Weight Sum of Helpfulness Metrics

Here we are going to create a function that prints out the importance of each helpfulness metric. Once we have the function, it can be used to look at the overall weights as well as weights for individual workers.

```
In [1]: import pandas as pd
        import numpy as np
        from sklearn.linear_model import LogisticRegression
        from sklearn.preprocessing import StandardScaler
        def calculate_metric_weights(df, worker_id=None):
            Calculate the weight of each helpfulness metric for a given writer.
            Input: csv file
            Output: dictionary of feature importance
            # Select relevant columns
            features = ['tone_similarity',
                                                 'pos_similarity',
                                                                         'coherence sco
            target = 'acceptance_status'
            # Filter for specific writer if worker id is provided
            if worker id:
                df = df[df['workerID'] == worker_id]
            # Drop rows with missing values
            df = df.dropna(subset=features + [target])
            # Convert target to binary, accepted is 1 and rejected is 0
            df[target] = df[target].apply(lambda x: 1 if x == 'accepted' else 0)
            if df[target].nunique() < 2:</pre>
                print(f"WorkerID: {worker_id} has only one class in target variable. S
                return {}
            # Standardize the features
            scaler = StandardScaler()
            X = scaler.fit_transform(df[features])
            y = df[target]
            # Train logistic regression model
            model = LogisticRegression()
            model.fit(X, y)
            # Get absolute values of coefficients and normalize them to sum to 1
            weights = np.abs(model.coef_[0])
            weights /= weights.sum()
            # Return dictionary of feature importance
            return dict(zip(features, weights))
```

```
In [2]: df = pd.read_csv("all_metrics.csv")

# Get overall weights
weights = calculate_metric_weights(df)
print(weights)

# Get weights for all writers
unique_workers = df['workerID'].unique()

for worker in unique_workers:
    weights = calculate_metric_weights(df, worker_id=worker)
    print(f'WorkerID: {worker}, Weights: {weights}')
```

```
{'tone_similarity': 0.06601884014650725, 'pos_similarity': 0.3046119144345303,
'coherence_score': 0.39001169399995467, 'user_score': 0.08892239417966982, 'ai
_score': 0.0759988759946788, 'ai_coverage': 0.0744362812446591}
WorkerID: A2WGW5Y3ZFBDEC, Weights: {'tone_similarity': 0.27247442149524514, 'p
os_similarity': 0.20853556418609345, 'coherence_score': 0.01804222935152794,
'user_score': 0.22152220273780235, 'ai_score': 0.04337483926572697, 'ai_covera
ge': 0.2360507429636043}
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r_score': 0.02289716796762064, 'ai_score': 0.060811963911849315, 'ai_coverag
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os_similarity': 0.044942835918885315, 'coherence_score': 0.36320778768291834,
'user score': 0.23219764267480042, 'ai score': 0.18592583436078933, 'ai covera
ge': 0.06124758820818416}
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os_similarity': 0.07971120610835707, 'coherence_score': 0.36003425262626326,
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r score': 0.03833218736474527, 'ai score': 0.03527534956691112, 'ai coverage':
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```

```
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ge': 0.13802758643152913}
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```

```
os_similarity': 0.11437187463238478, 'coherence_score': 0.2864065405479316, 'u
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WorkerID: A17Q4QN6UE0EZC has only one class in target variable. Skipping...
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s similarity': 0.09969714170438669, 'coherence score': 0.22306549925012273, 'u
ser score': 0.3998541839219472, 'ai score': 0.09883894665448084, 'ai coverag
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ge': 0.2710171752890695}
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e': 0.06533524708099557}
WorkerID: A173MXK429XAZQ has only one class in target variable. Skipping...
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ser_score': 0.03279793794079311, 'ai_score': 0.3493812251764123, 'ai_coverag
e': 0.22861631755515402}
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ge': 0.10433393939095109}
WorkerID: A377LTGWJKY2IW, Weights: { 'tone similarity': 0.006672610833185649,
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```

```
age': 0.4632518235572374}
        WorkerID: A3HE29W5IDR394 has only one class in target variable. Skipping...
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        ser_score': 0.09806894960865416, 'ai_score': 0.5117993692493543, 'ai_coverag
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        ge': 0.25742396824688296}
        WorkerID: A394J04NEPCY3M has only one class in target variable. Skipping...
        WorkerID: A394J04NEPCY3M, Weights: {}
In [3]: def calculate_weights_for_all_workers(df, unique_workers):
          Calculate the weight of each helpfulness metric for all writers.
          Input: csv file
          Output: dictionary of dictionary of feature importance
          worker weights = {}
          for worker in unique workers:
              weights = calculate metric weights(df, worker id=worker)
              worker_weights[worker] = weights
          return worker_weights
In [4]: import seaborn as sns
        import matplotlib.pyplot as plt
        worker_weights = calculate_weights_for_all_workers(df, unique_workers)
        df weights = pd.DataFrame.from dict(worker weights, orient='index')
        plt.figure(figsize=(12, 10))
        sns.heatmap(df_weights, annot=True, cmap="rocket_r", linewidths=0.5)
        plt.xlabel("Helpfulness Metrics")
        plt.ylabel("Worker ID")
        plt.title("Heatmap of Feature Weights Across Workers")
        plt.savefig('heatmap.png', dpi=300)
        plt.show()
        WorkerID: A17Q4QN6UE0EZC has only one class in target variable. Skipping...
        WorkerID: AFIK3VBMMX6G6 has only one class in target variable. Skipping...
        WorkerID: A1WKF2VH7TV0H2 has only one class in target variable. Skipping...
        WorkerID: A173MXK429XAZQ has only one class in target variable. Skipping...
        WorkerID: A6KOTWP7N7RLU has only one class in target variable. Skipping...
        WorkerID: A3HE29W5IDR394 has only one class in target variable. Skipping...
        WorkerID: A394J04NEPCY3M has only one class in target variable. Skipping...
```

Heatmap of Feature Weights Across Workers A2WGW5Y3ZFBDEC -0.018 0.5 AZCGF2D7QIO10 0.023 0.33 0.061 A345TDMHP3DQ3G 0.073 0.33 0.44 0.027 0.029 0.1 A2W121DQXNQK1 0.33 0.42 0.048 0.068 0.079 0.05 A324VBRLXHG5IB 0.095 A2ONILC0LZKG6Y 0.0056 0.32 0.076 0.0017 A30LRWACCCCUTU 0.081 0.1 0.24 A2EED3HLTA96CP 0.11 0.045 0.36 0.23 0.061 A2QKAA5YS0P4CI 0.08 0.36 0.099 0.039 0.21 A3S67QA0SQVPUJ 0.43 0.11 0.022 0.1 0.4 A1198W1SPF1R4 0.038 0.1 0.083 ANCIB6B6FBBII 0.41 0.038 0.035 0.27 AZZA3I049G7R5 0.22 0.28 A8C3WNWRBWUXO 0.075 0.28 0.0054 0.35 A1PTH9KTRO06EG 0.043 0.11 0.33 0.26 0.084 A2QX3YJXAAHHVV 0.022 0.33 A23KAJRDVCVGOE 0.28 0.096 0.21 0.03 A1QUQ0TV9KVD4C 0.053 0.25 0.21 0.089 0.3 A1VZSFHTU51JP0 0.033 0.088 A3MYPYBVHX7FQ2 0.027 0.3 0.23 0.063 0.055 ASVRLMDNQBUD9 0.049 0.34 0.052 0.34 0.0045 **A2YTQDLACTLIBA** 0.044 0.035 0.038 0.11 0.47 A1TW2BZRRS874Z 0.057 0.28 0.24 A2OVX9UW5WANOE 0.074 0.22 0.19 A2OVOVZBIYUO 0.029 0.1 0.3 A1FVXS8IM5OYO8 0.29 0.062 0.086 0.11 A3O5RKGH6VB19C 0.008 0.47 0.11 0.2 ABL2FXYMI00T6 0.048 0.07 0.015 0.3 0.4 AZLZA0Q87TJZO 0.32 0.066 0.0061 A3P9TM5PRYBH90 0.039 0.052 0.11 0.4 0.11 0.076 0.33 A118BQHK3S4UDV 0.2 0.041 0.26 0.094 A14OQ52EFQAN2W 0.29 A3DUPRZSMU9W5R 0.1 0.46 0.05 0.1 A3DS5B06ZCD3E3 0.065 0.08 0.19 0.39 0.042 A1E235KE3CSO7H 0.012 0.29 0.037 0.36 APRZ7BR8C0ZMQ 0.034 - 0.1 A2RUHO7I7Y4XFA 0.12 0.3 0.21 0.047 0.058 A3VEF4M5FIN7KH 0.027 0.42 0.065 0.02 A143XRCI1YXAFE 0.076 0.29 0.033 0.35 0.24 A23FWFNNOUS10B 0.088 0.27 0.0015 A1PBRKFHSF1OF8 0.027 0.36 0.2 0.085 0.1 A377LTGWIKY2IW 0.0067 0.017 0.46 AM2KK02IXXW48 0.055 0.066 0.51 0.098 A17AF42SNONH9C -0.062 0.24 tone_similarity pos_similarity coherence_score user_score ai_coverage ai_score Helpfulness Metrics

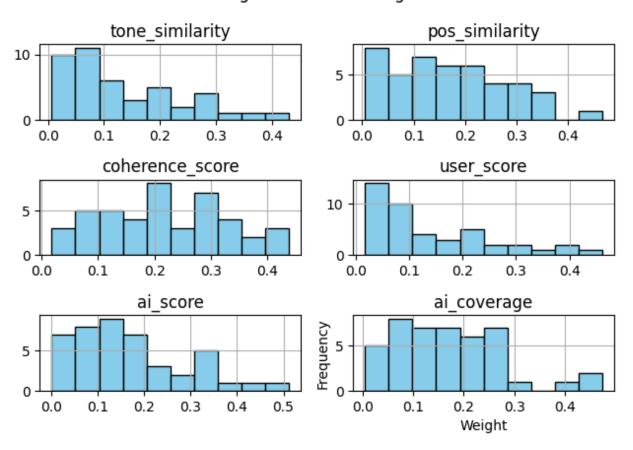
```
In [5]: # histogram of metric weights
   plt.figure(figsize=(4, 10))
   df_weights.hist(bins=10, color='skyblue', edgecolor='black')
   plt.suptitle('Histogram of Metric Weights', fontsize=12)
   plt.xlabel('Weight', fontsize=10)
   plt.ylabel('Frequency', fontsize=10)

#save img
   plt.savefig('metric_weights_histogram.png')

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

<Figure size 400x1000 with 0 Axes>

Histogram of Metric Weights



```
In [6]: from sklearn.cluster import KMeans
# Apply KMeans clustering
kmeans = KMeans(n_clusters=6, random_state=42)
df_weights['Cluster'] = kmeans.fit_predict(df_weights)

plt.figure(figsize=(10, 6))
sns.scatterplot(x=df_weights.iloc[:, 0], y=df_weights.iloc[:, 1], hue=df_weights.iloc[:, 1], hue=df_weights.iloc[:, 1], hue=df_weights.iloc[:, 1], hue=df_weights.iloc[:, 0], y=df_weights.iloc[:, 1], hue=df_weights.iloc[:, 1], hue=df_weights.iloc[:,
```



