Using my saved model in an example to see if it works

```
In [25]: import joblib
         import numpy as np
         from scipy.sparse import hstack, csr matrix
         import ast
         import praw
         import time
         import pandas as pd
         from sklearn.feature extraction.text import TfidfVectorizer
         import re
         from nltk.corpus import stopwords
         from nltk.stem import WordNetLemmatizer
         from wordcloud import WordCloud
         import matplotlib.pyplot as plt
 In [ ]: # === 1. Load model and vectorizer ===
         model = joblib.load("toxicity_model.pkl")
         tfidf = joblib.load("Preprocessing.pkl")
In [54]: # === 2. Set up Reddit API ===
         reddit = praw.Reddit(
             client id="sdf",
             client_secret="sdf",
             user agent="toxicity-detector by /u/data-anna"
         # === 3. Define subreddits ===
         subreddits = [
             "PoliticalDiscussion",
             "Ask Politics",
             "ModeratePolitics",
             "NeutralPolitics",
             "Republican",
             "AskTrumpSupporters",
             "AskConservatives",
             "PoliticalHumor",
             "ConservativeMemes",
             "EnoughLibertarianSpam",
             "PoliticalCompassMemes",
             "BuyCanadian",
             "Canada",
             "Conservative",
             "Politics",
             "Worldnews",
             "Democrats",
             "CanadianPolitics",
             "Ontario",
```

```
"vancouver",
    "Alberta",
    "Toronto",
    "Ottawa"
# Collect top comments from top posts
comments = []
for sub in subreddits:
    print(f" Q Collecting top comments from r/{sub}...")
    count = 0
   try:
        subreddit = reddit.subreddit(sub)
        for post in subreddit.top(limit=100, time filter="all"): # You can use 'day', 'week', 'month', 'year', 'all'
            post.comments.replace_more(limit=0) # Skip "Load more comments"
            for comment in post.comments:
                if comment.body: # Ignore deleted/empty comments
                    comments.append({"subreddit": sub, "comment": comment.body})
                   count += 1
        print(f" ✓ Collected {count} top comments from r/{sub}")
        print("∑ Waiting 60 seconds to respect Reddit's rate limit...")
        time.sleep(60)
    except Exception as e:
        print(f" X Failed to fetch from r/{sub}: {e}")
print(f"\n >> Total top comments collected: {len(comments)}")
# === 5. Save to CSV ===
df = pd.DataFrame(comments)
df.to csv("reddit comments with subreddits.csv", index=False, encoding="utf-8")
print(" Saved to 'reddit_comments_with_subreddits.csv'")
import os
df = pd.DataFrame(comments)
file_path = "reddit_comments_with_subreddits.csv"
# Append if file exists, otherwise create new with header
if os.path.exists(file_path):
    df.to_csv(file_path, mode="a", header=False, index=False, encoding="utf-8")
else:
    df.to_csv(file_path, mode="w", header=True, index=False, encoding="utf-8")
print(f" Appended to '{file_path}'")
```

```
Q Collecting top comments from r/PoliticalDiscussion...
✓ Collected 7237 top comments from r/PoliticalDiscussion

    Waiting 60 seconds to respect Reddit's rate limit...

Q Collecting top comments from r/Ask Politics...
✓ Collected 2496 top comments from r/Ask Politics

▼ Waiting 60 seconds to respect Reddit's rate limit...

Q Collecting top comments from r/ModeratePolitics...
✓ Collected 4994 top comments from r/ModeratePolitics

    Waiting 60 seconds to respect Reddit's rate limit...

Q Collecting top comments from r/NeutralPolitics...
✓ Collected 2816 top comments from r/NeutralPolitics
Waiting 60 seconds to respect Reddit's rate limit...
Collecting top comments from r/Republican...
✓ Collected 6228 top comments from r/Republican

    Waiting 60 seconds to respect Reddit's rate limit...

Collecting top comments from r/AskTrumpSupporters...
✓ Collected 2060 top comments from r/AskTrumpSupporters

    Waiting 60 seconds to respect Reddit's rate limit...

Q Collecting top comments from r/AskConservatives...
✓ Collected 4143 top comments from r/AskConservatives

    Waiting 60 seconds to respect Reddit's rate limit...

    Collecting top comments from r/PoliticalHumor...

✓ Collected 10914 top comments from r/PoliticalHumor

    Waiting 60 seconds to respect Reddit's rate limit...

Q Collecting top comments from r/ConservativeMemes...

✓ Collected 747 top comments from r/ConservativeMemes

    Waiting 60 seconds to respect Reddit's rate limit...

Collecting top comments from r/EnoughLibertarianSpam...
✓ Collected 2243 top comments from r/EnoughLibertarianSpam
☑ Waiting 60 seconds to respect Reddit's rate limit...
Q Collecting top comments from r/PoliticalCompassMemes...
✓ Collected 10327 top comments from r/PoliticalCompassMemes

    Waiting 60 seconds to respect Reddit's rate limit...

Collecting top comments from r/BuyCanadian...
✓ Collected 18788 top comments from r/BuyCanadian

    Waiting 60 seconds to respect Reddit's rate limit...

Collecting top comments from r/Canada...
✓ Collected 14147 top comments from r/Canada

    Waiting 60 seconds to respect Reddit's rate limit...

Q Collecting top comments from r/Conservative...
✓ Collected 10801 top comments from r/Conservative

    Waiting 60 seconds to respect Reddit's rate limit...

Collecting top comments from r/Politics...
✓ Collected 11204 top comments from r/Politics

    Waiting 60 seconds to respect Reddit's rate limit...

Collecting top comments from r/Worldnews...
✓ Collected 6524 top comments from r/Worldnews

    Waiting 60 seconds to respect Reddit's rate limit...

Q Collecting top comments from r/Democrats...
✓ Collected 14807 top comments from r/Democrats

    Waiting 60 seconds to respect Reddit's rate limit...

Collecting top comments from r/CanadianPolitics...
```

```
✓ Collected 1157 top comments from r/CanadianPolitics

    Waiting 60 seconds to respect Reddit's rate limit...

Collecting top comments from r/Ontario...
✓ Collected 15351 top comments from r/Ontario

    Waiting 60 seconds to respect Reddit's rate limit...

Collecting top comments from r/vancouver...
✓ Collected 12190 top comments from r/vancouver

    Waiting 60 seconds to respect Reddit's rate limit...

Collecting top comments from r/Alberta...
✓ Collected 16287 top comments from r/Alberta

    Waiting 60 seconds to respect Reddit's rate limit...

Collecting top comments from r/Toronto...
✓ Collected 12764 top comments from r/Toronto

    Waiting 60 seconds to respect Reddit's rate limit...

Collecting top comments from r/Ottawa...
✓ Collected 11279 top comments from r/Ottawa

    Waiting 60 seconds to respect Reddit's rate limit...

> Total top comments collected: 199504
Appended to 'reddit_comments_with_subreddits.csv'
```

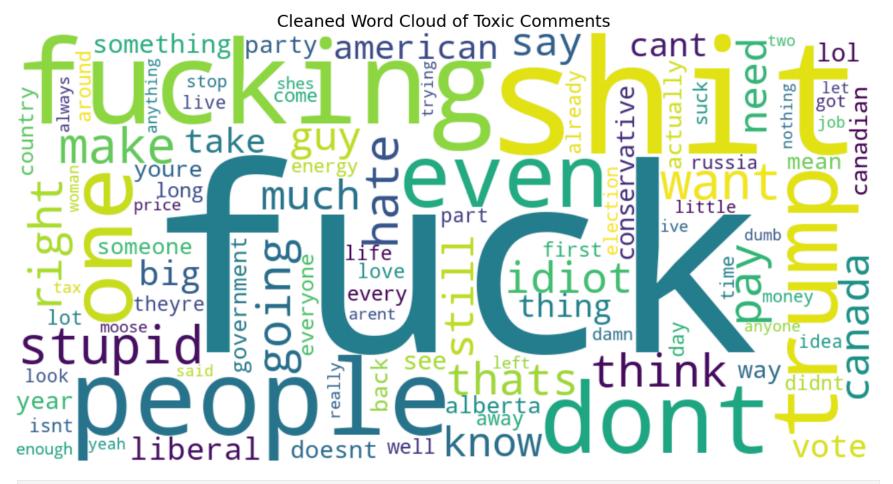
In [56]: df = pd.read_csv("reddit_comments_with_subreddits.csv") df

Out[56]:		subreddit	comment
	0	Canada	All skirt, no knickers. As they say.
	1	Canada	The key word is in your own comment "Either wa
	2	Canada	Gasoline is a minor part of oil companies reve
	3	Canada	>DEI establishes quotas which make sure that r
	4	Canada	We all stole the best continent in the world t
	•••		
	219846	Ottawa	Bhahahaha. God keep our land glorious and fr
	219847	Ottawa	It's classless. I wouldn't expect anything els
	219848	Ottawa	This is bad Orange man will use this as an
	219849	Ottawa	No, it's petty. Insane and rude even if you ha
	219850	Ottawa	I don't agree with it. Their anthem had existe

219851 rows × 2 columns

```
In [36]: comments = df['comment'].tolist()
```

```
In [38]: # === 3. Preprocess: TF-IDF only (no BERT) ===
         X tfidf = tfidf.transform(comments)
         # === 4. Predict ===
         predictions = model.predict(X_tfidf)
         probs = model.predict proba(X tfidf)[:, 1]
         # === 5. Show ALL comments with prediction ===
         toxic count = 0
         for i in range(len(comments)):
             label = " Toxic" if predictions[i] == 1 else " Non-toxic"
             if predictions[i] == 1:
                 toxic_count += 1
         print(f"\n Total toxic comments found: {toxic_count} out of {len(comments)}")
        ✓ Total toxic comments found: 642 out of 10081
In [40]: stop_words = set(stopwords.words('english'))
         lemmatizer = WordNetLemmatizer()
         def clean text(text):
             # Remove URLs and special characters
             text = re.sub(r"http\S+|www\S+|https\S+", '', text)
             text = re.sub(r"[^a-zA-Z\s]", '', text)
             text = text.lower()
             words = text.split()
             # Remove stopwords and short words, lemmatize the rest
             words = [lemmatizer.lemmatize(word) for word in words if word not in stop_words and len(word) > 2]
             return " ".join(words)
         # Apply to only toxic comments
         toxic_comments = [comments[i] for i in range(len(comments)) if predictions[i] == 1]
         cleaned_toxic_text = " ".join([clean_text(comment) for comment in toxic_comments])
In [42]: wordcloud = WordCloud(
             width=1000,
             height=500,
             background_color='white',
             max words=100,
             colormap='viridis', # try 'plasma', 'inferno', etc. for aesthetics
         ).generate(cleaned toxic text)
         plt.figure(figsize=(14, 7))
         plt.imshow(wordcloud, interpolation='bilinear')
         plt.axis("off")
         plt.title("Cleaned Word Cloud of Toxic Comments", fontsize=18)
         plt.tight layout()
         plt.show()
```



```
In []:

import pandas as pd
   import matplotlib.pyplot as plt
   from sklearn.preprocessing import MinMaxScaler

# === 1. Calculate toxicity rate ===
   df['toxicity_label'] = predictions
   df['toxicity_prob'] = probs

# Group and count
   toxic_counts = df[df['toxicity_label'] == 1].groupby('subreddit').size()
   total_counts = df.groupby('subreddit').size()

# Create summary table
   toxicity_summary = pd.DataFrame({
        'total_comments': total_counts,
```

```
'toxic_comments': toxic_counts
 }).fillna(0)
 toxicity_summary['toxicity_rate'] = toxicity_summary['toxic_comments'] / toxicity_summary['total_comments']
 # === 2. Normalize toxicity rate (0 to 1) ===
 scaler = MinMaxScaler()
 toxicity_summary['toxicity_rate_normalized'] = scaler.fit_transform(
     toxicity_summary[['toxicity_rate']]
 # === 3. Plot normalized toxicity rates ===
 toxicity_summary.sort_values('toxicity_rate_normalized', ascending=False).plot(
     kind='bar',
     y='toxicity rate normalized',
     legend=False,
     figsize=(10, 6),
     title=" Normalized Toxicity Rate by Subreddit (0-1)",
     color='darkred'
 plt.ylabel("Normalized Toxicity Score")
 plt.xlabel("Subreddit")
 plt.xticks(rotation=45, ha='right')
 plt.tight layout()
 plt.grid(axis='y')
 plt.show()
C:\Users\annaj\AppData\Local\Temp\ipykernel_10588\1729952249.py:40: UserWarning: Glyph 128202 (\N{BAR CHART}) missing from font(s) DejaVu Sans.
  plt.tight_layout()
C:\Users\annaj\anaconda3\Lib\site-packages\IPython\core\pylabtools.py:170: UserWarning: Glyph 128202 (\N{BAR CHART}) missing from font(s) DejaVu Sans.
fig.canvas.print figure(bytes io, **kw)
```

