

```
# IMPORTANT: SOME KAGGLE DATA SOURCES ARE PRIVATE
# RUN THIS CELL IN ORDER TO IMPORT YOUR KAGGLE DATA SOURCES.
import kagglehub
kagglehub.login()
```



Kaggle credentials successfully validated.

```
Kaggle credentials set.
Warning: Looks like you're using an outdated `kagglehub` version (installed: 0.3.11), please consider upgrading to the l
Kaggle credentials successfully validated.
```

```
# IMPORTANT: RUN THIS CELL IN ORDER TO IMPORT YOUR KAGGLE DATA SOURCES,
# THEN FEEL FREE TO DELETE THIS CELL.
# NOTE: THIS NOTEBOOK ENVIRONMENT DIFFERS FROM KAGGLE'S PYTHON
# ENVIRONMENT SO THERE MAY BE MISSING LIBRARIES USED BY YOUR
# NOTEBOOK.
```

```
wsdm_cup_multilingual_chatbot_arena_path = kagglehub.competition_download('wsdm-cup-multilingual-chatbot-arena')
```

```
print('Data source import complete.')
```



```
Downloading from https://www.kaggle.com/api/v1/competitions/data/download-all/wsdm-cup-multilingual-chatbot-arena...
100%|██████████| 108M/108M [00:01<00:00, 84.3MB/s] Extracting files...
```

```
Data source import complete.
```

```
# Data Overview and Basic Statistics
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import os
```

▼ Data Overview and Basic Statistics

```
# Load the dataset
df = pd.read_parquet(os.path.join(wsdm_cup_multilingual_chatbot_arena_path, 'train.parquet')) # Update with your actual file
```

```
# 1. Basic dataset information
print("Dataset Shape:", df.shape)
print("\nFirst 5 rows:")
print(df.head())
```



```
Dataset Shape: (48439, 8)
```

```
First 5 rows:
```

	id \	prompt \	response_a \	response_b	winner \
0	00007cff95d7f7974642a785aca248b0f26e60d3312fac...	0	00007cff95d7f7974642a785aca248b0f26e60d3312fac...	0	Áno, veď som tu! Môžem ti pomôcť s otázkami al...
1	00010ed04b536f56ebe43eef1100c13906abea12bf9855...	1	00010ed04b536f56ebe43eef1100c13906abea12bf9855...	1	Let's break down the news and analyze it accor...
2	0003800d510e38803efba5ceaec122bc66408fe367b0be...	2	0003800d510e38803efba5ceaec122bc66408fe367b0be...	2	Bu soruyu çözmek için, verilen koşulları adım ...
3	00072026c68f5418ef2da238394e418ce72a534b9b22d5...	3	00072026c68f5418ef2da238394e418ce72a534b9b22d5...	3	현재 추천된 탑 3 종목에 순위를 매기기 위해서는 여러 가지 요소들을 고려해야 합니...
4	0007ce7cf6bc1b5a8f8a4669b854fb12030863c970d9dc...	4	0007ce7cf6bc1b5a8f8a4669b854fb12030863c970d9dc...	4	Understood. Here is a straightforward, unadorn...

	model_a	model_b	language
0	o1-preview	reka-core-20240904	Slovak
1	gemma-2-27b-it	gemini-1.5-flash-002	Russian
2	gpt-4-0125-preview	claude-3-5-sonnet-20240620	Turkish
3	gemma-2-2b-it	llama-3.1-nemotron-70b-instruct	English

✓ Data types and missing values

```
# 2. Data types and missing values
print("\nData Types:")
print(df.dtypes)
print("\nMissing Values Summary:")
print(df.isnull().sum())
print(f"Total Missing Values: {df.isnull().sum().sum()}")
```



```
Data Types:
id          object
prompt      object
response_a  object
response_b  object
winner      object
model_a     object
model_b     object
language    object
dtype: object

Missing Values Summary:
id          0
prompt      0
response_a  0
response_b  0
winner      0
model_a     0
model_b     0
language    0
dtype: int64
Total Missing Values: 0
```

✓ Language distribution

```
# 3. Language distribution
print("\nLanguage Distribution:")
lang_counts = df['language'].value_counts()
print(lang_counts)
print("\nLanguage Distribution (Percentage):")
print(100 * lang_counts / len(df))
```



```
Language Distribution:
language
English      25211
Russian      6455
Chinese      4310
Vietnamese   3103
German       1402
...
Klingon       1
Kurdish       1
Hawaiian      1
Telugu        1
Sindhi        1
Name: count, Length: 128, dtype: int64

Language Distribution (Percentage):
language
English      52.046904
Russian      13.326039
Chinese       8.897789
Vietnamese    6.405995
German       2.894362
...
Klingon       0.002064
Kurdish       0.002064
Hawaiian      0.002064
Telugu        0.002064
Sindhi        0.002064
Name: count, Length: 128, dtype: float64
```

✓ winner distribution

```
# Get winner distribution
winner_counts = df['winner'].value_counts()
```

```
total_samples = len(df)

# Create pie chart
plt.figure(figsize=(10, 6))
plt.pie(winner_counts, labels=winner_counts.index, autopct='%1.1f%%',
        colors=['#ff9999', '#66b3ff'], startangle=90, explode=[0.05, 0.05])
plt.title('Winner Distribution (A vs B)', fontsize=16)
plt.axis('equal') # Equal aspect ratio ensures that pie is drawn as a circle
plt.legend([f"Model A: {winner_counts.get('A', 0)} ({winner_counts.get('A', 0)/total_samples*100:.1f}%)",
            f"Model B: {winner_counts.get('B', 0)} ({winner_counts.get('B', 0)/total_samples*100:.1f}%)",
            loc="best"])
plt.tight_layout()
plt.show()

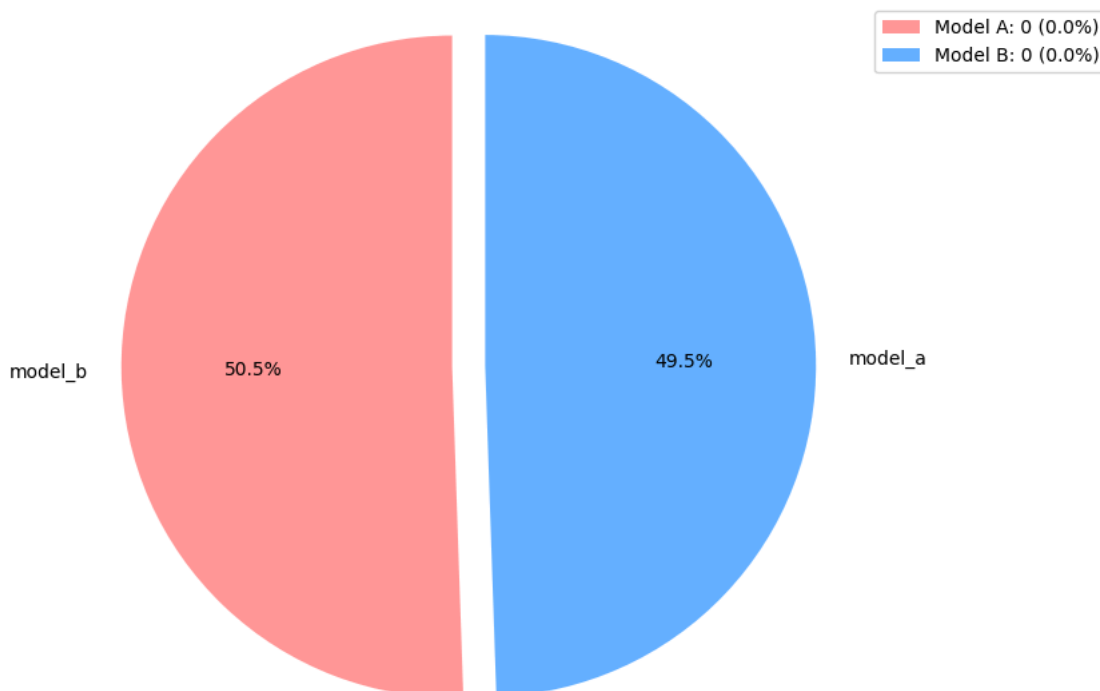
# Create bar chart
plt.figure(figsize=(8, 6))
sns.barplot(x=winner_counts.index, y=winner_counts.values, palette=['#ff9999', '#66b3ff'])
plt.title('Winner Distribution (A vs B)', fontsize=16)
plt.xlabel('Winner', fontsize=12)
plt.ylabel('Count', fontsize=12)

# Add count and percentage on top of bars
for i, v in enumerate(winner_counts.values):
    plt.text(i, v + 5, f"{v} ({v/total_samples*100:.1f}%)", ha='center')

plt.tight_layout()
plt.show()
```



Winner Distribution (A vs B)

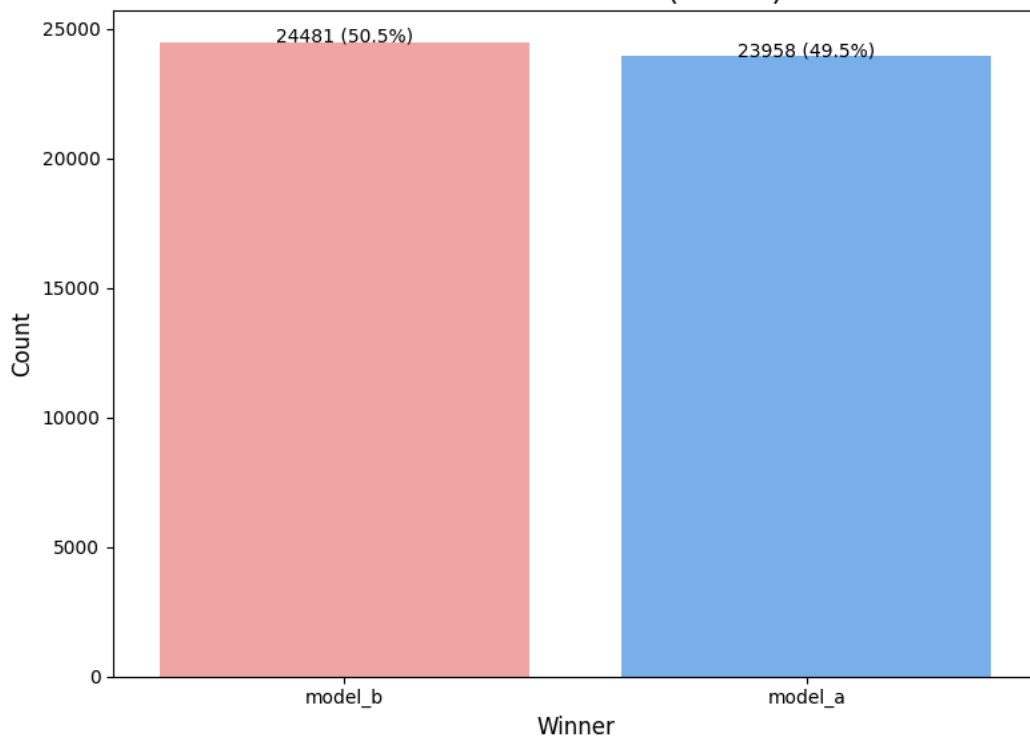


<ipython-input-10-f4dc21bfde41>:19: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue`

```
sns.barplot(x=winner_counts.index, y=winner_counts.values, palette=['#ff9999','#66b3ff'])
```

Winner Distribution (A vs B)



✓ Model distribution

4. Model distribution

```
model_a_counts = df['model_a'].value_counts()
```

```
model_b_counts = df['model_b'].value_counts()
```

```
model_counts_df = pd.DataFrame({  
    'Model A': model_a_counts,
```

```
'Model B': model_b_counts
}).fillna(0).astype(int) # Fill NaNs with 0 and convert to int
```

```
# Print the resulting DataFrame
print("\nModel A vs Model B Distribution:")
print(model_counts_df)
```

chatgpt-4o-latest-20240808	668	707
chatgpt-4o-latest-20240903	1863	1839
claude-3-5-sonnet-20240620	1255	1281
claude-3-5-sonnet-20241022	552	533
claude-3-haiku-20240307	866	909
claude-3-opus-20240229	1665	1748
command-r-08-2024	1037	976
command-r-plus-08-2024	1001	995
deepseek-coder-v2-0724	114	118
deepseek-v2-api-0628	59	72
deepseek-v2.5	830	806
gemini-1.5-flash-001	372	389
gemini-1.5-flash-002	1329	1357
gemini-1.5-flash-8b-001	1436	1427
gemini-1.5-flash-8b-exp-0827	1047	1072
gemini-1.5-flash-exp-0827	553	615
gemini-1.5-pro-001	317	294
gemini-1.5-pro-002	1836	1751
gemini-1.5-pro-exp-0827	518	485
gemma-2-27b-it	857	802
gemma-2-2b-it	1178	1213
gemma-2-9b-it	448	377
gemma-2-9b-it-simpo	228	211
glm-4-plus	939	992
gpt-4-0125-preview	784	793
gpt-4-1106-preview	368	352
gpt-4-turbo-2024-04-09	498	474
gpt-4o-2024-05-13	838	877
gpt-4o-2024-08-06	883	862
gpt-4o-mini-2024-07-18	816	931
grok-2-2024-08-13	1401	1405
grok-2-mini-2024-08-13	914	944
internlm2_5-20b-chat	1215	1091
jamba-1.5-large	226	189
jamba-1.5-mini	210	182
llama-3.1-405b-instruct-bf16	1432	1519
llama-3.1-405b-instruct-fp8	812	867
llama-3.1-70b-instruct	884	865
llama-3.1-8b-instruct	884	900
llama-3.1-nemotron-51b-instruct	230	226
llama-3.1-nemotron-70b-instruct	576	565
llama-3.2-1b-instruct	1075	1044
llama-3.2-3b-instruct	990	976
mistral-large-2407	1020	967
mixtral-8x22b-instruct-v0.1	68	97
o1-mini	1167	1236
o1-preview	1165	1201
phi-3-medium-4k-instruct	102	114
qwen-max-0919	1630	1582
qwen-plus-0828	1318	1340
qwen2-72b-instruct	190	174
qwen2.5-72b-instruct	919	926
reka-core-20240722	90	84
reka-core-20240904	735	709
reka-flash-20240722	86	98
reka-flash-20240904	719	711
yi-lightning	1131	1149
yi-lightning-lite	1681	1647

```
# 5. Overall models used (combining model_a and model_b)
all_models = pd.concat([df['model_a'], df['model_b']]).value_counts()
print("\nAll Models Distribution:")
print(all_models)
```

yi-lightning-lite	3328
qwen-max-0919	3212
llama-3.1-405b-instruct-bf16	2951
gemini-1.5-flash-8b-001	2863
grok-2-2024-08-13	2806
gemini-1.5-flash-002	2686
qwen-plus-0828	2658
claude-3-5-sonnet-20240620	2536
o1-mini	2403
gemma-2-2b-it	2391
o1-preview	2366
internlm2_5-20b-chat	2306

mistral-large-2401	1901
llama-3.2-3b-instruct	1966
glm-4-plus	1931
grok-2-mini-2024-08-13	1858
qwen2.5-72b-instruct	1845
llama-3.1-8b-instruct	1784
claude-3-haiku-20240307	1775
llama-3.1-70b-instruct	1749
gpt-4o-mini-2024-07-18	1747
gpt-4o-2024-08-06	1745
gpt-4o-2024-05-13	1715
llama-3.1-405b-instruct-fp8	1679
gemma-2-27b-it	1659
deepseek-v2.5	1636
gpt-4-0125-preview	1577
reka-core-20240904	1444
reka-flash-20240904	1430
chatgpt-4o-latest-20240808	1375
gemini-1.5-flash-exp-0827	1168
llama-3.1-nemotron-70b-instruct	1141
claude-3-5-sonnet-20241022	1085
gemini-1.5-pro-exp-0827	1003
gpt-4-turbo-2024-04-09	972
gemma-2-9b-it	825
gemini-1.5-flash-001	761
gpt-4-1106-preview	720
gemini-1.5-pro-001	611
llama-3.1-nemotron-51b-instruct	456
c4ai-aya-expense-32b	449
gemma-2-9b-it-simpo	439
jamba-1.5-large	415
jamba-1.5-mini	392
qwen2-72b-instruct	364
athene-70b-0725	338
deepseek-coder-v2-0724	232
phi-3-medium-4k-instruct	216
reka-flash-20240722	184
reka-core-20240722	174
mixtral-8x22b-instruct-v0.1	165
deepseek-v2-api-0628	131
Name: count, dtype: int64	

✓ Top 20 languages

```
# Get language distribution
language_counts = df['language'].value_counts()

# Get top 20 languages
top_20_languages = language_counts.head(20)


# Create a bar chart
plt.figure(figsize=(12, 8))
sns.barplot(x=top_20_languages.values, y=top_20_languages.index, palette='viridis')

# Add labels and title
plt.title('Top 20 Languages in the Dataset', fontsize=16)
plt.xlabel('Count', fontsize=12)
plt.ylabel('Language', fontsize=12)

# Add count values at the end of each bar
for i, v in enumerate(top_20_languages.values):
    plt.text(v + 0.5, i, str(v), va='center')

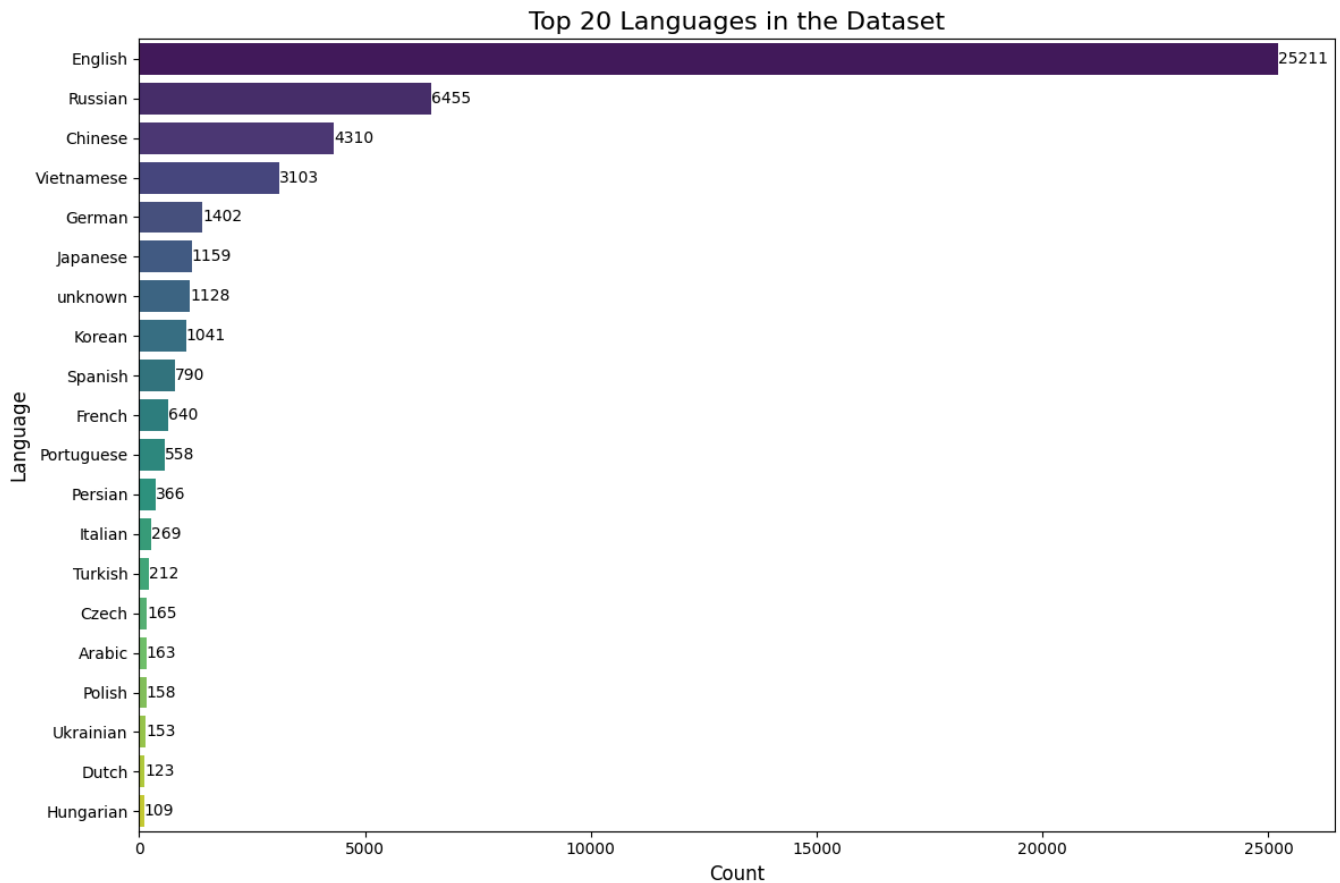
# Improve layout
plt.tight_layout()

# Show the plot
plt.show()
```

 <ipython-input-13-553f2e6d385a>:9: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `y` variable to `hue`

```
sns.barplot(x=top_20_languages.values, y=top_20_languages.index, palette='viridis')
```



✓ Model Distribution

```
# Get top models (if there are too many)
top_models = all_models.head(20) # Adjust number if needed


# Create a bar chart
plt.figure(figsize=(12, 8))
sns.barplot(x=top_models.values, y=top_models.index, palette='coolwarm')

# Add labels and title
plt.title('Model Distribution in the Dataset', fontsize=16)
plt.xlabel('Count', fontsize=12)
plt.ylabel('Model', fontsize=12)

# Add count values at the end of each bar
for i, v in enumerate(top_models.values):
    plt.text(v + 0.5, i, str(v), va='center')

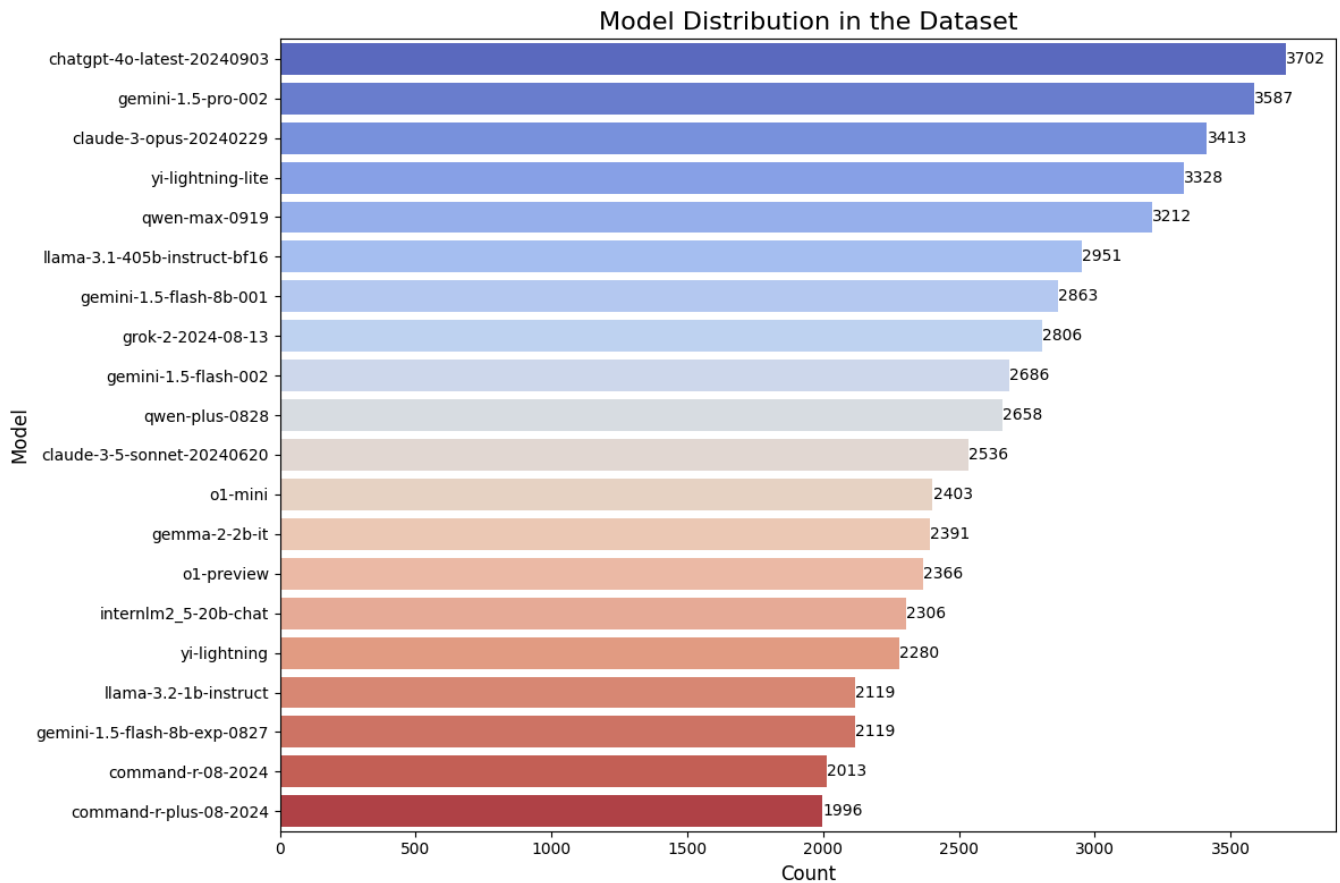
# Improve layout
plt.tight_layout()

# Show the plot
plt.show()
```

 <ipython-input-14-d0d1b01a5d28>:6: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `y` variable to `hue`

```
sns.barplot(x=top_models.values, y=top_models.index, palette='coolwarm')
```



✓ Word Count

```
# Function to count words
def count_words(text):
    if isinstance(text, str):
        return len(text.split())
    return 0 # Handle non-string entries

# Calculate word counts
df['prompt_word_count'] = df['prompt'].apply(count_words)
df['response_a_word_count'] = df['response_a'].apply(count_words)
df['response_b_word_count'] = df['response_b'].apply(count_words)

# Create combined text column (prompt + both responses)
df['combined_text'] = df['prompt'] + ' ' + df['response_a'] + ' ' + df['response_b']
df['combined_word_count'] = df['combined_text'].apply(count_words)

# Get descriptive statistics for word counts
word_count_stats = df[['prompt_word_count', 'response_a_word_count',
                        'response_b_word_count', 'combined_word_count']].describe()

print("Descriptive Statistics for Word Counts:")
print(word_count_stats)

# Create visualization for word count distributions
plt.figure(figsize=(14, 10))

# Prompt word count
plt.subplot(2, 2, 1)
sns.histplot(df['prompt_word_count'], kde=True, color='blue')
plt.title('Prompt Word Count Distribution')
plt.xlabel('Word Count')
```



```

plt.axvline(df['prompt_word_count'].mean(), color='red', linestyle='--',
            label=f'Mean: {df["prompt_word_count"].mean():.1f}')
plt.axvline(df['prompt_word_count'].median(), color='green', linestyle='-',
            label=f'Median: {df["prompt_word_count"].median():.1f}')
plt.legend()

# Response A word count
plt.subplot(2, 2, 2)
sns.histplot(df['response_a_word_count'], kde=True, color='orange')
plt.title('Response A Word Count Distribution')
plt.xlabel('Word Count')
plt.axvline(df['response_a_word_count'].mean(), color='red', linestyle='--',
            label=f'Mean: {df["response_a_word_count"].mean():.1f}')
plt.axvline(df['response_a_word_count'].median(), color='green', linestyle='-',
            label=f'Median: {df["response_a_word_count"].median():.1f}')
plt.legend()

# Response B word count
plt.subplot(2, 2, 3)
sns.histplot(df['response_b_word_count'], kde=True, color='green')
plt.title('Response B Word Count Distribution')
plt.xlabel('Word Count')
plt.axvline(df['response_b_word_count'].mean(), color='red', linestyle='--',
            label=f'Mean: {df["response_b_word_count"].mean():.1f}')
plt.axvline(df['response_b_word_count'].median(), color='green', linestyle='-',
            label=f'Median: {df["response_b_word_count"].median():.1f}')
plt.legend()

# Combined text word count
plt.subplot(2, 2, 4)
sns.histplot(df['combined_word_count'], kde=True, color='purple')
plt.title('Combined Text Word Count Distribution')
plt.xlabel('Word Count')
plt.axvline(df['combined_word_count'].mean(), color='red', linestyle='--',
            label=f'Mean: {df["combined_word_count"].mean():.1f}')
plt.axvline(df['combined_word_count'].median(), color='green', linestyle='-',
            label=f'Median: {df["combined_word_count"].median():.1f}')
plt.legend()

plt.tight_layout()
plt.show()

# Create boxplots to visualize the distribution and identify outliers
plt.figure(figsize=(14, 6))
word_counts_df = df[['prompt_word_count', 'response_a_word_count',
                    'response_b_word_count', 'combined_word_count']]
sns.boxplot(data=word_counts_df)
plt.title('Word Count Boxplots')
plt.ylabel('Word Count')
plt.xticks(rotation=45)
plt.tight_layout()
plt.show()

# Compare response lengths between winning and losing responses
df['winner_word_count'] = df.apply(
    lambda row: row['response_a_word_count'] if row['winner'] == 'A' else row['response_b_word_count'],
    axis=1
)
df['loser_word_count'] = df.apply(
    lambda row: row['response_b_word_count'] if row['winner'] == 'A' else row['response_a_word_count'],
    axis=1
)

print("\nWord Count Statistics for Winners vs Losers:")
print(df[['winner_word_count', 'loser_word_count']].describe())

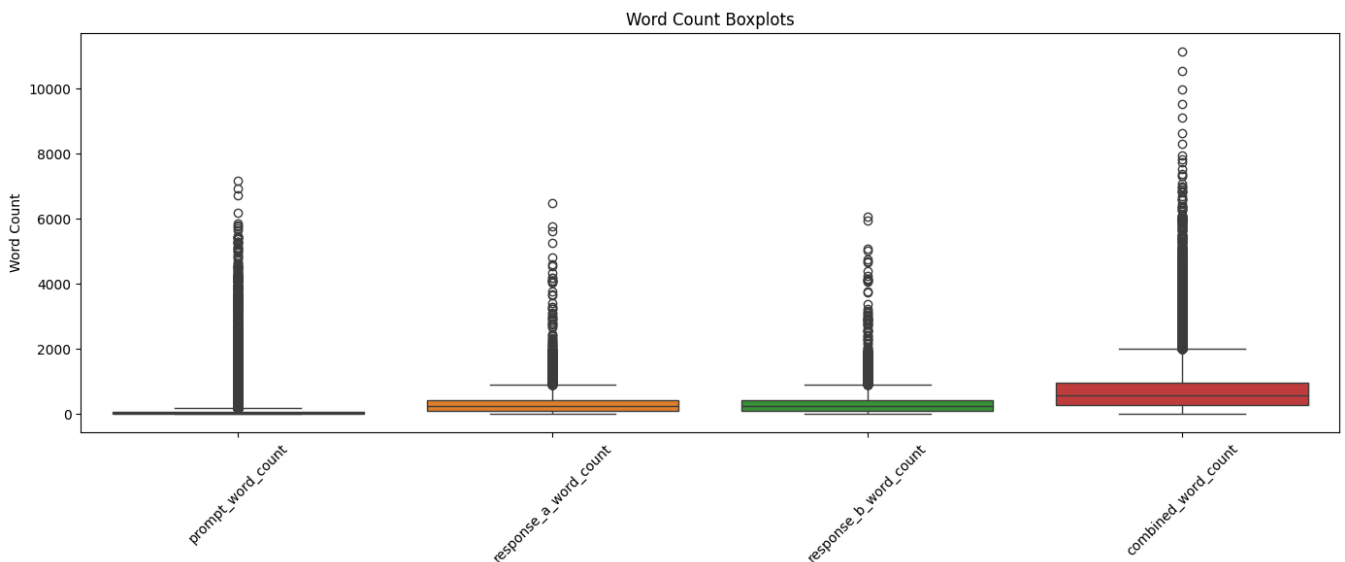
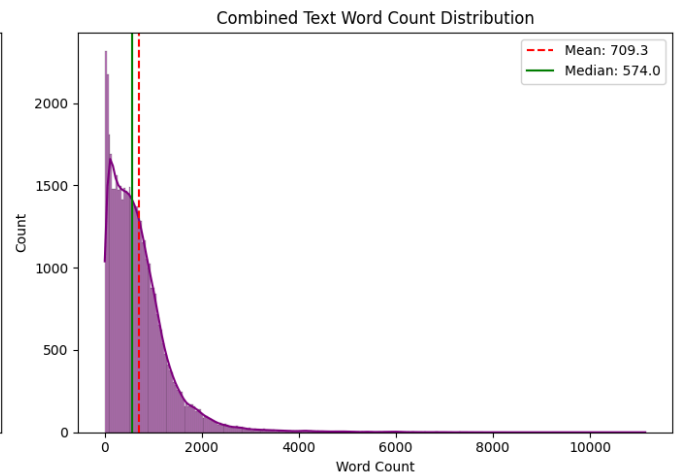
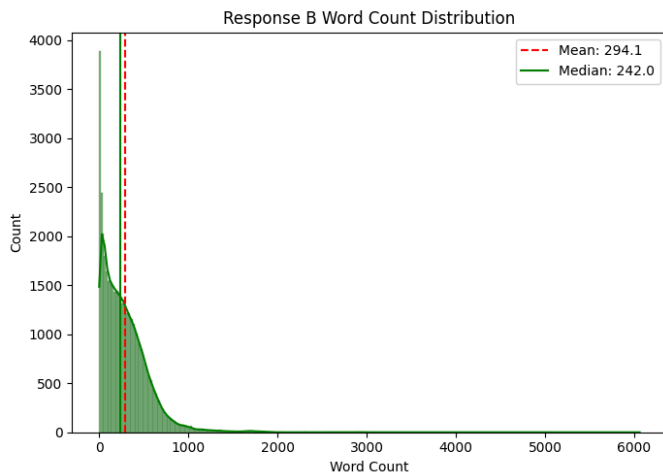
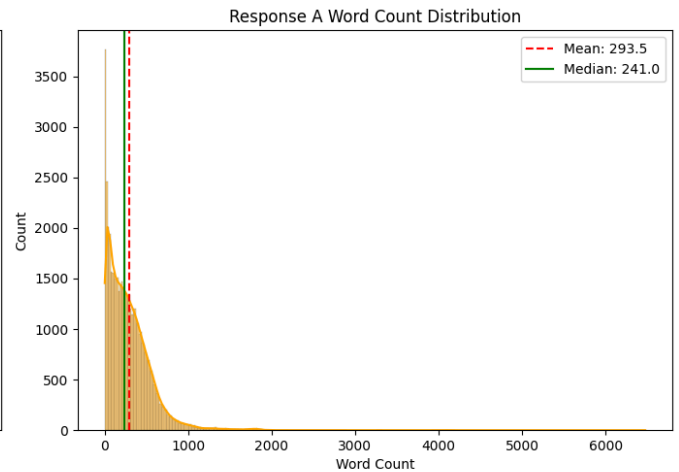
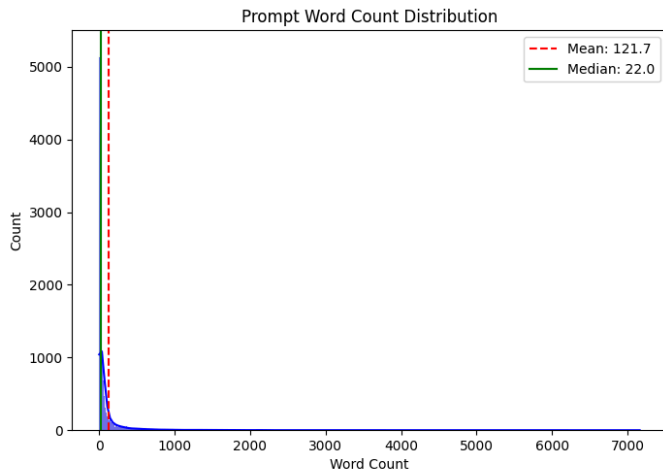
# Visualize winner vs loser word counts
plt.figure(figsize=(10, 6))
sns.boxplot(data=df[['winner_word_count', 'loser_word_count']])
plt.title('Word Count: Winners vs Losers')
plt.ylabel('Word Count')
plt.tight_layout()
plt.show()

```

Descriptive Statistics for Word Counts:

	prompt_word_count	response_a_word_count	response_b_word_count	\
count	48439.000000	48439.000000	48439.000000	
mean	121.658808	293.452714	294.143294	
std	364.760289	276.278870	275.319280	
min	0.000000	1.000000	1.000000	
25%	9.000000	94.000000	94.000000	
50%	22.000000	241.000000	242.000000	
75%	73.000000	419.000000	421.000000	
max	7160.000000	6476.000000	6061.000000	

	combined_word_count
count	48439.000000
mean	709.254815
std	667.944098
min	3.000000
25%	260.000000
50%	574.000000
75%	954.000000
max	11140.000000



Word Count Statistics for Winners vs Losers:

	winner_word_count	loser_word_count
count	48439.000000	48439.000000