

Codagami Anime Recommender

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Data Mining Project

Group-15

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Data Source

MyAnimeList

Grand Prize WEB NOVEL FINALISTS Vote Now!

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99+


chinmay453

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Fate/Zero

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Top > Anime > Fate/Zero

SCORE

8.31

716,491 users

Ranked #219Popularity #67Members 1,276,140

Fall 2011TVufotable

Completed(10) MasterpieceEpisodes: 13/13

* Your list is public by default.

Synopsis

With the promise of granting any wish, the omnipotent Holy Grail triggered three wars in the past, each too cruel and fierce to leave a victor. In spite of that, the wealthy Einzbern family is confident that the Fourth Holy Grail War will be different; namely, with a vessel of the Holy Grail now in their grasp. Solely for this reason, the much hated "Magus Killer" Kiriotsugu Emiya is hired by the Einzberns, with marriage to their only daughter Irisviel as binding contract.

Kiriotsugu now stands at the center of a cutthroat game of survival, facing off against six other participants, each armed with an ancient familiar, and fueled by unique desires and ideals. Accompanied by his own familiar, Saber, the notorious mercenary soon finds his greatest opponent in Kirei Kotomine, a priest who seeks salvation from the emptiness within himself in pursuit of Kiriotsugu.

Based on the light novel written by Gen Urobuchi, *Fate/Zero* depicts the events of the Fourth Holy Grail War—10 years prior to *Fate/stay night*. Witness a battle royale in which no one is guaranteed to survive.

[Written by MAL Rewrite]

Background

Fate/Zero was simulcasted around the world, with subtitles covering eight major languages. It is currently licensed by Aniplex of America for release in North America.

Edit Status

* Your list is public by default.


Status:Completed

Eps Seen:13 / 13

Your Score:(10) Masterpiece

UpdateEdit Details

Remove from Favorites



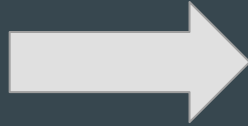
More videos

Edit

Edit

Data retrieval and processing

```
{  
  "anime_list_final_231.json"  
}  
1  
2  
3   "1": {  
4     "id": 1,  
5     "title": "Cowboy Bebop",  
6     "main_picture": {  
7       "medium": "https://api-cdn.myanimelist.net/im...",  
8       "large": "https://api-cdn.myanimelist.net/im...",  
9     },  
10    "alternative_titles": {  
11      "synonyms": [],  
12      "en": "Cowboy Bebop",  
13      "ja": "カウボーイビバップ"  
14    },  
15    "start_date": "1998-04-03",  
16    "end_date": "1999-04-24",  
17    "synopsis": "In the year 2071, humanity has expanded to other planets, but the Earth is still the center of the world. In the year 2071, humanity has expanded to other planets, but the Earth is still the center of the world. In the year 2071, humanity has expanded to other planets, but the Earth is still the center of the world.",  
18    "mean": 8.77,  
19    "rank": 33,  
20    "popularity": 45,  
21  }
```



id	title	main_picture	alternative_titles	start_date	end_date
1.0	Cowboy Bebop	{'medium': 'https://api-cdn.myanimelist.net/im...'	{'synonyms': [], 'en': 'Cowboy Bebop', 'ja': 'カウボーイビバップ'}	1998-04-03	1999-04-24
6.0	Trigun	{'medium': 'https://api-cdn.myanimelist.net/im...'	{'synonyms': [], 'en': 'Trigun', 'ja': 'トライガン'}	1998-04-01	1998-09-30
8.0	Bouken Ou Beet	{'medium': 'https://api-cdn.myanimelist.net/im...'	{'synonyms': ['Adventure King Beet'], 'en': 'Bouken Ou Beet'}	2004-09-30	2005-09-29
16.0	Hachimitsu to Clover	{'medium': 'https://api-cdn.myanimelist.net/im...'	{'synonyms': ['HachiKuro', 'Honey & Clover'], 'en': 'Honey & Clover'}	2005-04-15	2005-09-27
18.0	Initial D Fourth Stage	{'medium': 'https://api-cdn.myanimelist.net/im...'	{'synonyms': ['Initial D 4th Stage'], 'en': 'Initial D Fourth Stage'}	2004-04-17	2006-02-18
s x 33 columns					

Json Data collected from MAL API

Data converted and processed to pandas DataFrame for analysis

ALS based Recommender

Alternating Least Square (ALS) Matrix Factorization based Collaborative Filtering

Make recommendations

Enter your anime name in my_anime_list, and the recommender will generate recommendations

```
my_anime_list = ['Cowboy Bebop']

recommends = make_recommendation(
    best_model_params={'iterations': 10, 'rank': 20, 'lambda_': 0.05},
    ratings_data=rating_data,
    df_animes=animes,
    fav_anime_list=my_anime_list,
    n_recommendations=10,
    spark_context=sc)

print('Recommendations for {}'.format(my_anime_list[0]))
for i, title in enumerate(recommends):
    print('{0}: {1}'.format(i+1, title))
```

[15] ✓ 11.6s

...

Recommendations for Cowboy Bebop:

- 1: Fullmetal Alchemist: Brotherhood
- 2: Hunter x Hunter (2011)
- 3: Shingeki no Kyojin Season 3
- 4: Shokugeki no Souma: San no Sara
- 5: "Violet Evergarden: Kitto ""Ai"" wo Shiru Hi ga Kuru no Darou"
- 6: Seishun Buta Yarou wa Yumemiru Shoujo no Yume wo Minai
- 7: Death Note
- 8: Kimi no Na wa.
- 9: Saiki Kusuo no Ψ-nan 2
- 10: Shingeki no Kyojin Season 3 Part 2

TF-IDF based Recommender

Term Frequency — Inverse Document Frequency (TF-IDF) based recommendation systems are content based recommenders

```
def recommend_anime(title, max_reco = 10, cosine_sim = cos_sim, cosine_sim_g
    print("Anime Recommendations for: "+title)
    recommended_animes = []
    index = anime_names[anime_names == title].index[0]

    similar_scores = pd.Series(cosine_sim[index])
    similar_scores_genre = pd.Series(cosine_sim_genre[index])
    mean = anime_df['mean']
    mean = mean.apply(lambda x: x if x >= 7 else 0)
    anime_mean_score = pd.Series(np.array(mean))

    similar_scores_mul = similar_scores.mul(similar_scores_genre)

    similar_scores_mul = similar_scores_mul.mul(anime_mean_score)
    similar_scores_mul = similar_scores_mul.sort_values(ascending=False)

    top_animes = list(similar_scores_mul.iloc[1:max_reco+1].index)
    for anime_index in top_animes:
        anime_row = anime_df.iloc[anime_index]
        anime_name = anime_row['title']
        recommended_animes.append(anime_name)
    return recommended_animes
```

```
recommend_anime("Cowboy Bebop")
```

✓ 56.1s

Anime Recommendations for: Cowboy Bebop

```
['Cowboy Bebop: Yose Atsume Blues',
 'Seihou Bukyou Outlaw Star',
 'Cowboy Bebop: Tengoku no Tobira',
 'Space Adventure Cobra',
 'Waga Seishun no Arcadia: Mugen Kidou SSX',
 'Ginga Tetsudou Monogatari',
 'Iria: Zeiram The Animation',
 'Uchuu Senkan Yamato',
 'Freedom',
 'Sayonara Ginga Tetsudou 999: Andromeda Shuuchakueki']
```

KNN based Recommender

K- Nearest Neighbours (KNN) item based collaborative filtering

```
reverse_mapper = {v: k for k, v in mapper.items()}\nprint('Recommendations for {}'.format(fav_anime))\nfor i, (idx, dist) in enumerate(raw_recommends):\n    print('{0}: {1}, with distance of {2}'.format(i+1, reverse_mapper[idx], dist))
```

✓ 0.1s

```
my_favorite = 'Cowboy Bebop'
```

```
make_recommendation(\n    model_knn=model_knn,\n    data=anime_user_mat_sparse,\n    fav_anime=my_favorite,\n    mapper=anime_to_idx,\n    n_recommendations=10)
```

✓ 0.4s

You have input anime: Cowboy Bebop

Found possible matches in our database: ['Cowboy Bebop']

Recommendation system start to make inference

.....

Recommendations for Cowboy Bebop:

1: FLCL, with distance of 0.31903903051406135

2: Akira, with distance of 0.31690885637703625

3: Monster, with distance of 0.3168795683168413

4: Samurai Champloo, with distance of 0.3042900227126648

5: Bleach, with distance of 0.30396440374767253

6: Cowboy Bebop: Tengoku no Tobira, with distance of 0.30382183444633903

7: Neon Genesis Evangelion: The End of Evangelion, with distance of 0.30313470443520196

8: Death Note, with distance of 0.28743321445921133

9: Fullmetal Alchemist: Brotherhood, with distance of 0.28349179908653377

10: Neon Genesis Evangelion, with distance of 0.28263286514808483

SVD based Recommender

```
def cosine_similarity_sort(r_data, anime_id, top_n=10):  
    ind = anime_id - 1  
    anime_row = r_data[ind, :]  
    magnitude = np.sqrt(np.einsum('ij, ij -> i', r_data, r_data))  
    matrix_similarity = np.dot(anime_row, r_data.T) / (magnitude[ind] * magnitude)  
    sorted_indices = np.argsort(-matrix_similarity)  
    return sorted_indices[:top_n]
```

```
def get_most_similar_anime(anime_df, anime_id, index_list):  
    print('Best recommendations for {0}: \n'.format(  
        anime_df[anime_df.anime_id == anime_id].title.values[0]))  
    for id in index_list + 1:  
        print(anime_df[anime_df.anime_id == id].title.values[0])
```

```
k = 50          #k-principal components to represent anime, anime_id to find recommendat  
anime_id = 2    # id for which we want recommendation  
top_n = 10  
rep_data = V.T[:, :k] # representative data  
index_list = cosine_similarity_sort(rep_data, anime_id, top_n)  
  
#Get the top N recommendations  
get_most_similar_anime(anime_df, anime_id, index_list)
```

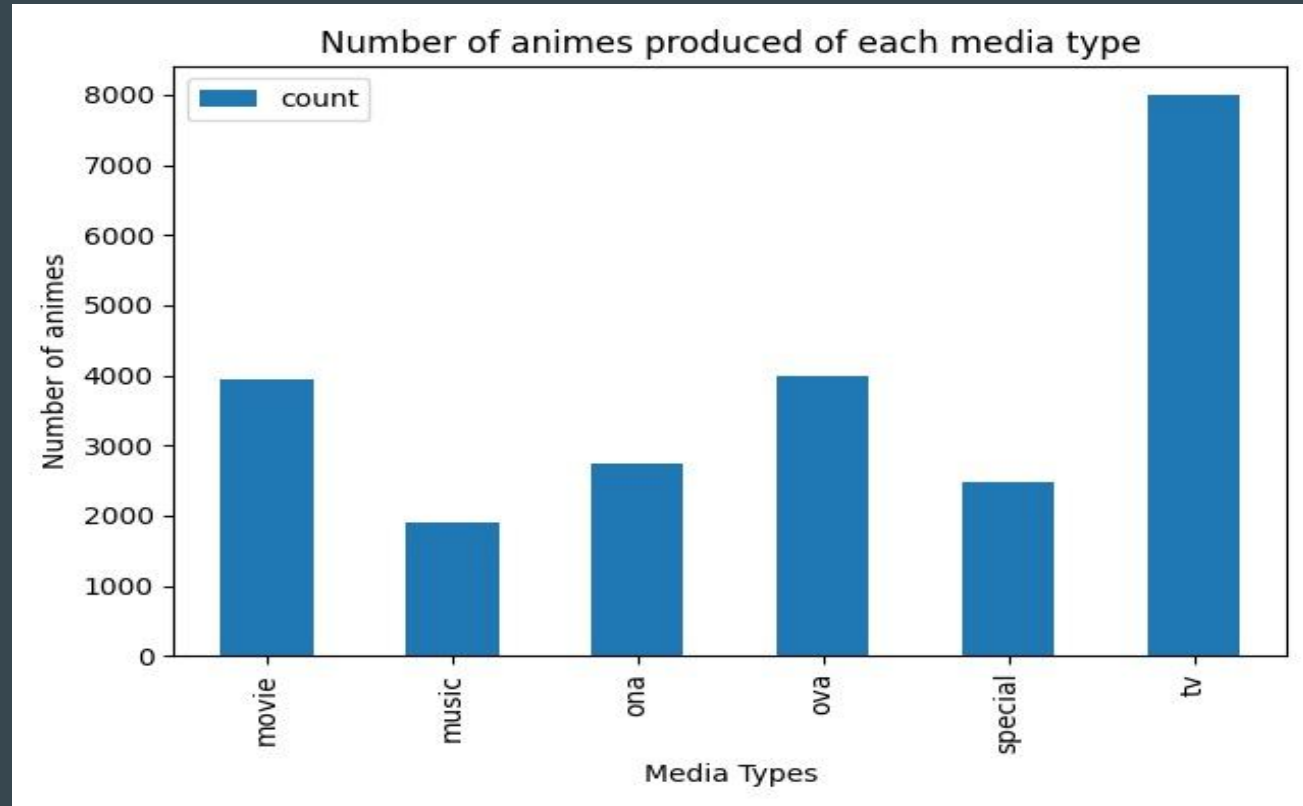
Best recommendations for Cowboy Bebop:

Initial D First Stage
Trigun
El Hazard: The Alternative World
Mobile Suit Gundam SEED
Beck
Saiyuuiki Reload Gunlock
Yakitate!! Japan
Hunter x Hunter: Greed Island Final
Pita Ten
Green Green

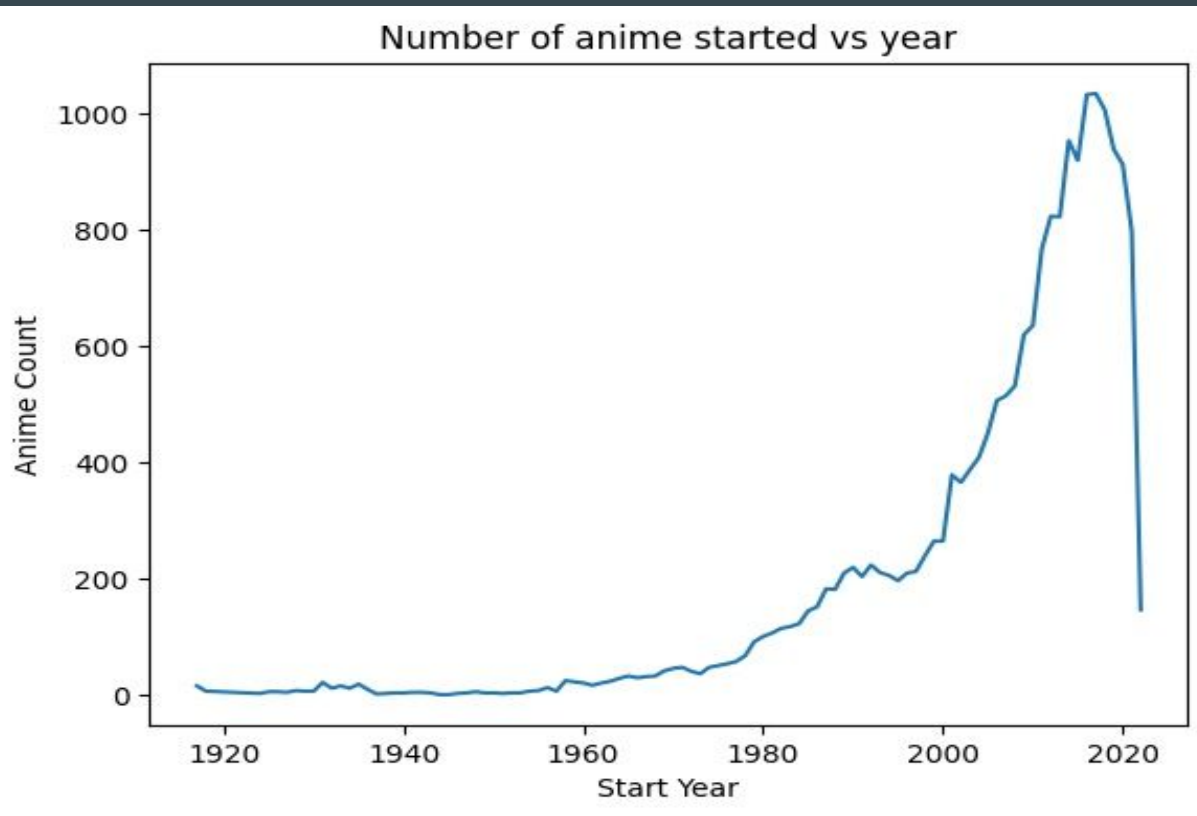
Insights and Analysis

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Number of animes produced of each media type

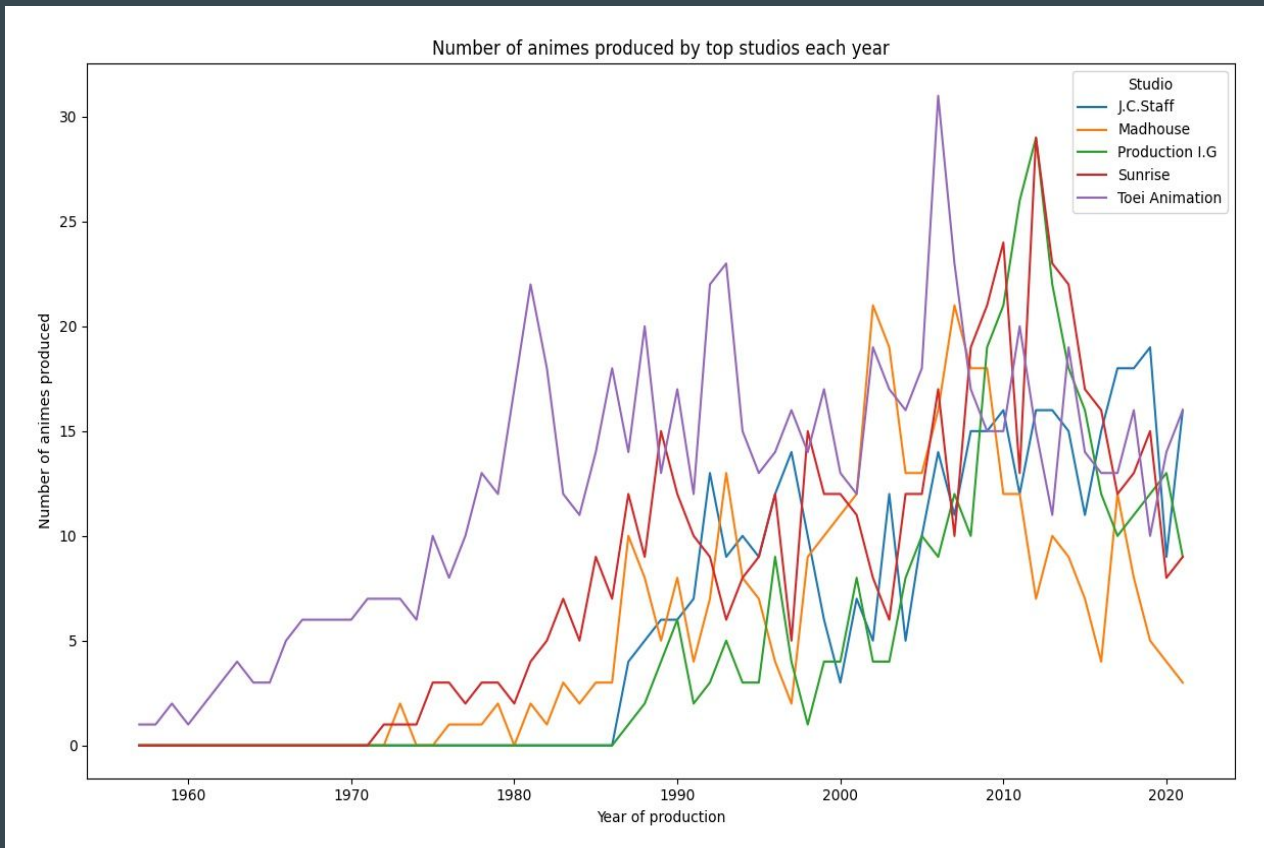


Number of animes aired with time



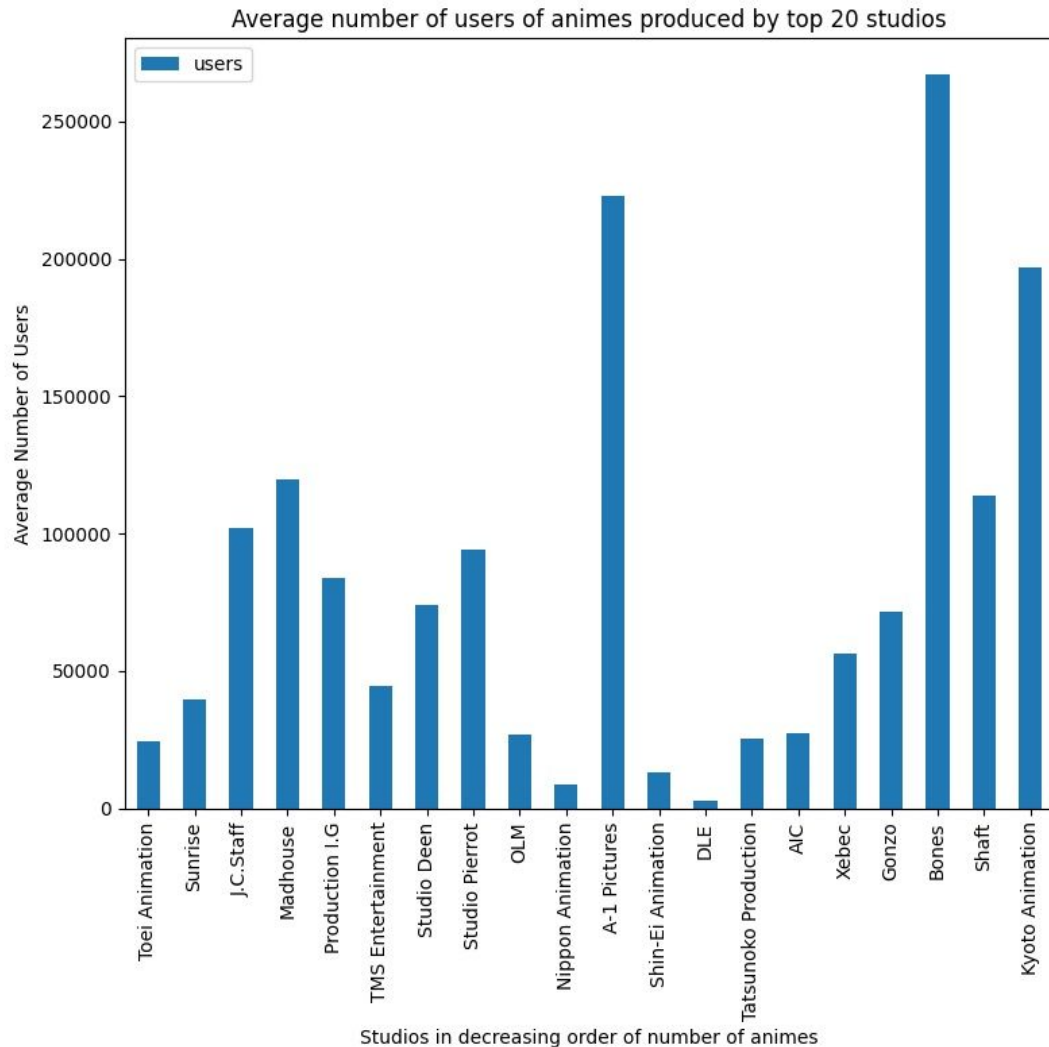
The dip in the number of animes started during 2020–21 is due to the COVID-19 pandemic.

Number of animes produced by top studios with time

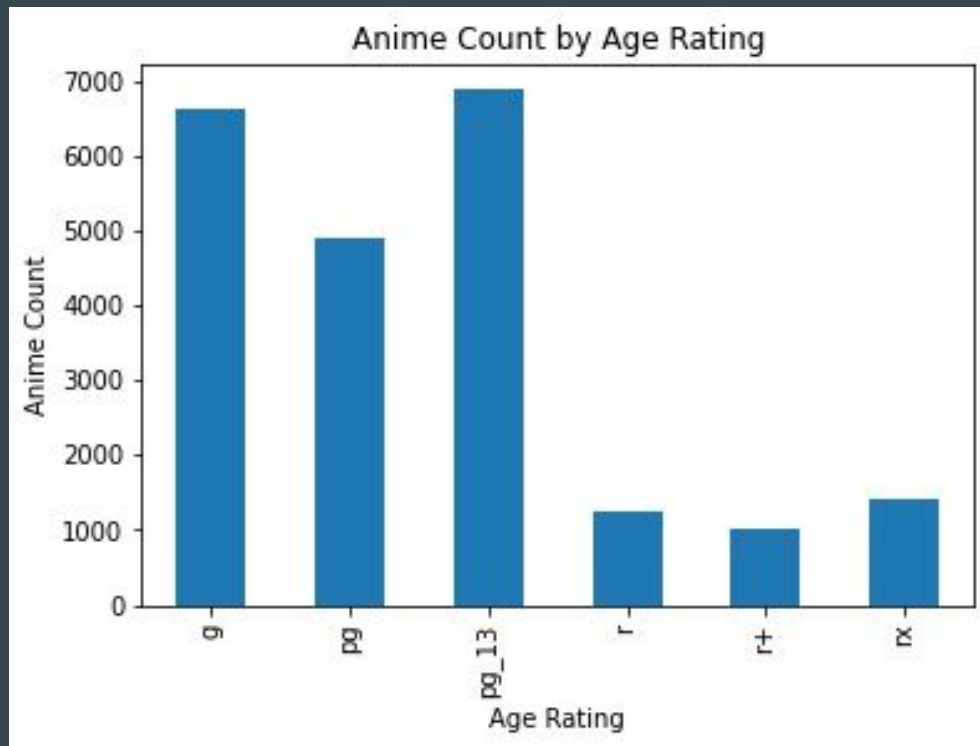


Viewership of Top Studios

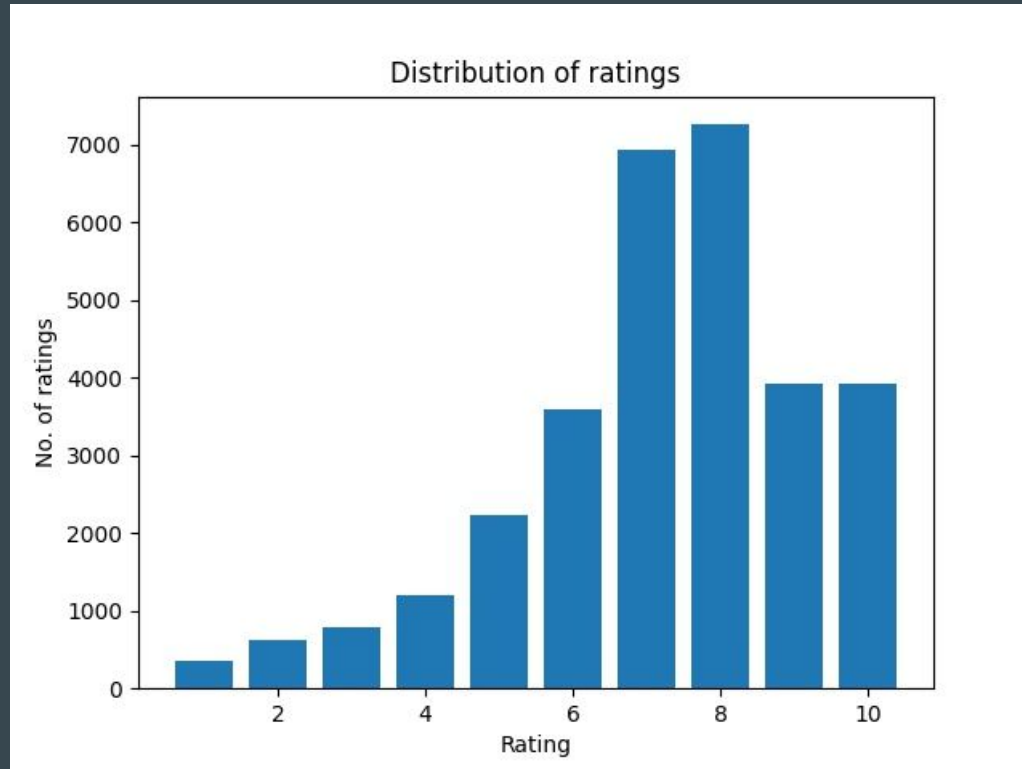
Bones has the largest viewership among top studios



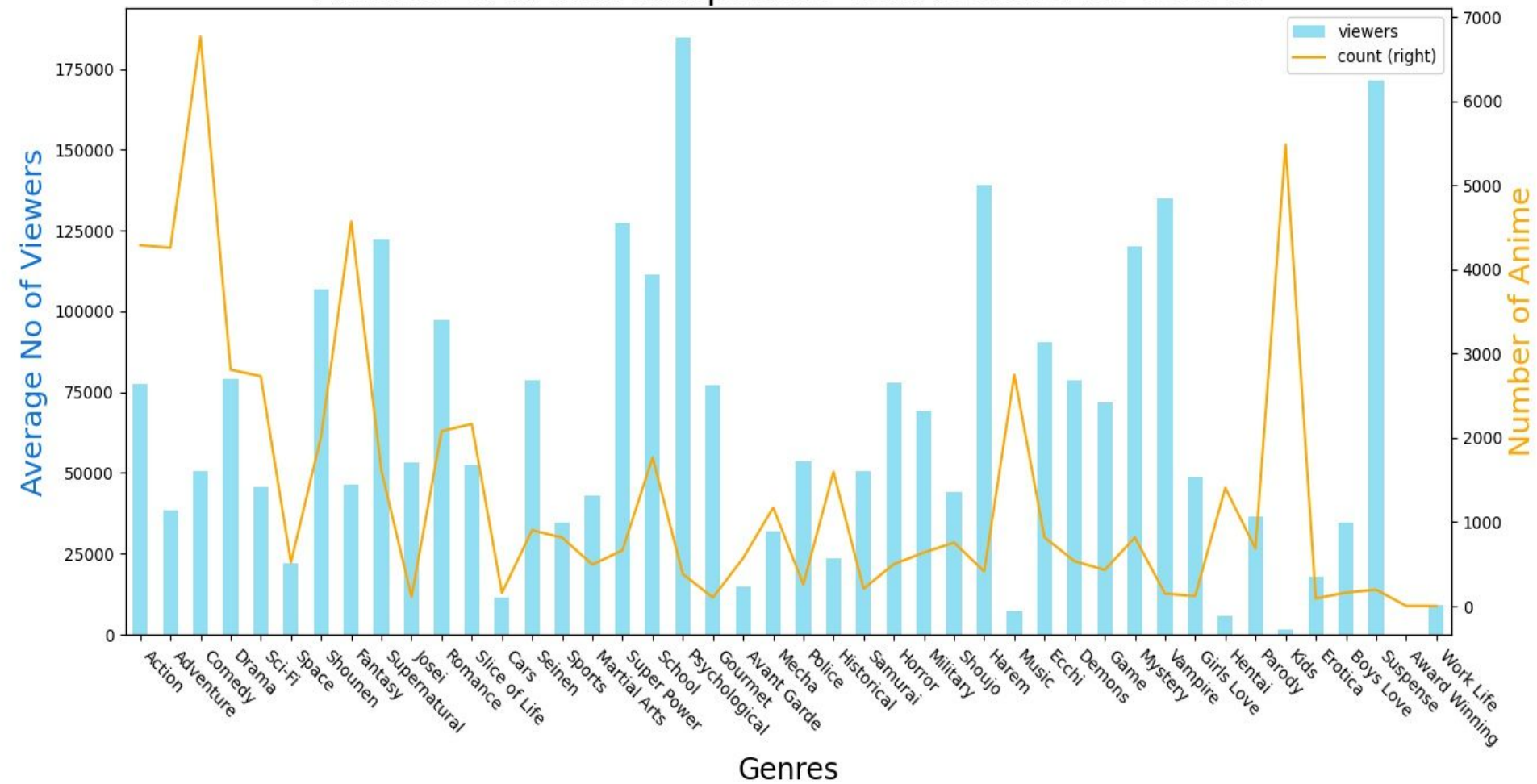
Distribution of animes based on age rating



Frequency of various ratings given by viewers



Number of Anime comparison with Viewers for Genres



Arigatō Gozaimasu!