

# SENAProject-Code.docx

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Identical Words	0
Words with Minor Changes	0
Paraphrased Words	0
Omitted Words	0



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## Results

Sources that matched your submitted document.

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#### IDENTICAL

Identical matches are one to one exact wording in the text.

#### MINOR CHANGES

Nearly identical with different form, ie "slow" becomes "slowly".

#### PARAPHRASED

Close meaning but different words used to convey the same message.

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### **Scanned Text**

ax.set\_xlim(0,45000)

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IDENTICAL MINOR CHANGES PARAPHRASED

Read, clean and structure data to make it suitable for analysis: from google.colab import drive drive.mount('/content/drive') import pandas as pd import numpy as np import matplotlib.pyplot as plt import seaborn as sns github\_data\_df = pd.read\_csv('/content/drive/MyDrive/Sena\_Project/Github\_data.csv') github\_data\_df.head() github\_data\_df = github\_data\_df.drop(['Unnamed: 0','Unnamed: 0.1'],axis=1) github\_df = github\_data\_df[['topic','name','user','star','fork','watch','issue','pull\_requests','topic \_tag','commits','contributers']] new\_names = ['Topic','Repo\_Name','User\_Name','Star','Fork','Watch','Issues','Pull\_Requests', 'Topic\_Tags','Commits','Contributors'] old\_names = github\_df.columns github\_df = github\_df.rename(columns=dict(zip(old\_names, new\_names))) github\_df['Star'] = github\_df['Star'].apply(lambda x: float(x.rstrip('k'))\*1000 if x.endswith('k') else float(x)) github\_df['Fork'] = github\_df['Fork'].apply(lambda x: float(x.rstrip('k'))\*1000 if x.endswith('k ') else float(x)) github\_df['Watch'] = github\_df['Watch'].apply(lambda x: float(x.rstrip('k'))\*1000 if 'k' in x el se float(x)) cols = ['Issues','Pull\_Requests','Commits','Contributors'] github\_df[cols] = github\_df[cols].apply(pd.to\_numeric, errors='coerce', axis=1) Analysis 1: Top repositories based on popularity pop\_mean\_df = github\_df.groupby('Topic').mean().reset\_index() 1.1 Analysis of stars fig, ax = plt.subplots(figsize=(6,4), dpi=100) plt.rcParams['axes.edgecolor']='#333F4B' ax.spines['top'].set\_visible(False) ax.spines['right'].set\_visible(False) ax.spines['left'].set\_visible(False) ax.tick\_params(axis='both', which='both', labelsize=10, bottom=True, left=False)

```
ax.set facecolor('white')
sns.barplot(data=pop_mean_df, x='Star', y='Topic');
ax.set_xlabel('Stars', fontsize=13, color = '#333F4B')
ax.set_ylabel('Topic', fontsize=13, color = '#333F4B')
fig.suptitle('Average stars on each topic',fontsize=18, color = '#333F4B');
github_df.nlargest(n=10, columns='Star')[['Repo_Name','Topic','Star']]
print('Most starred repository:')
print('Repository Name: ',github_df.iloc[github_df['Star'].idxmax()]['Repo_Name'])
print('Topic: ',github_df.iloc[github_df['Star'].idxmax()]['Topic'])
print('Star: ',github_df.iloc[github_df['Star'].idxmax()]['Star'])
1.2 Analysis of watch
fig, ax = plt.subplots(figsize=(6,4), dpi=100)
plt.rcParams['axes.edgecolor']='#333F4B'
ax.spines['top'].set_visible(False)
ax.spines['right'].set_visible(False)
ax.spines['left'].set_visible(False)
ax.tick_params(axis='both', which='both', labelsize=10, bottom=True, left=False)
ax.set_xlim(0,1600)
ax.grid(False)
ax.set_facecolor('white')
sns.barplot(data=pop_mean_df, x='Watch', y='Topic');
ax.set_xlabel('Watchers', fontsize=13, color = '#333F4B')
ax.set_ylabel('Topic', fontsize=13, color = '#333F4B')
fig.suptitle('Average watchers on each topic',fontsize=18, color = '#333F4B');
github_df.nlargest(n=10, columns='Watch')[['Repo_Name','Topic','Watch']]
print('Most watched repository:')
print('Repository Name: ',github_df.iloc[github_df['Watch'].idxmax()]['Repo_Name'])
print('Topic: ',github_df.iloc[github_df['Watch'].idxmax()]['Topic'])
print('Watch: ',github_df.iloc[github_df['Watch'].idxmax()]['Watch'])
1.3 Analysis of fork
fig, ax = plt.subplots(figsize=(6,4), dpi=100)
plt.rcParams['axes.edgecolor']='#333F4B'
ax.spines['top'].set_visible(False)
ax.spines['right'].set_visible(False)
ax.spines['left'].set_visible(False)
ax.tick_params(axis='both', which='both', labelsize=10, bottom=True, left=False)
ax.set_xlim(0,8000)
ax.grid(False)
ax.set_facecolor('white')
sns.barplot(data=pop_mean_df, x='Fork', y='Topic');
ax.set_xlabel('Forks', fontsize=13, color = '#333F4B')
ax.set_ylabel('Topic', fontsize=13, color = '#333F4B')
fig.suptitle('Average forks on each topic',fontsize=18, color = '#333F4B');
github_df.nlargest(n=10, columns='Fork')[['Repo_Name','Topic','Fork']]
print('Most forked repository:')
```

ax.grid(False)

```
print('Repository Name: ',github_df.iloc[github_df['Fork'].idxmax()]['Repo_Name'])
print('Topic: ',github_df.iloc[github_df['Fork'].idxmax()]['Topic'])
print('Fork: ',github_df.iloc[github_df['Fork'].idxmax()]['Fork'])
Relationship between Star, Fork and Watch
fig, ax = plt.subplots(figsize=(8,4), dpi=100)
```

```
sns.set_theme('paper')
sns.regplot(data=github_df, x='Star', y='Fork', color='purple');
ax.set_xlabel('Star', fontsize=13, color = '#333F4B')
```

```
ax.set_ylabel('Fork', fontsize=13, color = '#333F4B')
fig.suptitle('Relationship between Star and Fork',fontsize=18, color = '#333F4B');
fig, ax = plt.subplots(figsize=(8,4), dpi=100)
sns.set theme('paper')
sns.regplot(data=github_df, x='Watch', y='Fork', color='purple');
ax.set_xlabel('Watch', fontsize=13, color = '#333F4B')
ax.set ylabel('Fork', fontsize=13, color = '#333F4B')
fig.suptitle('Relationship between Watch and Fork',fontsize=18, color = '#333F4B');
Analysis 2: Contribution activities using issues, pull requests, commits, and
contributors across the repositories
corr_df = github_df.dropna(axis=0, subset = ['Issues','Pull_Requests','Commits','Contributors'
])[['Issues','Pull_Requests','Commits','Contributors']]
fig, ax = plt.subplots(figsize=(6,4), dpi=100)
sns.heatmap(corr_df.corr(), linewidths=0.1, vmax=1.0, square=True, linecolor='white', annot
=True, cmap='winter');
fig.suptitle('Correlation between the contribution columns',fontsize=16, color = '#333F4B');
popular_df = github_df.nlargest(n=100,columns=['Star'])
[['Issues','Pull_Requests','Commits','Contributors']]
fig, ax = plt.subplots(figsize=(6,4), dpi=100)
sns.heatmap(popular_df.corr(), linewidths=0.1, vmax=1.0, square=True, linecolor='white', an
not=True, cmap='winter');
fig.suptitle('Correlation of contributions in Top 100 popular repositories',fontsize=16, color =
'#333F4B');
users_with_more_repos = github_df.groupby('User_Name').size().nlargest(n=10).reset_index
(name='Count')['User_Name'].to_list()
more_repos_users_df = github_df[github_df['User_Name'].isin(users_with_more_repos)]
[['Issues','Pull_Requests','Commits','Contributors']]
fig, ax = plt.subplots(figsize=(6,4), dpi=100)
sns.heatmap(more_repos_users_df.corr(), linewidths=0.1, vmax=1.0, square=True, linecolor=
'white', annot=True, cmap='summer');
fig.suptitle('Correlation of contributions among users with more repositories',fontsize=16, col
or = '#333F4B');
Analysis 3: Topic Tags
import ast
from collections import Counter
topic_tags = github_df['Topic_Tags'].apply(lambda x: ast.literal_eval(x)).tolist()
all_tags = [tag for topic in topic_tags for tag in topic]
tags_dict = Counter(all_tags)
toptags_df = pd.DataFrame(tags_dict.most_common(15), columns=['Name of the Tag', 'Count
fig, ax = plt.subplots(figsize=(7,4), dpi=100)
plt.xticks(rotation=90)
ax.grid(False)
ax.set_facecolor('white')
sns.despine()
sns.barplot(data=toptags_df, x='Name of the Tag', y='Count', palette='twilight_shifted');
ax.set_xlabel('Topic Tags', fontsize=13, color = '#333F4B')
```

```
plt.xticks(rotation=90)
ax.grid(False)
ax.set_facecolor('white')
sns.despine()
sns.barplot(data=toptags_df, x='Name of the Tag', y='Count', palette='twilight_shifted');
ax.set_xlabel('Topic Tags', fontsize=13, color = '#333F4B')
ax.set_ylabel('Count', fontsize=13, color = '#333F4B')
fig.suptitle('Most popular topic tags',fontsize=18, color = '#333F4B');
len_tags = [len(tag) for tag in topic_tags]
github_df['Total_Tags'] = len_tags
topic_wise_tags = github_df.groupby('Topic').sum()['Total_Tags'].reset_index(name='Total T ags')
fig, ax = plt.subplots(figsize=(7,4), dpi=100)
ax.grid(False)
ax.set_facecolor('white')
sns.despine()
sns.barplot(data=topic_wise_tags,x='Total Tags', y='Topic', ci=None, palette='gist_rainbow');
ax.set_xlabel('Total Tags', fontsize=13, color = '#333F4B')
ax.set_ylabel('Topic', fontsize=13, color = '#333F4B')
```

fig.suptitle('Tags distribution across topics',fontsize=18, color = '#333F4B');