# netflix analysis

#### June 11, 2025

```
[3]: import pandas as pd
     import numpy as np
     import matplotlib.pyplot as plt
     import seaborn as sns
     from wordcloud import WordCloud
     from sklearn.preprocessing import MultiLabelBinarizer
     from datetime import datetime
     import warnings
     #Configuration
     warnings.filterwarnings('ignore')
     pd.set_option('display.max_columns', None)
     import seaborn as sns
     sns.set_theme(style="whitegrid", palette="muted")
     sns.set_theme(style="whitegrid", palette="muted")
     COLOR_PALETTE = ['#E50914', '#221F1F', '#B81D24', '#F5F5F1', '#000000']
     #DATA LOADING & CLEANING
     def load_and_clean_data():
         """Load and preprocess Netflix data with comprehensive cleaning"""
             df = pd.read_csv('netflix.csv')
            print(" Data loaded successfully (rows: {}, cols: {})".format(*df.
      ⇔shape))
         except FileNotFoundError:
            raise FileNotFoundError(" Error: File 'netflix.csv' not found")
         # Date processing
         df['date_added'] = pd.to_datetime(df['date_added'], errors='coerce')
         df['year_added'] = df['date_added'].dt.year
         df['month_added'] = df['date_added'].dt.month_name()
         # Country processing (extract primary country)
         df['country'] = df['country'].fillna('Unknown').apply(
             lambda x: x.split(',')[0].strip() if isinstance(x, str) else 'Unknown')
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# Duration processing
   df['duration'] = df['duration'].fillna('0')
   df['duration_min'] = df.apply(
        lambda x: int(x['duration'].split()[0]) if 'min' in x['duration'] else__
 →np.nan, axis=1)
   df['duration seasons'] = df.apply(
        lambda x: int(x['duration'].split()[0]) if 'Season' in x['duration']
 ⇔else np.nan, axis=1)
    # Genre processing
   df['genres'] = df['listed_in'].str.split(', ').fillna('').apply(
        lambda x: [genre.strip() for genre in x])
    # Director/actor processing
   df['directors'] = df['director'].str.split(', ').fillna('')
   df['actors'] = df['cast'].str.split(', ').fillna('')
    # Content rating standardization
   rating_map = {
        'TV-MA': 'Adult',
        'TV-14': 'Teen',
        'TV-PG': 'Older Kids',
        'R': 'Adult',
        'PG-13': 'Teen',
        'PG': 'Older Kids',
        'G': 'Kids',
        'NC-17': 'Adult',
        'TV-Y7': 'Older Kids',
        'TV-Y7-FV': 'Older Kids',
        'TV-Y': 'Kids',
        'NR': 'Adult',
        'UR': 'Adult'
   }
   df['rating_category'] = df['rating'].map(rating_map).fillna('Other')
   return df
df = load_and_clean_data()
#EXPLORATORY ANALYSIS
def comprehensive_eda(df):
   print("\n COMPREHENSIVE EXPLORATORY ANALYSIS")
    # Basic statistics
   print("\n BASIC STATISTICS:")
   print(f"- Dataset spans {df['release_year'].min()} to {df['release_year'].
 →max()}")
   print(f"- Latest addition: {df['date_added'].max().strftime('%Y-%m-%d')}")
   print(f"- {df['type'].value_counts(normalize=True).to_string()}")
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# Missing values analysis
   print("\n MISSING VALUES:")
   missing = df.isnull().sum().sort_values(ascending=False)
   print(missing[missing > 0].to_string())
   # Temporal analysis
   print("\n TEMPORAL PATTERNS:")
   print(f"- Peak release year: {df['release year'].mode()[0]}")
   print(f"- Peak addition year: {df['year_added'].mode()[0]}")
    # Content characteristics
   print("\n CONTENT CHARACTERISTICS:")
   print(f"- Avg movie duration: {df['duration min'].mean():.1f} minutes")
   print(f"- Avg show seasons: {df['duration_seasons'].mean():.1f}")
   return df
df = comprehensive_eda(df)
#ADVANCED VISUALIZATIONS
def create_advanced_visualizations(df):
    """Generate sophisticated visualizations with insights"""
   plt.figure(figsize=(20, 18))
    # 1. Content Type Evolution Over Time
   plt.subplot(3, 2, 1)
   type year = df.groupby(['release year', 'type']).size().unstack()
   type_year.plot(kind='area', stacked=True, color=['#E50914', '#221F1F'],__
 ⇒alpha=0.8, ax=plt.gca())
   plt.title('Content Type Evolution (1925-2021)', fontsize=12)
   plt.xlabel('Release Year')
   plt.ylabel('Count')
   plt.legend(title='Content Type')
   # 2. Monthly Addition Patterns
   plt.subplot(3, 2, 2)
   month_order = ['January', 'February', 'March', 'April', 'May', 'June',
                  'July', 'August', 'September', 'October', 'November',
 monthly = df['month_added'].value_counts().reindex(month_order)
   sns.barplot(x=monthly.index, y=monthly.values, palette='viridis')
   plt.title('Monthly Addition Patterns', fontsize=12)
   plt.xlabel('Month')
   plt.ylabel('Titles Added')
   plt.xticks(rotation=45)
    # 3. Movie Duration Distribution
   plt.subplot(3, 2, 3)
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sns.boxplot(x='rating_category', y='duration_min', data=df,
                palette='rocket', order=['Kids', 'Older Kids', 'Teen', 'Adult'])
   plt.title('Movie Duration by Rating Category', fontsize=12)
   plt.xlabel('Rating Category')
   plt.ylabel('Duration (minutes)')
    # 4. Top Production Countries
   plt.subplot(3, 2, 4)
    countries = df['country'].value_counts().nlargest(10).sort_values()
    countries.plot(kind='barh', color='#E50914')
   plt.title('Top 10 Production Countries', fontsize=12)
   plt.xlabel('Number of Titles')
    # 5. Genre Word Cloud
   plt.subplot(3, 2, 5)
   genres_text = ' '.join([' '.join(genres) for genres in df['genres']])
   wordcloud = WordCloud(width=800, height=400, background_color='white',
                         colormap='Reds').generate(genres_text)
   plt.imshow(wordcloud, interpolation='bilinear')
   plt.title('Genre Word Cloud', fontsize=12)
   plt.axis('off')
   # 6. Content Addition Growth
   plt.subplot(3, 2, 6)
   additions = df.groupby('year_added').size().cumsum()
   additions.plot(color='#E50914', marker='o')
   plt.title('Cumulative Content Growth', fontsize=12)
   plt.xlabel('Year')
   plt.ylabel('Total Titles')
   plt.tight_layout()
   plt.suptitle('Netflix Content: Advanced Analytics Dashboard', y=1.02, ___
 ⇔fontsize=16)
   plt.savefig('netflix_advanced_dashboard.png', dpi=300, bbox_inches='tight')
   plt.show()
create_advanced_visualizations(df)
#DEEP DIVE ANALYSIS
def perform_deep_dive_analysis(df):
    """Execute sophisticated analytical techniques"""
   print("\n DEEP DIVE ANALYSIS")
    # Genre Analysis
   mlb = MultiLabelBinarizer()
   genre_matrix = pd.DataFrame(mlb.fit_transform(df['genres']),
                              columns=mlb.classes_,
                              index=df.index)
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top_genre_combos = genre_matrix.sum().sort_values(ascending=False).head(10)
    # Content Longevity Analysis
   df['content_age'] = df['year_added'] - df['release_year']
   longevity_stats = df.groupby('type')['content_age'].describe()
    # Seasonal Analysis
   seasonal = df.groupby(['month_added', 'type']).size().unstack()
   # Print insights
   print("\n TOP GENRES:")
   print(top_genre_combos.to_string())
   print("\n CONTENT LONGEVITY (years between release and addition):")
   print(longevity_stats.to_string())
   print("\n SEASONAL ADDITION PATTERNS:")
   print(seasonal.idxmax().to_string())
   return {
        'genre_matrix': genre_matrix,
        'longevity_stats': longevity_stats,
        'seasonal_patterns': seasonal
analysis_results = perform_deep_dive_analysis(df)
#TREND ANALYSIS & FORECASTING
def analyze_trends_and_forecast(df):
    """Perform time series analysis and simple forecasting"""
   print("\n TREND ANALYSIS & FORECASTING")
    # Prepare time series data
   ts = df.groupby(['year_added', 'type']).size().unstack().fillna(0)
   ts['Total'] = ts.sum(axis=1)
   # Calculate growth rates
   ts['Growth_Rate'] = ts['Total'].pct_change() * 100
   ts['3Yr_Avg_Growth'] = ts['Growth_Rate'].rolling(3).mean()
   # Simple forecasting (linear extrapolation of 3-year growth)
   last_year = int(ts.index.max()) # FIXED
   forecast years = 3
   forecast = pd.DataFrame(index=range(last_year, last_year + forecast_years +_u
 →1))
   forecast['Total'] = [ts.loc[last_year, 'Total'] *
                    (1 + ts.loc[last_year, '3Yr_Avg_Growth']/100)**i
                    for i in range(forecast_years + 1)]
    # Plot historical and forecasted data
   plt.figure(figsize=(12, 6))
```

```
plt.plot(ts.index, ts['Total'], 'o-', label='Historical', color='#E50914')
   ⇔color='#221F1F')
   plt.fill between(forecast.index, forecast['Total']*0.9, forecast['Total']*1.
 →1,
                   color='gray', alpha=0.1)
   plt.title('Netflix Content Growth: Historical & Forecast', fontsize=14)
   plt.xlabel('Year')
   plt.ylabel('Total Titles')
   plt.legend()
   plt.grid(True, alpha=0.3)
   plt.savefig('netflix_growth_forecast.png', dpi=300, bbox_inches='tight')
   print(f"\n INSIGHT: Based on 3-year average growth of {ts.loc[last_year,__

¬'3Yr_Avg_Growth']:.1f}%")
   print(forecast.to string())
   return ts, forecast
ts_data, forecast = analyze_trends_and_forecast(df)
#SAVE RESULTS
def save_analysis_results(df, analysis_results):
    """Save all analysis outputs systematically"""
   timestamp = datetime.now().strftime("%Y%m%d_%H%M%S")
   # Save cleaned data
   df.to_csv(f'netflix_cleaned_{timestamp}.csv', index=False)
   # Save analysis results
   with pd.ExcelWriter(f'netflix analysis {timestamp}.xlsx') as writer:
       df.describe().to_excel(writer, sheet_name='Summary Stats')
       analysis_results['genre_matrix'].sum().sort_values(ascending=False).
 →to_excel(
           writer, sheet_name='Genre Analysis')
       analysis results['longevity stats'].to excel(
           writer, sheet_name='Content Longevity')
       ts_data.to_excel(writer, sheet_name='Time Series Data')
       forecast.to_excel(writer, sheet_name='Growth Forecast')
   print(f"\n Results saved with timestamp: {timestamp}")
save_analysis_results(df, analysis_results)
print("\n Analysis completed successfully! ")
```

Data loaded successfully (rows: 8807, cols: 12)

COMPREHENSIVE EXPLORATORY ANALYSIS

BASIC STATISTICS:

- Dataset spans 1925 to 2021 - Latest addition: 2021-09-25

- type

Movie 0.696151 TV Show 0.303849

### MISSING VALUES:

duration\_seasons 6131 duration\_min 2679 director 2634 825 cast year\_added 98 month\_added 98 98 date\_added 4 rating

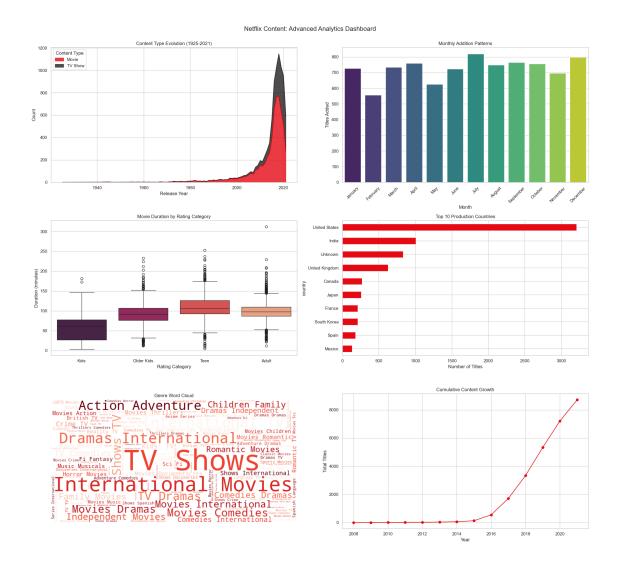
#### TEMPORAL PATTERNS:

Peak release year: 2018Peak addition year: 2019.0

### CONTENT CHARACTERISTICS:

- Avg movie duration: 99.6 minutes

- Avg show seasons: 1.8



### DEEP DIVE ANALYSIS

### TOP GENRES:

International Movies	2752
Dramas	2427
Comedies	1674
International TV Shows	1351
Documentaries	869
Action & Adventure	859
TV Dramas	763
Independent Movies	756
Children & Family Movies	641
Romantic Movies	616

CONTENT LONGEVITY (years between release and addition):

count mean std min 25% 50% 75% max

type

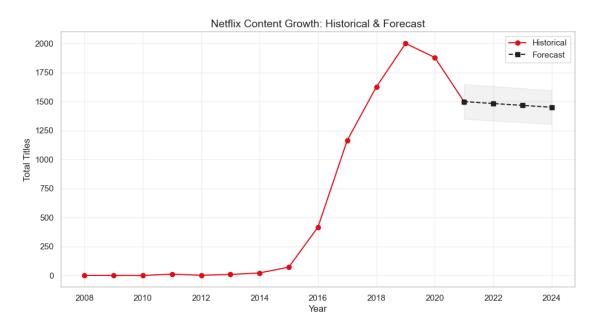
Movie 6131.0 5.727777 9.742631 -1.0 0.0 2.0 7.0 75.0 TV Show 2578.0 2.224981 5.175744 -3.0 0.0 0.0 2.0 93.0

#### SEASONAL ADDITION PATTERNS:

type

Movie July TV Show July

### TREND ANALYSIS & FORECASTING



INSIGHT: Based on 3-year average growth of -1.1%

Total

2021 1498.000000

2022 1481.662144

2023 1465.502477

2024 1449.519054

Results saved with timestamp: 20250509\_232136

Analysis completed successfully!

## []: