#EMOTIONS DECODING

```
import torch
from transformers import AutoTokenizer, AutoModelForSequenceClassification
import torch.nn.functional as F
# Load pre-trained model and tokenizer for emotion analysis
model_name = "nateraw/bert-base-uncased-emotion"
tokenizer = AutoTokenizer.from pretrained(model name)
model = AutoModelForSequenceClassification.from pretrained(model name)
# List of emotions
emotions = ['admiration', 'amusement', 'anger', 'annoyance', 'approval',
      'caring', 'confusion', 'curiosity', 'desire', 'disappointment',
      'disapproval', 'disgust', 'embarrassment', 'excitement',
      'fear', 'gratitude', 'grief', 'joy', 'love', 'nervousness',
      'optimism', 'pride', 'realization', 'relief', 'remorse',
      'sadness', 'surprise', 'neutral']
def analyze emotion(text):
  # Tokenize input text
  inputs = tokenizer(text, return_tensors="pt", truncation=True, padding=True)
  # Get model outputs
  with torch.no_grad():
    outputs = model(**inputs)
  # Convert logits to probabilities
  probs = F.softmax(outputs.logits, dim=1)
```

```
# Get top emotion
top_prob, top_class = torch.max(probs, dim=1)
emotion = emotions[top_class.item()]

return emotion, top_prob.item()

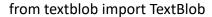
# Example usage
if _name_ == "_main_":
    print("Enter social media conversation (type 'exit' to quit):")
    while True:
    text = input(">> ")
    if text.lower() == "exit":
        break
    emotion, confidence = analyze_emotion(text)
    print(f"Detected Emotion: {emotion} (Confidence: {confidence:.2f})")
```

#MEAN EMOTIONS DEFECTION FROM MULTIPLE TEXTS

```
import torch
from transformers import AutoTokenizer, AutoModelForSequenceClassification
import torch.nn.functional as F
from collections import Counter
# Load model and tokenizer
model_name = "nateraw/bert-base-uncased-emotion"
tokenizer = AutoTokenizer.from pretrained(model name)
model = AutoModelForSequenceClassification.from pretrained(model name)
# Emotion labels
emotions = ['admiration', 'amusement', 'anger', 'annoyance', 'approval',
      'caring', 'confusion', 'curiosity', 'desire', 'disappointment',
      'disapproval', 'disgust', 'embarrassment', 'excitement',
      'fear', 'gratitude', 'grief', 'joy', 'love', 'nervousness',
      'optimism', 'pride', 'realization', 'relief', 'remorse',
      'sadness', 'surprise', 'neutral']
def analyze_emotion(text):
  inputs = tokenizer(text, return_tensors="pt", truncation=True, padding=True)
  with torch.no_grad():
    outputs = model(**inputs)
  probs = F.softmax(outputs.logits, dim=1)
  top_prob, top_class = torch.max(probs, dim=1)
  return emotions[top class.item()]
def mean_emotion(text_list):
```

```
emotion_list = [analyze_emotion(text) for text in text_list]
  emotion count = Counter(emotion list)
  most common = emotion count.most common(1)[0]
  return most_common[0], emotion_count
# Example usage
if _name_ == "_main_":
  texts = [
    "I love how this turned out!",
    "This makes me so angry!",
    "I'm feeling quite happy and grateful.",
    "Why did this happen to me?",
    "This is so annoying and unfair."
  ]
  mean_emotion_label, all_counts = mean_emotion(texts)
  print(f"Mean (most common) emotion: {mean_emotion_label}")
  print("All detected emotions:", dict(all_counts))
```

#BASIC SENTIMENT ANALYSIS USING TEXTLOB



text = "I am really happy and excited about my future!"

blob = TextBlob(text)

sentiment = blob.sentiment

print("Text:", text)

print("Polarity:", sentiment.polarity)

print("Subjectivity:", sentiment.subjectivity)

#Emotion Detection Using Pretrained Transformer (Hugging Face)

```
import torch
from transformers import AutoTokenizer, AutoModelForSequenceClassification
import torch.nn.functional as F

tokenizer = AutoTokenizer.from_pretrained("nateraw/bert-base-uncased-emotion")

model = AutoModelForSequenceClassification.from_pretrained("nateraw/bert-base-uncased-emotion")

text = "I am feeling so grateful for your help."
inputs = tokenizer(text, return_tensors="pt")
outputs = model(**inputs)
probs = F.softmax(outputs.logits, dim=1)

labels = model.config.id2label
emotion = labels[probs.argmax().item()]
```

print("Detected Emotion:", emotion)

#Mean Emotion from Multiple Sentences

```
from collections import Counter

texts = ["I'm so proud of you!", "This is annoying.", "I'm happy!", "Why am I so sad?", "Thank
you!"]
emotion_list = []

for t in texts:
    inputs = tokenizer(t, return_tensors="pt")
    outputs = model(**inputs)
    probs = F.softmax(outputs.logits, dim=1)
    emotion = labels[probs.argmax().item()]
    emotion_list.append(emotion)

mean_emotion = Counter(emotion_list).most_common(1)[0][0]

print("Texts:", texts)
print("Detected Emotions:", emotion_list)
print("Most Common Emotion:", mean_emotion)
```

#Sentiment Classification Using Vader (Great for Social Media)

from vaderSentiment.vaderSentiment import SentimentIntensityAnalyzer
analyzer = SentimentIntensityAnalyzer()
text = "I can't believe how amazing this is!"
score = analyzer.polarity_scores(text)
print("Text:", text)
print("Scores:", score)
<pre>print("Sentiment:", "Positive" if score['compound'] > 0 else "Negative" if score['compound'] < 0 else "Neutral")</pre>

#Emotion Detection from a CSV of Comments

```
import pandas as pd

# Example comments

data = {'Comment': [
    "I am feeling sad.",
    "Wow, this made my day!",
    "You are so rude.",
    "Thanks a lot for helping me."

]}

df = pd.DataFrame(data)

# Detect emotion for each comment

df['Emotion'] = df['Comment'].apply(lambda x:
labels[torch.argmax(F.softmax(model(**tokenizer(x, return_tensors="pt")).logits, dim=1)).item()])
print(df)
```

#Real-Time Emotion Detection from User Input

```
while True:
    text = input("Enter your message (or 'exit'): ")
    if text.lower() == 'exit':
        break
    inputs = tokenizer(text, return_tensors="pt")
    outputs = model(**inputs)
    probs = F.softmax(outputs.logits, dim=1)
    emotion = labels[probs.argmax().item()]
    print(f"Detected Emotion: {emotion}")
```

#Detecting Top 3 Emotions with Probabilities

```
def top_emotions(text, top_k=3):
    inputs = tokenizer(text, return_tensors="pt")
    outputs = model(**inputs)
    probs = F.softmax(outputs.logits, dim=1)[0]
    top_probs, top_labels = torch.topk(probs, top_k)
    for i in range(top_k):
        print(f"{labels[top_labels[i].item()]}: {top_probs[i].item():.2f}")

top_emotions("I'm feeling a mix of excitement, fear, and pride.")
```

Emotion Detection in YouTube Comments (Dummy Data)

```
comments = [
   "This video is hilarious!",
   "I didn't like the ending.",
   "The song gave me chills.",
   "I cried watching this."
]

for comment in comments:
   inputs = tokenizer(comment, return_tensors="pt")
   outputs = model(**inputs)
   probs = F.softmax(outputs.logits, dim=1)
   emotion = labels[probs.argmax().item()]
   print(f"Comment: {comment} => Emotion: {emotion}")
```

#Emoji-Based Emotion Estimation

```
emoji_texts = [
  "I love this! \infty",
  "I'm furious right now! ",
  "Yay! ",
  "Why me... ",
]

for text in emoji_texts:
  inputs = tokenizer(text, return_tensors="pt")
  outputs = model(**inputs)
  emotion = labels[torch.argmax(F.softmax(outputs.logits, dim=1)).item()]
  print(f"Text: {text} => Emotion: {emotion}")
```

Compare Emotions Between Two Comments

```
def compare_emotions(text1, text2):
    def detect(text):
        inputs = tokenizer(text, return_tensors="pt")
        logits = model(**inputs).logits
        return labels[torch.argmax(F.softmax(logits, dim=1)).item()]

e1 = detect(text1)
    e2 = detect(text2)
    print(f"Text 1: {text1} => {e1}")
    print(f"Text 2: {text2} => {e2}")
    print("Same emotion" if e1 == e2 else "Different emotions")

compare_emotions("I'm scared about my exam", "I'm excited for the test!")
```