BERT Analysis- Actionability

```
import torch
from transformers import BertTokenizer, BertModel
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean_squared_error
import pandas as pd
```