

BookReview_EDA_Satyajit_Ghana

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Book Review - An Exploratory Data Analysis and Data Transformation Notebook
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```
[191]: import gdown

url = 'https://drive.google.com/uc?id=1UPZiTughL3iDtPwreoUs_SX-LfVktrI3'
output = 'BX-CSV-Dump.zip'
gdown.download(url, output, quiet=False)
```

Downloading...

From: https://drive.google.com/uc?id=1UPZiTughL3iDtPwreoUs_SX-LfVktrI3

To: /content/BX-CSV-Dump.zip

26.1MB [00:00, 156MB/s]

```
[191]: 'BX-CSV-Dump.zip'
```

```
[2]: ! unzip BX-CSV-Dump.zip
```

Archive: BX-CSV-Dump.zip

inflating: BX-Book-Ratings.csv

inflating: BX-Books.csv

inflating: BX-Users.csv

1 Book Crossing EDA

```
[3]: %matplotlib inline

import scipy
import seaborn as sns
import matplotlib.pyplot as plt
import pandas as pd
import numpy as np

from sklearn.preprocessing import StandardScaler, MinMaxScaler

sns.set()
palette = sns.color_palette("icefire")
```

```
plt.style.use('ggplot')

sns.set_context("talk")
```

1.1 Loading, Cleaning and Merging the Dataset

Reading the csv files

```
[4]: users = pd.read_csv(
    '/content/BX-Users.csv',
    names=['user_id', 'location', 'age'],
    sep=';',
    skiprows=1,
    encoding='ISO-8859-1',
    low_memory=False,
    error_bad_lines=False
)
users
```

```
[4]:
```

	user_id	location	age
0	1	nyc, new york, usa	NaN
1	2	stockton, california, usa	18.0
2	3	moscow, yukon territory, russia	NaN
3	4	porto, v.n.gaia, portugal	17.0
4	5	farnborough, hants, united kingdom	NaN
...
278853	278854	portland, oregon, usa	NaN
278854	278855	tacoma, washington, united kingdom	50.0
278855	278856	brampton, ontario, canada	NaN
278856	278857	knoxville, tennessee, usa	NaN
278857	278858	dublin, n/a, ireland	NaN

[278858 rows x 3 columns]

parse the datatypes properly

```
[5]: users.dtypes
```

```
[5]: user_id      int64
location      object
age           float64
dtype: object
```

```
[6]: users.describe().T
```

```
[6]:
```

	count	mean	std	...	50%	75%
max						
user_id	278858.0	139429.500000	80499.515020	...	139429.5	209143.75
278858.0						
age	168096.0	34.751434	14.428097	...	32.0	44.00

244.0

[2 rows x 8 columns]

age cannot be 244! so let's fix that

```
[7]: users.loc[(users.age > 100) | (users.age < 5), 'age'] = np.nan
users.age = users.age.fillna(users.age.mean())
```

```
[8]: users['age'] = users['age'].astype(np.uint8)
```

```
[9]: users['age'].describe()
```

```
[9]: count    278858.000000
mean         34.446733
std          10.551712
min           5.000000
25%          29.000000
50%          34.000000
75%          35.000000
max          100.000000
Name: age, dtype: float64
```

```
[10]: users.isna().sum()
```

```
[10]: user_id    0
location    0
age         0
dtype: int64
```

```
[11]: books = pd.read_csv(
    '/content/BX-Books.csv',
    names=['isbn', 'book_title', 'book_author', 'year_of_publication',
    'publisher', 'img_s', 'img_m', 'img_l'],
    sep=';',
    skiprows=1,
    encoding='ISO-8859-1',
    low_memory=False,
    error_bad_lines=False
)
books
```

```
[11]:
```

	isbn	...	img_l
0	0195153448	...	http://images.amazon.com/images/P/0195153448.0...
1	0002005018	...	http://images.amazon.com/images/P/0002005018.0...
2	0060973129	...	http://images.amazon.com/images/P/0060973129.0...
3	0374157065	...	http://images.amazon.com/images/P/0374157065.0...
4	0393045218	...	http://images.amazon.com/images/P/0393045218.0...
...
271374	0440400988	...	http://images.amazon.com/images/P/0440400988.0...
271375	0525447644	...	http://images.amazon.com/images/P/0525447644.0...
271376	006008667X	...	http://images.amazon.com/images/P/006008667X.0...

```
271377 0192126040 ... http://images.amazon.com/images/P/0192126040.0...
271378 0767409752 ... http://images.amazon.com/images/P/0767409752.0...
```

[271379 rows x 8 columns]

parse the data types properly

```
[12]: books.dtypes
```

```
[12]: isbn                object
book_title              object
book_author             object
year_of_publication     object
publisher               object
img_s                   object
img_m                   object
img_l                   object
dtype: object
```

drop ['img_s', 'img_m', 'img_l'] since they are not useful for us

```
[13]: books = books.drop(['img_s', 'img_m', 'img_l'], axis=1)
```

year_of_publication should be a integer

```
[14]: books['year_of_publication'] = pd.to_numeric(books['year_of_publication'],
→errors='coerce')
```

```
[15]: books.loc[(books['year_of_publication'] == 0) | (books['year_of_publication'] >
→2008), 'year_of_publication'] = np.nan
books['year_of_publication'] = books['year_of_publication'].
→fillna(round(books['year_of_publication'].mean()))
books['year_of_publication'] = pd.to_numeric(books['year_of_publication'],
→downcast='unsigned')
```

```
[16]: books.isna().sum()
```

```
[16]: isbn                0
book_title              0
book_author             1
year_of_publication     0
publisher               2
dtype: int64
```

```
[17]: books = books.dropna()
```

```
[18]: books.describe().T
```

```
[18]:
```

	count	mean	std	...	50%	75%
max						
year_of_publication	271376.0	1993.692427	8.248715	...	1995.0	2000.0
2008.0						

[1 rows x 8 columns]

```
[19]: ratings = pd.read_csv(
        '/content/BX-Book-Ratings.csv',
        names=['user_id', 'isbn', 'book_rating'],
        sep=';',
        skiprows=1,
        encoding='ISO-8859-1',
        low_memory=False,
        error_bad_lines=False
    )
ratings
```

```
[19]:
```

	user_id	isbn	book_rating
0	276725	034545104X	0
1	276726	0155061224	5
2	276727	0446520802	0
3	276729	052165615X	3
4	276729	0521795028	6
...
1149775	276704	1563526298	9
1149776	276706	0679447156	0
1149777	276709	0515107662	10
1149778	276721	0590442449	10
1149779	276723	05162443314	8

[1149780 rows x 3 columns]

```
[20]: ratings['book_rating'] = ratings['book_rating'].astype(np.uint8)
```

```
[21]: ratings.dtypes
```

```
[21]: user_id      int64
       isbn       object
       book_rating  uint8
       dtype: object
```

```
[22]: ratings.isna().sum()
```

```
[22]: user_id      0
       isbn       0
       book_rating  0
       dtype: int64
```

```
[23]: ratings.describe().T.astype(np.int32)
```

```
[23]:
```

	count	mean	std	min	25%	50%	75%	max
user_id	1149780	140386	80562	2	70345	141010	211028	278854
book_rating	1149780	2	3	0	0	0	7	10

Join the three datasets based on user_id and isbn

```
[24]: temp = pd.merge(users, ratings, on='user_id')
       temp = pd.merge(temp, books, on='isbn')
```

```
dataset = temp.copy()
```

```
[25]: dataset
```

```
[25]:      user_id  ...      publisher
0           2  ...  Oxford University Press
1           8  ...    HarperFlamingo Canada
2        11400  ...    HarperFlamingo Canada
3        11676  ...    HarperFlamingo Canada
4        41385  ...    HarperFlamingo Canada
...         ...  ...
1031167  278851  ...    Simon & Schuster
1031168  278851  ...    Broadway Books
1031169  278851  ...    Lone Star Books
1031170  278851  ...    Kqed Books
1031171  278851  ...  American Map Corporation
```

```
[1031172 rows x 9 columns]
```

Split the location into city, state and country and replacing missing location details with just n/a

```
[26]: location = dataset['location'].str.split(' ', n=2, expand=True)
location.columns = ['city', 'state', 'country']
location = location.fillna('n/a')
```

```
[27]: dataset['city'] = location['city'] ; dataset['state'] = location['state'] ;
      dataset['country'] = location['country']
```

```
[28]: dataset = dataset.drop(['location'], axis=1)
```

```
[29]: dataset.describe().T.astype(np.int32)
```

```
[29]:      count    mean    std  ...    50%    75%    max
user_id    1031172  140594  80524  ...  141210  211426  278854
age         1031172     36    10  ...    34    41    100
book_rating  1031172     2     3  ...     0     7    10
year_of_publication  1031172  1995     7  ...  1997  2001  2008
```

```
[4 rows x 8 columns]
```

```
[30]: dataset.isna().sum()
```

```
[30]: user_id      0
age          0
isbn         0
book_rating  0
book_title   0
book_author  0
year_of_publication  0
publisher    0
city         0
```

```
state          0
country        0
dtype: int64
```

```
[31]: dataset.shape
```

```
[31]: (1031172, 11)
```

```
[32]: dataset.dtypes
```

```
[32]: user_id          int64
age                uint8
isbn              object
book_rating        uint8
book_title         object
book_author        object
year_of_publication uint16
publisher          object
city              object
state             object
country           object
dtype: object
```

This will be the final dataset we will be working with !

```
[33]: dataset.head(5)
```

```
[33]:  user_id  age  isbn  ...  city  state country
0         2   18  0195153448  ...  stockton  california  usa
1         8   34  0002005018  ...  timmins   ontario  canada
2      11400   49  0002005018  ...  ottawa   ontario  canada
3      11676   34  0002005018  ...    n/a     n/a    n/a
4      41385   34  0002005018  ...  sudbury   ontario  canada
```

```
[5 rows x 11 columns]
```

```
[34]: dataset.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 1031172 entries, 0 to 1031171
Data columns (total 11 columns):
#   Column                Non-Null Count  Dtype
---  -
0   user_id               1031172 non-null  int64
1   age                  1031172 non-null  uint8
2   isbn                 1031172 non-null  object
3   book_rating          1031172 non-null  uint8
4   book_title           1031172 non-null  object
5   book_author          1031172 non-null  object
6   year_of_publication  1031172 non-null  uint16
7   publisher            1031172 non-null  object
8   city                 1031172 non-null  object
```

```

9    state          1031172 non-null object
10   country        1031172 non-null object
dtypes: int64(1), object(7), uint16(1), uint8(2)
memory usage: 74.7+ MB

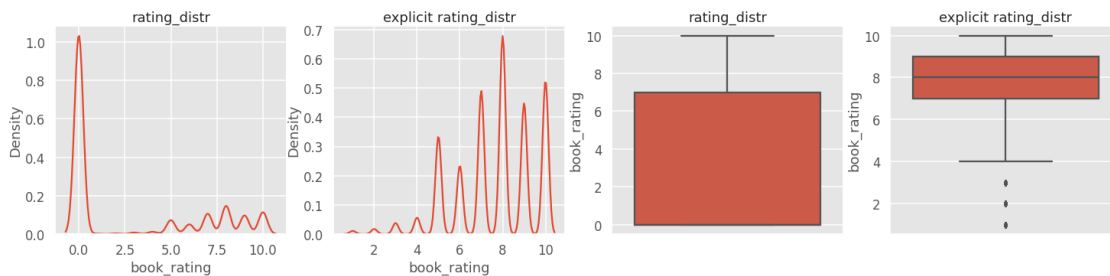
```

```
[35]: cleaned_data = dataset.copy()
```

```
[53]: # dataset = cleaned_data.copy()
```

1.2 Analyzing the Feature Space

```
[55]: f, axes = plt.subplots(ncols = 4, figsize=(25, 5))
sns.kdeplot(x="book_rating", data=dataset, ax=axes[0]).set_title('rating_distr')
sns.kdeplot(x="book_rating", data=dataset[dataset['book_rating'] != 0],
            ax=axes[1]).set_title('explicit rating_distr')
sns.boxplot(y="book_rating", data=dataset, orient='v', ax=axes[2]).
            set_title('rating_distr')
sns.boxplot(y="book_rating", data=dataset[dataset['book_rating'] != 0],
            orient='v', ax=axes[3]).set_title('explicit rating_distr')
plt.show()
```



We can remove 0 ratings, since these are unrated, and why would someone rate a book as 0 ?

```
[56]: dataset = dataset[dataset['book_rating'] != 0]
```

```
[57]: def plot_univariate(dataset, column_name, subtitle = None, kde_only = True):
    f, axes = plt.subplots(ncols = 4, figsize=(25, 5))
    if kde_only:
        sns.kdeplot(x=column_name, data=dataset, ax=axes[0]).
            set_title(f'{column_name}_distr')
    else:
        sns.histplot(x=column_name, data=dataset, kde=True, ax=axes[0]).
            set_title(f'{column_name}_distr')
        sns.boxplot(y=column_name, data=dataset, orient='v', ax=axes[1]).
            set_title(f'{column_name}_box_plot')
        sns.violinplot(y=column_name, data=dataset, orient='v', ax=axes[2]).
            set_title(f'{column_name}_violin_plot')
```



```

    scipy.stats.probplot(dataset[column_name], dist="norm", plot=axes[3])
    if suptitle:
        plt.suptitle(suptitle)
        plt.subplots_adjust(top=0.80)
    plt.show()

```

```

[58]: plot_univariate(dataset=dataset, column_name='book_rating', suptitle='Book_
      ↳Rating')

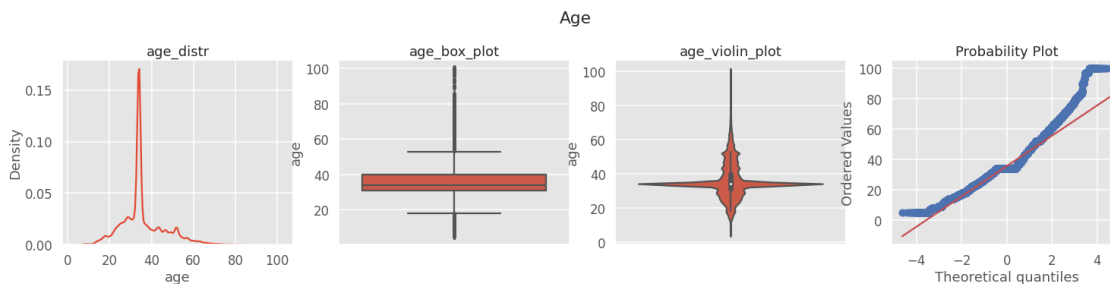
```



```

[59]: plot_univariate(dataset=dataset, column_name='age', suptitle='Age')

```



1.3 Data Transformation

1.3.1 Min-Max Normalization

$$x_{scaled} = \frac{x - x_{min}}{x_{max} - x_{min}}$$

```

[82]: scaler = MinMaxScaler()
      scaled = scaler.fit_transform(dataset['book_rating'].values.reshape(-1, 1)).
      ↳reshape(-1)

```

```

[83]: scaled_rating = pd.DataFrame(data=scaled, columns=['book_rating'])
      scaled_rating

```

```

[83]:    book_rating
      0      0.444444
      1      0.777778

```

```

2          0.777778
3          0.888889
4          0.888889
...
383844     0.666667
383845     0.444444
383846     0.666667
383847     0.666667
383848     1.000000

```

[383849 rows x 1 columns]

```
[84]: scaled_rating.describe().T
```

```

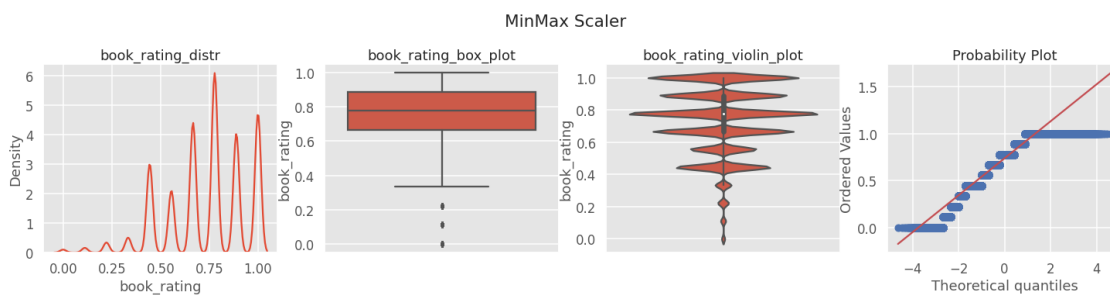
[84]:          count      mean      std  min      25%      50%      75%  max
book_rating 383849.0  0.7363  0.204593  0.0  0.666667  0.777778  0.888889  1.0

```

```

[85]: plot_univariate(dataset=scaled_rating, column_name='book_rating',
    ↳suptitle='MinMax Scaler')

```



1.3.2 Z-Score Standardization

$$Z = \frac{x - \mu}{\sigma}$$

```

[86]: scaler = StandardScaler()
scaled = scaler.fit_transform(dataset['book_rating'].values.reshape(-1, 1)).
    ↳reshape(-1)

```

```

[87]: scaled_rating = pd.DataFrame(data=scaled, columns=['book_rating'])
scaled_rating

```

```

[87]:    book_rating
0      -1.426522
1       0.202732
2       0.202732
3       0.745817
4       0.745817
...
383844 -0.340353

```

```
383845    -1.426522
383846    -0.340353
383847    -0.340353
383848     1.288902
```

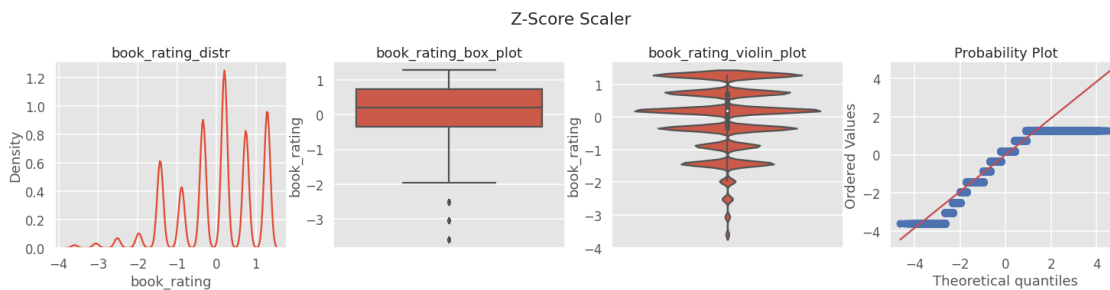
```
[383849 rows x 1 columns]
```

```
[88]: scaled_rating.describe().T
```

```
[88]:          count          mean         std  ...         50%         75%         max
book_rating 383849.0  2.953229e-14  1.000001  ...  0.202732  0.745817  1.288902
```

```
[1 rows x 8 columns]
```

```
[89]: plot_univariate(dataset=scaled_rating, column_name='book_rating',
    ↳ subtitle='Z-Score Scaler')
```



1.3.3 Decimal Scaling

$$v'_i = \frac{v_i}{10^q}$$

```
[90]: p = dataset['book_rating'].max()
      q = len(str(abs(p)))
      scaled = dataset['book_rating'].values / 10 ** q
```

```
[91]: scaled_rating = pd.DataFrame(data=scaled, columns=['book_rating'])
      scaled_rating
```

```
[91]:    book_rating
0         0.05
1         0.08
2         0.08
3         0.09
4         0.09
...         ...
383844     0.07
383845     0.05
383846     0.07
383847     0.07
```

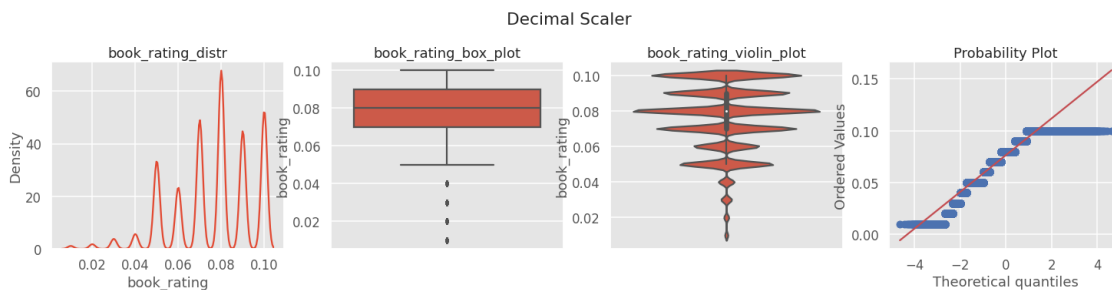
383848 0.10

[383849 rows x 1 columns]

```
[92]: scaled_rating.describe().T
```

```
[92]:          count      mean      std   min   25%   50%   75%   max
book_rating 383849.0  0.076267  0.018413  0.01  0.07  0.08  0.09  0.1
```

```
[93]: plot_univariate(dataset=scaled_rating, column_name='book_rating',
→suptitle='Decimal Scaler')
```



1.4 Data Normality

1.4.1 Natural Log Transform

```
[162]: transformed = np.log(dataset['book_rating'].astype(np.float32))
```

```
[163]: trans_rating = pd.DataFrame(data=transformed, columns=['book_rating'])
trans_rating
```

```
[163]:      book_rating
1      1.609438
3      2.079442
5      2.079442
8      2.197225
9      2.197225
...      ...
1031166  1.945910
1031168  1.609438
1031169  1.945910
1031170  1.945910
1031171  2.302585
```

[383849 rows x 1 columns]

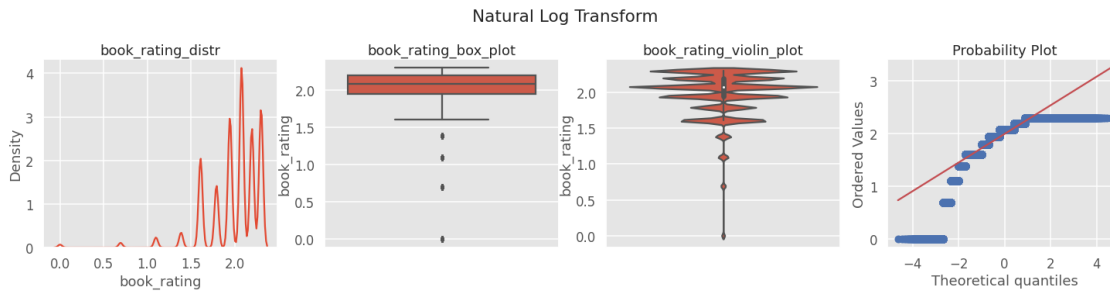
```
[164]: trans_rating.describe().T
```

```
[164]:
```

	count	mean	std	...	50%	75%	max
book_rating	383849.0	1.994574	0.301864	...	2.079442	2.197225	2.302585

[1 rows x 8 columns]

```
[165]: plot_univariate(dataset=trans_rating, column_name='book_rating',
→subtitle='Natural Log Transform')
```



1.4.2 Square Root Transform

```
[166]: transformed = np.sqrt(dataset['book_rating'].astype(np.float32))
```

```
[167]: trans_rating = pd.DataFrame(data=transformed, columns=['book_rating'])
trans_rating
```

```
[167]:
```

	book_rating
1	2.236068
3	2.828427
5	2.828427
8	3.000000
9	3.000000
...	...
1031166	2.645751
1031168	2.236068
1031169	2.645751
1031170	2.645751
1031171	3.162278

[383849 rows x 1 columns]

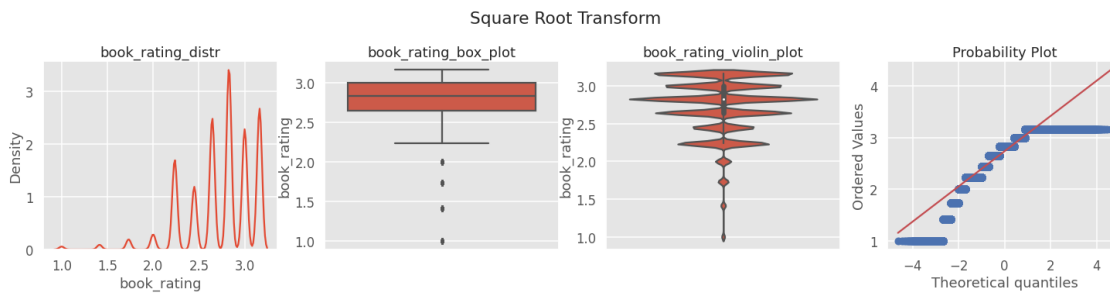
```
[168]: trans_rating.describe().T
```

```
[168]:
```

	count	mean	std	...	50%	75%	max
book_rating	383849.0	2.738279	0.362097	...	2.828427	3.0	3.162278

[1 rows x 8 columns]

```
[169]: plot_univariate(dataset=trans_rating, column_name='book_rating',
→suptitle='Square Root Transform')
```



1.4.3 Inverse Square Root Transformation

```
[170]: transformed = np.power(dataset['book_rating'].astype(np.float32), -1/2)
```

```
[171]: trans_rating = pd.DataFrame(data=transformed, columns=['book_rating'])
trans_rating
```

```
[171]:      book_rating
1      0.447214
3      0.353553
5      0.353553
8      0.333333
9      0.333333
...      ...
1031166  0.377964
1031168  0.447214
1031169  0.377964
1031170  0.377964
1031171  0.316228
```

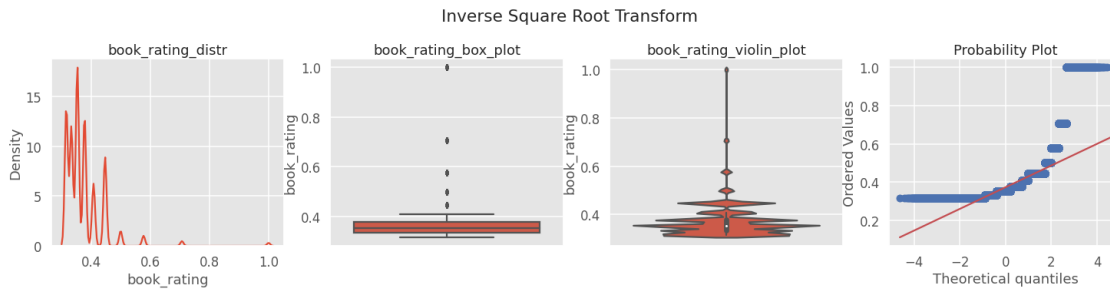
[383849 rows x 1 columns]

```
[172]: trans_rating.describe().T
```

```
[172]:      count      mean      std      ...      50%      75%      max
book_rating  383849.0  0.373803  0.06913      ...  0.353553  0.377964  1.0
```

[1 rows x 8 columns]

```
[173]: plot_univariate(dataset=trans_rating, column_name='book_rating',
→suptitle='Inverse Square Root Transform')
```



1.5 Exploratory Data Analysis

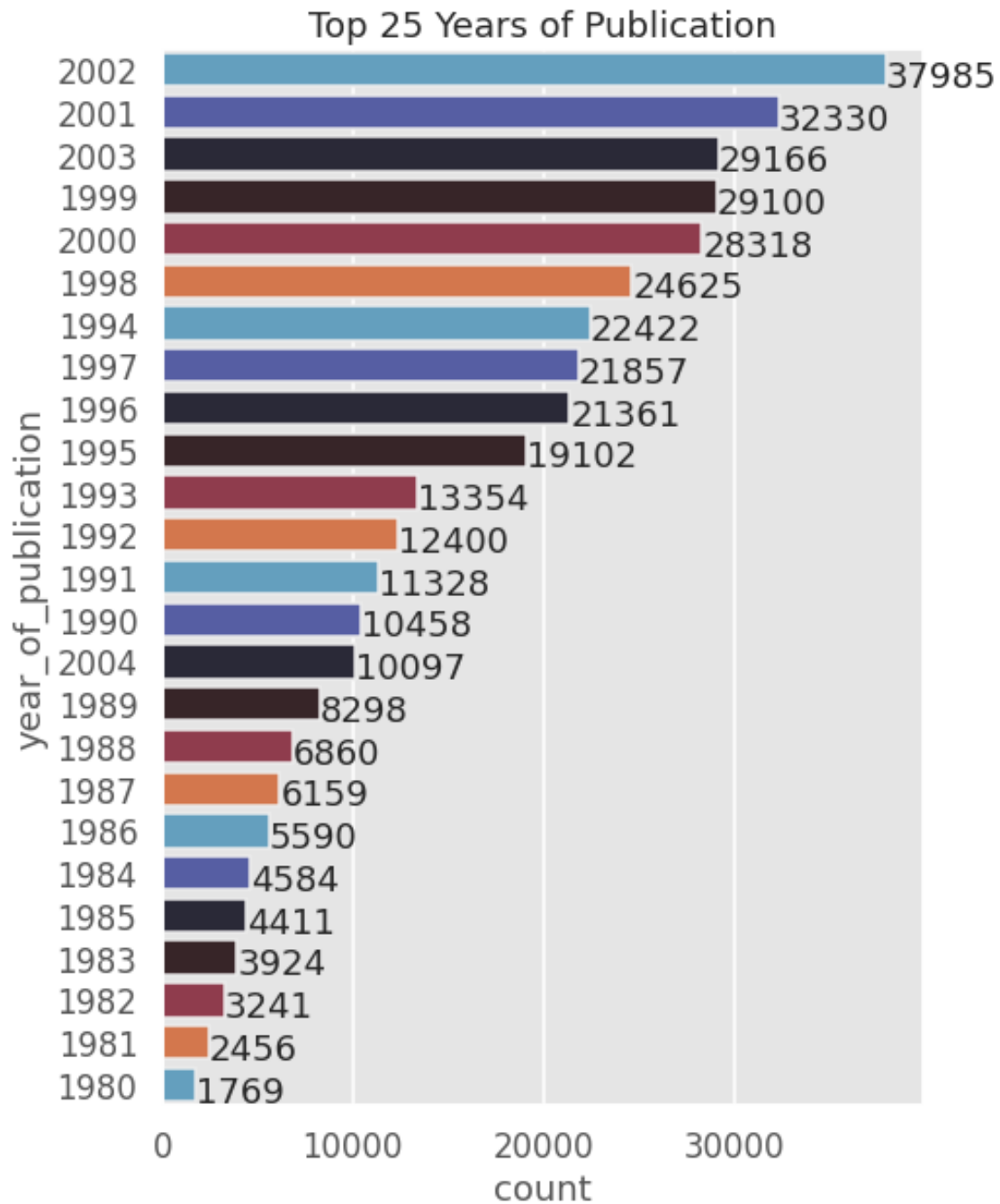
[186]: *# to plot values in barplot, <https://stackoverflow.com/a/56780852>*

```
def show_values_on_bars(axes, h_v="v", space=0.4):
    def _show_on_single_plot(ax):
        if h_v == "v":
            for p in ax.patches:
                _x = p.get_x() + p.get_width() / 2
                _y = p.get_y() + p.get_height()
                value = int(p.get_height())
                ax.text(_x, _y, value, ha="center")
        elif h_v == "h":
            for p in ax.patches:
                _x = p.get_x() + p.get_width() + float(space)
                _y = p.get_y() + p.get_height()
                value = int(p.get_width())
                ax.text(_x, _y, value, ha="left")

    if isinstance(axes, np.ndarray):
        for idx, ax in np.ndenumerate(axes):
            _show_on_single_plot(ax)
    else:
        _show_on_single_plot(axes)
```

```
[256]: eda = dataset['year_of_publication'].copy().value_counts().head(25).
        ↪reset_index()
eda.columns = ['year_of_publication', 'count']
eda = eda.sort_values(by=['count'], ascending=False)
```

```
[257]: plt.figure(figsize=(7, 10))
splot = sns.barplot(x='count', y='year_of_publication', data=eda,
        ↪order=eda['year_of_publication'], orient='h', palette=palette)
show_values_on_bars(splot, h_v="h")
plt.title('Top 25 Years of Publication')
plt.show()
```

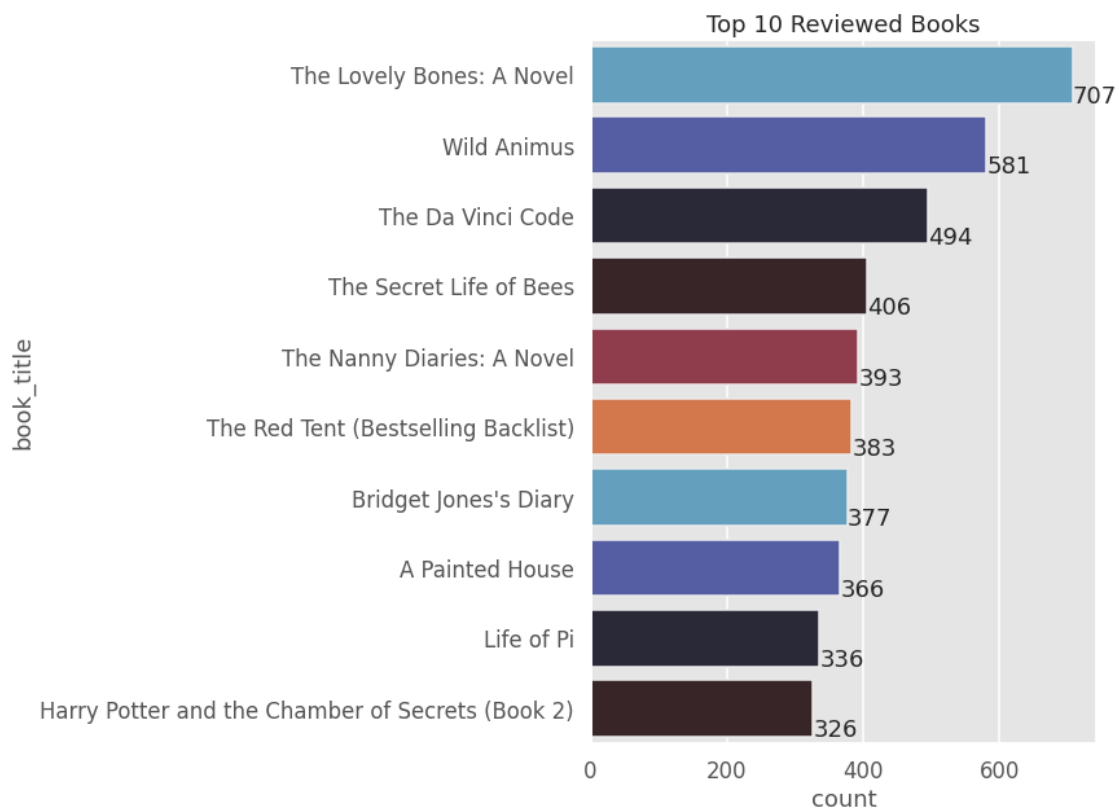


```
[258]: eda = dataset['book_title'].value_counts().head(10).reset_index()
eda.columns = ['book_title', 'count']
```

```
[259]: plt.figure(figsize=(7, 10))
splot = sns.barplot(x='count', y='book_title', data=eda,
                    order=eda['book_title'], orient='h', palette=palette)
show_values_on_bars(splot, h_v="h")
```



```
plt.title('Top 10 Reviewed Books')
plt.show()
```



```
[280]: rating_count = dataset['book_title'].value_counts().head(25).reset_index().
        ↳sort_values(by='book_title').reset_index(drop=True)
rating_count.columns = ['book_title', 'rating_count']
rating_count.head(5)
```

```
[280]:
```

	book_title	rating_count
0	The Girls' Guide to Hunting and Fishing	259
1	The Testament	261
2	Timeline	263
3	The Catcher in the Rye	265
4	To Kill a Mockingbird	267

```
[281]: rating_sum = dataset[dataset['book_title'].isin(rating_count['book_title'])].
        ↳groupby(['book_title'])['book_rating'].sum().reset_index().
        ↳sort_values(by=['book_title'])
rating_sum.columns = ['book_title', 'rating_sum']
rating_sum.head(5)
```

```
[281]:
```

	book_title	rating_sum
0	A Painted House	2708.0

1	Angels & Demons	2485.0
2	Bridget Jones's Diary	2875.0
3	Divine Secrets of the Ya-Ya Sisterhood: A Novel	2544.0
4	Girl with a Pearl Earring	2219.0

```
[282]: avg_rating = pd.merge(rating_count, rating_sum, on='book_title')
avg_rating['rating_avg'] = avg_rating['rating_sum'] / avg_rating['rating_count']
avg_rating = avg_rating.sort_values(by='rating_avg', ascending=False).
      ↪reset_index(drop=True)
```

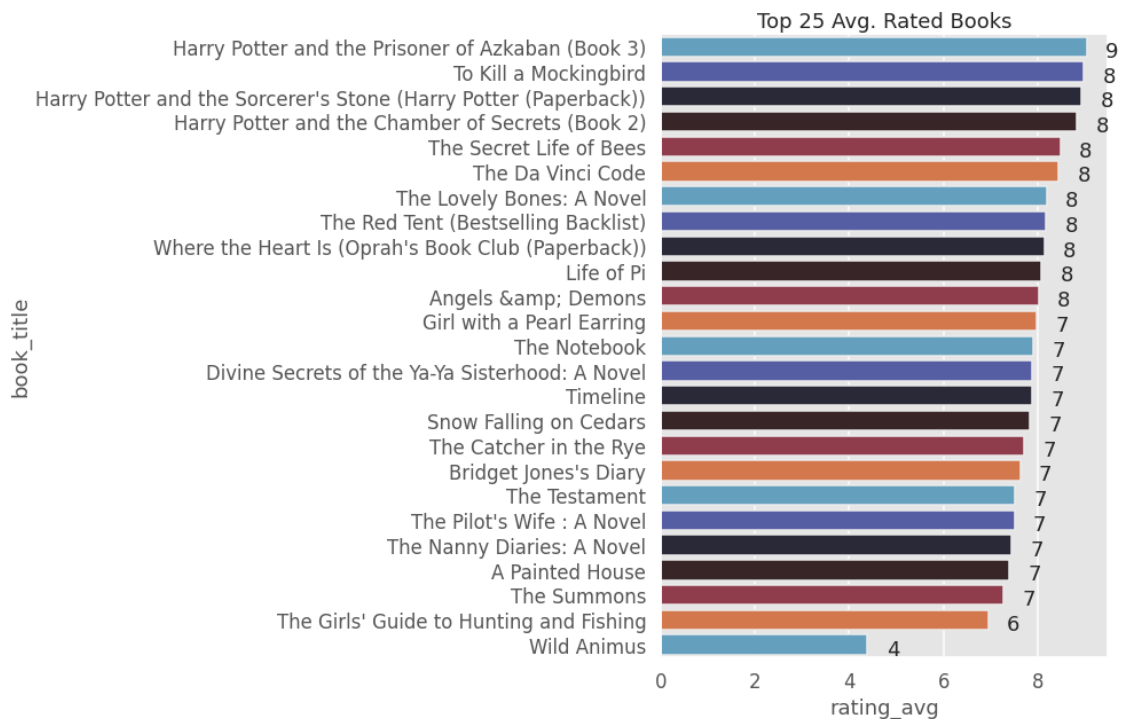
```
[283]: avg_rating.head(5)
```

```
[283]:
```

	book_title	...	rating_avg
0	Harry Potter and the Prisoner of Azkaban (Book 3)	...	9.043321
1	To Kill a Mockingbird	...	8.977528
2	Harry Potter and the Sorcerer's Stone (Harry P...	...	8.936508
3	Harry Potter and the Chamber of Secrets (Book 2)	...	8.840491
4	The Secret Life of Bees	...	8.477833

[5 rows x 4 columns]

```
[284]: plt.figure(figsize=(7, 10))
splot = sns.barplot(x='rating_avg', y='book_title', data=avg_rating,
      ↪order=avg_rating['book_title'], orient='h', palette=palette)
show_values_on_bars(splot, h_v="h")
plt.title('Top 25 Avg. Rated Books')
plt.show()
```

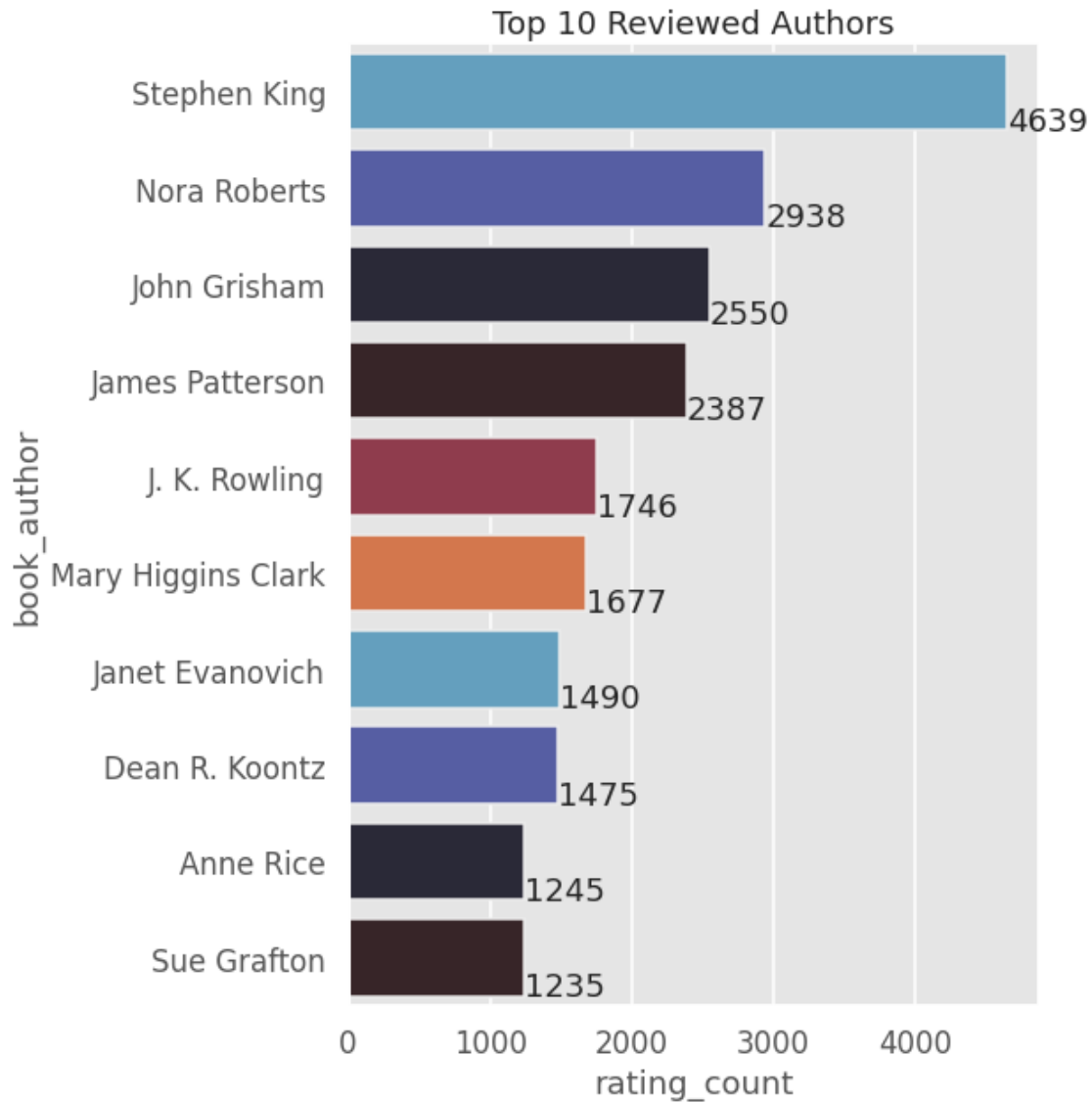


```
[285]: author_count = dataset['book_author'].value_counts().head(10).reset_index()
author_count.columns = ['book_author', 'rating_count']
author_count.head(5)
```

```
[285]:
```

	book_author	rating_count
0	Stephen King	4639
1	Nora Roberts	2938
2	John Grisham	2550
3	James Patterson	2387
4	J. K. Rowling	1746

```
[286]: plt.figure(figsize=(7, 10))
splot = sns.barplot(x='rating_count', y='book_author', data=author_count,
→order=author_count['book_author'], orient='h', palette=palette)
show_values_on_bars(splot, h_v="h")
plt.title('Top 10 Reviewed Authors')
plt.show()
```

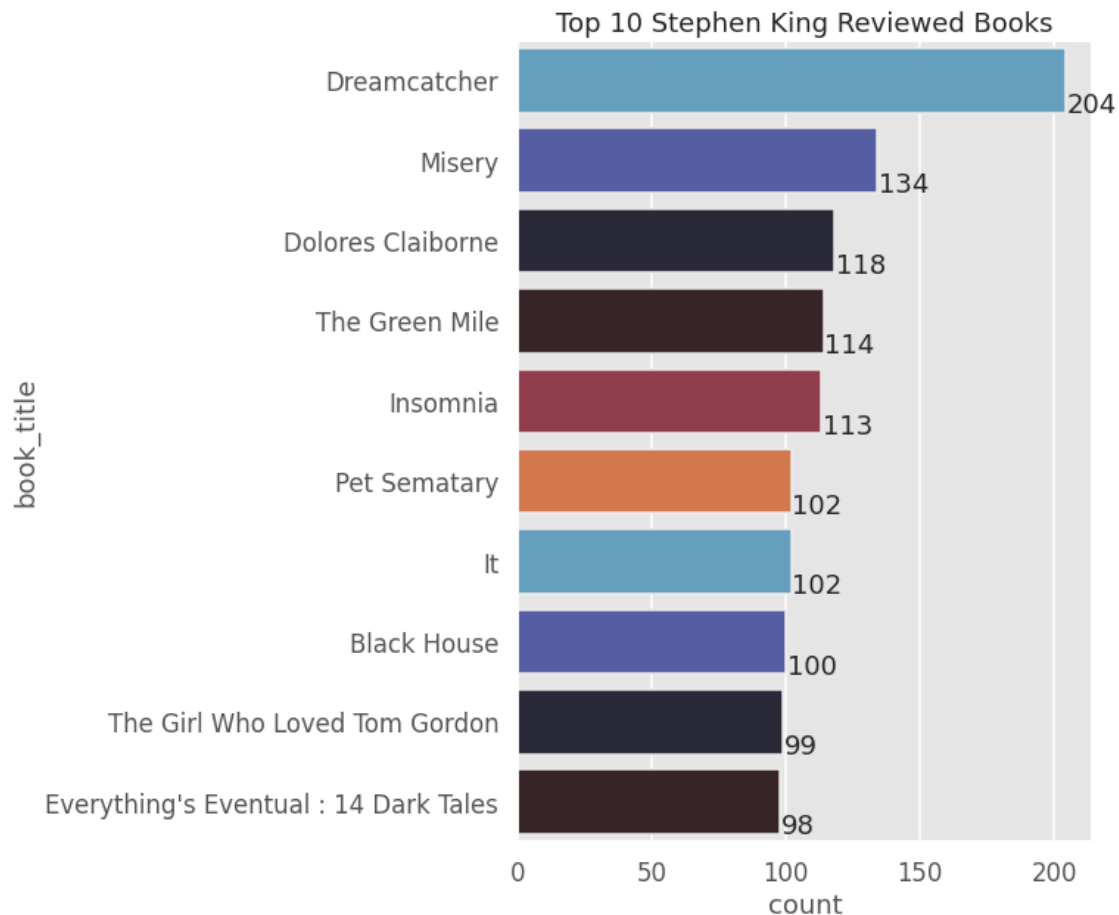


1.5.1 It seems Stephen King is a very popular author !

```
[288]: stephen_king = dataset[dataset['book_author'] == 'Stephen King']
```

```
[289]: eda = stephen_king['book_title'].value_counts().head(10).reset_index()
eda.columns = ['book_title', 'count']
```

```
[290]: plt.figure(figsize=(7, 10))
splot = sns.barplot(x='count', y='book_title', data=eda,
                    order=eda['book_title'], orient='h', palette=palette)
show_values_on_bars(splot, h_v="h")
plt.title('Top 10 Stephen King Reviewed Books')
plt.show()
```



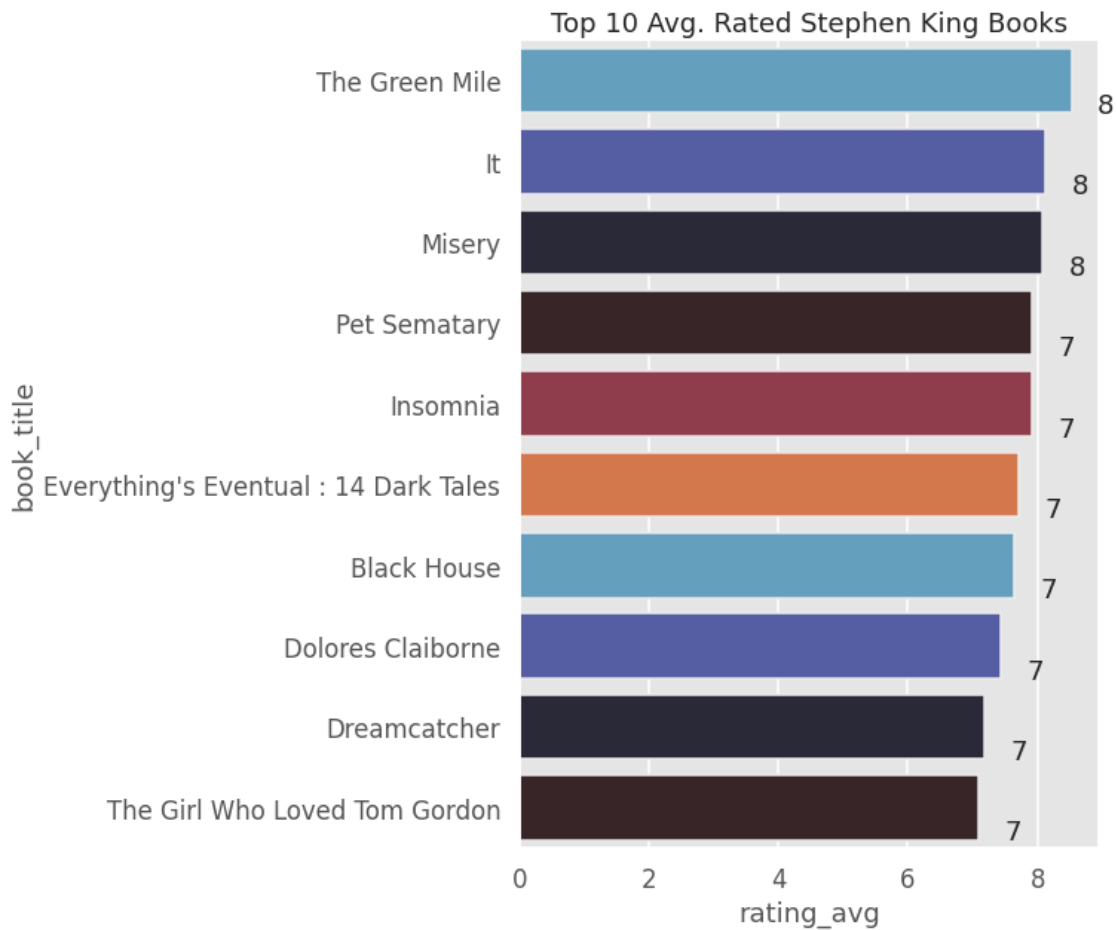
```
[291]: rating_count = stephen_king['book_title'].value_counts().head(10).reset_index().
        ↳sort_values(by='book_title').reset_index(drop=True)
rating_count.columns = ['book_title', 'rating_count']
rating_sum = stephen_king[stephen_king['book_title'].
        ↳isin(rating_count['book_title'])].groupby(['book_title'])['book_rating'].
        ↳sum().reset_index().sort_values(by=['book_title'])
rating_sum.columns = ['book_title', 'rating_sum']
avg_rating = pd.merge(rating_count, rating_sum, on='book_title')
avg_rating['rating_avg'] = avg_rating['rating_sum'] / avg_rating['rating_count']
avg_rating = avg_rating.sort_values(by='rating_avg', ascending=False).
        ↳reset_index(drop=True)
```

```
[292]: avg_rating.head(5)
```

```
[292]:
```

	book_title	rating_count	rating_sum	rating_avg
0	The Green Mile	114	972.0	8.526316
1	It	102	829.0	8.127451
2	Misery	134	1082.0	8.074627
3	Pet Sematary	102	808.0	7.921569
4	Insomnia	113	895.0	7.920354

```
[293]: plt.figure(figsize=(7, 10))
sns.barplot(x='rating_avg', y='book_title', data=avg_rating,
            order=avg_rating['book_title'], orient='h', palette=palette)
show_values_on_bars(splot, h_v="h")
plt.title('Top 10 Avg. Rated Stephen King Books')
plt.show()
```



1.6 Convert this Notebook to PDF

```
[181]: %%capture
! apt update
! apt install texlive-xetex texlive-fonts-recommended
      ↳ texlive-generic-recommended
```

```
[182]: import subprocess
import shlex
```

Convert to PDF

```
[188]: s = subprocess.Popen(shlex.split(  
      f'jupyter nbconvert /content/BookReview_EDA.ipynb --to pdf'  
      ), shell = False, stdout = subprocess.PIPE, stderr = subprocess.PIPE)  
s.wait()  
s.stdout.read()
```

```
[188]: b''
```

Convert to L^AT_EX

```
[189]: s = subprocess.Popen(shlex.split(  
      f'jupyter nbconvert /content/BookReview_EDA.ipynb --to latex'  
      ), shell = False, stdout = subprocess.PIPE, stderr = subprocess.PIPE)  
s.wait()  
s.stdout.read()
```

```
[189]: b''
```

```
[190]: ! zip -r BookReview_EDA_latex.zip BookReview_EDA_files BookReview_EDA.tex
```

```
updating: BookReview_EDA_files/ (stored 0%)  
updating: BookReview_EDA_files/BookReview_EDA_53_0.png (deflated 7%)  
updating: BookReview_EDA_files/BookReview_EDA_99_0.png (deflated 12%)  
updating: BookReview_EDA_files/BookReview_EDA_105_0.png (deflated 13%)  
updating: BookReview_EDA_files/BookReview_EDA_83_0.png (deflated 6%)  
updating: BookReview_EDA_files/BookReview_EDA_101_0.png (deflated 13%)  
updating: BookReview_EDA_files/BookReview_EDA_78_0.png (deflated 6%)  
updating: BookReview_EDA_files/BookReview_EDA_92_0.png (deflated 13%)  
updating: BookReview_EDA_files/BookReview_EDA_94_0.png (deflated 14%)  
updating: BookReview_EDA_files/BookReview_EDA_108_0.png (deflated 14%)  
updating: BookReview_EDA_files/BookReview_EDA_52_0.png (deflated 5%)  
updating: BookReview_EDA_files/BookReview_EDA_66_0.png (deflated 5%)  
updating: BookReview_EDA_files/BookReview_EDA_48_0.png (deflated 6%)  
updating: BookReview_EDA_files/BookReview_EDA_88_0.png (deflated 8%)  
updating: BookReview_EDA_files/BookReview_EDA_60_0.png (deflated 5%)  
updating: BookReview_EDA_files/BookReview_EDA_72_0.png (deflated 5%)  
updating: BookReview_EDA.tex (deflated 88%)
```

```
[ ]:
```