

Book Store Data Analysis

This project analyzes a book store dataset to explore sales performance, popular genres, publishers, and the relationship between ratings and sales.

Dataset: Book Store

Objective: Identify top-performing publishers, explore sales patterns, and analyze the effect of language, genre, and ratings on book sales.

```
In [3]: # Import the Main Library
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

```
In [4]: # Import the dataset and check it Structure
df=pd.read_csv('D:/Books_Data.csv')
df.head()
```

Out[4]:

	index	Publishing Year	Book Name	Author	language_code	Author_Rating	Book_ave
0	0	1975.0	Beowulf	Unknown, Seamus Heaney	en-US	Novice	
1	1	1987.0	Batman: Year One	Frank Miller, David Mazzucchelli, Richmond Lew...	eng	Intermediate	
2	2	2015.0	Go Set a Watchman	Harper Lee	eng	Novice	
3	3	2008.0	When You Are Engulfed in Flames	David Sedaris	en-US	Intermediate	
4	4	2011.0	Daughter of Smoke & Bone	Laini Taylor	eng	Intermediate	



```
In [5]: # Check the Shape of the Data
print(f'The Number of Rows is: {df.shape[0]}')
print(f'The Number of Columns is: {df.shape[1]}')
```

The Number of Rows is: 1070
The Number of Columns is: 15

```
In [6]: # Check the Columns Name
```

```
df.columns
```

```
Out[6]: Index(['index', 'Publishing Year', 'Book Name', 'Author', 'language_code',  
             'Author_Rating', 'Book_average_rating', 'Book_ratings_count', 'genre',  
             'gross sales', 'publisher revenue', 'sale price', 'sales rank',  
             'Publisher ', 'units sold'],  
            dtype='object')
```

```
In [7]: # Check the null values  
df.isnull().sum()
```

```
Out[7]: index                0  
Publishing Year            1  
Book Name                 23  
Author                    0  
language_code             53  
Author_Rating             0  
Book_average_rating       0  
Book_ratings_count        0  
genre                     0  
gross sales               0  
publisher revenue         0  
sale price                0  
sales rank                0  
Publisher                 0  
units sold                0  
dtype: int64
```

```
In [8]: # Handling the Missing Value  
df.dropna(subset='Book Name',inplace=True)  
df.fillna({'language_code': 'Unknown'},inplace=True)
```

```
In [9]: # Check the Duplicate Rows  
print(df.duplicated().sum())
```

0


```
In [10]: #After Handle the Missing Value  
df.isnull().sum()
```

```
Out[10]: index                0  
Publishing Year            0  
Book Name                 0  
Author                    0  
language_code             0  
Author_Rating             0  
Book_average_rating       0  
Book_ratings_count        0  
genre                     0  
gross sales               0  
publisher revenue         0  
sale price                0  
sales rank                0  
Publisher                 0  
units sold                0  
dtype: int64
```

```
In [11]: # Check the Descriptive Statistics of Dataset  
df.describe()
```

Out[11]:

	index	Publishing Year	Book_average_rating	Book_ratings_count	gross sales
count	1047.000000	1047.000000	1047.000000	1047.000000	1047.000000
mean	533.494747	1971.115568	4.00255	94987.737345	1847.212875
std	308.421094	186.735278	0.24679	31449.738538	3940.055642
min	0.000000	-560.000000	2.97000	27308.000000	104.940000
25%	267.500000	1984.500000	3.85000	70414.000000	357.465000
50%	534.000000	2002.000000	4.01000	89410.000000	792.000000
75%	796.500000	2009.000000	4.17000	113946.000000	1487.645000
max	1069.000000	2016.000000	4.77000	206792.000000	47795.000000



```
In [12]: # Converted 'Publishing Year' to integer and handled invalid values by filtering
df=df[df['Publishing Year']>1900]
df['Publishing Year']=df['Publishing Year'].astype('int64')
```

```
In [14]: # Check the Data type of all Columns
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Index: 988 entries, 0 to 1069
Data columns (total 15 columns):
#   Column                Non-Null Count  Dtype
---  -
0   index                  988 non-null   int64
1   Publishing Year        988 non-null   int64
2   Book Name              988 non-null   object
3   Author                 988 non-null   object
4   language_code          988 non-null   object
5   Author_Rating          988 non-null   object
6   Book_average_rating     988 non-null   float64
7   Book_ratings_count      988 non-null   int64
8   genre                  988 non-null   object
9   gross sales            988 non-null   float64
10  publisher revenue       988 non-null   float64
11  sale price              988 non-null   float64
12  sales rank              988 non-null   int64
13  Publisher               988 non-null   object
14  units sold              988 non-null   int64
dtypes: float64(4), int64(5), object(6)
memory usage: 123.5+ KB
```

```
In [15]: # Check the number of unique value in all columns
df.nunique()
```

```
Out[15]: index          988
Publishing Year      101
Book Name           987
Author              669
language_code         9
Author_Rating         4
Book_average_rating  133
Book_ratings_count   983
genre                4
gross sales          774
publisher revenue    570
sale price           143
sales rank           818
Publisher             9
units sold           470
dtype: int64
```

Top Selling Year

```
In [16]: df.groupby('Publishing Year')['gross sales'].sum().sort_values(ascending=False).
```

```
Out[16]: Publishing Year
2011      127763.77
2008      107665.37
2012       92435.27
2015       86692.83
2006       84589.66
Name: gross sales, dtype: float64
```

Top Most Expensive Books

```
In [17]: df.groupby('Book Name')['sale price'].sum().sort_values(ascending=False).head()
```

```
Out[17]: Book Name
The Wind in the Willows      33.86
We Need to Talk About Kevin  25.89
Sabriel                     19.98
The Tenth Circle             17.98
Alanna: The First Adventure  16.98
Name: sale price, dtype: float64
```

Top Sales Rank Books

```
In [18]: print(df[['Book Name', 'sales rank']].sort_values(by='sales rank').head().to_stri
```

```

          Book Name  sales rank
1          Beowulf            1
2  Batman: Year One            2
3  When You Are Engulfed in Flames  3
3          Go Set a Watchman       3
4  Daughter of Smoke & Bone       4
```

Top Rating Books

```
In [19]: df[['Book Name', 'Book_average_rating', 'Book_ratings_count']].sort_values(by='Book_ratings_count')
```

```
Out[19]:
```

	Book Name	Book_average_rating	Book_ratings_count
0	Words of Radiance	4.77	73572
1	A Court of Mist and Fury	4.72	108384
2	The Essential Calvin and Hobbes: A Calvin and ...	4.65	93001
3	The Way of Kings	4.64	144822
4	Calvin and Hobbes	4.61	117788

Top Authors By Sales

```
In [20]: df.groupby('Author', as_index=False)[['units sold', 'gross sales']].sum().sort_val
```

```
Out[20]:
```

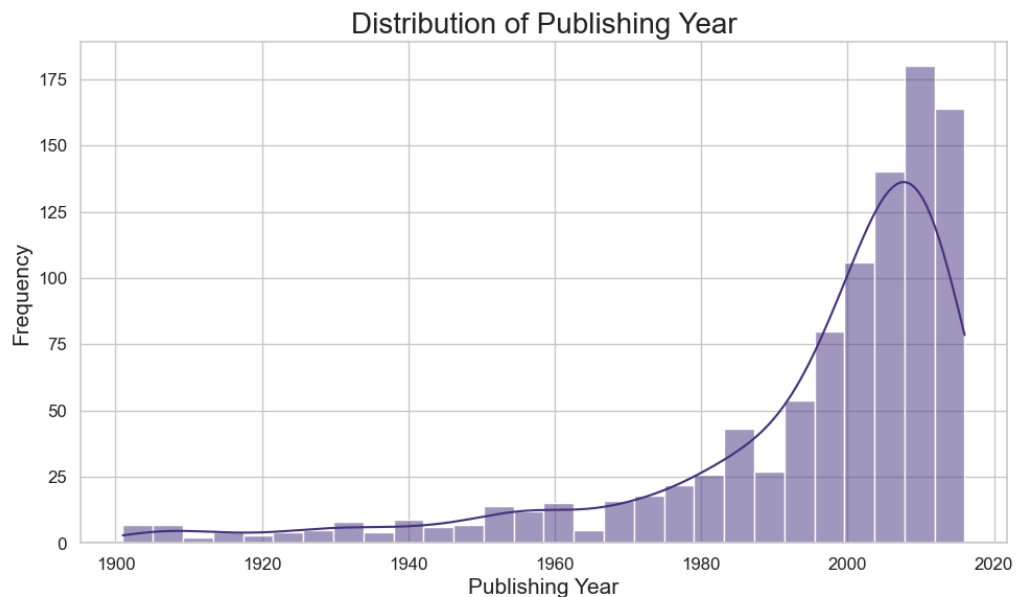
	Author	units sold	gross sales
0	Harper Lee	5500	47795.00
1	Stephen King	278322	43322.65
2	David Sedaris	15193	42323.41
3	Charlaine Harris	47249	39453.08
4	Laini Taylor	8990	38278.41

```
In [26]: # Plot Theme Setting
sns.set_theme(
    style="whitegrid",
    palette="viridis",
    font_scale=1.1
)
plt.rcParams['figure.figsize'] = (10, 6)
```

Distribution of Books Published Each Year Since 1900

```
In [27]: # Histogram

sns.histplot(x='Publishing Year', data=df, kde=True)
plt.title('Distribution of Publishing Year', fontsize=20)
plt.xlabel('Publishing Year', fontsize=15)
plt.ylabel('Frequency', fontsize=15)
plt.figtext(0.5, -0.05, 'Insight: Most of the books were published after the year 2000', ha='center')
plt.tight_layout()
plt.show()
```



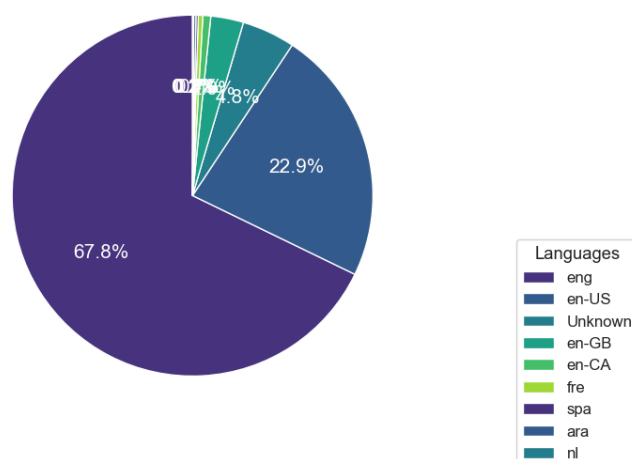
Insight: Most of the books were published after the year 2000. Before that, especially in the early 1900s, very few books were published.

Distribution of Books by Language

```
In [28]: # Count of books by language
lang_counts= df['language_code'].value_counts()

# Pie chart
plt.pie(lang_counts, labels=None, autopct='%1.1f%%', startangle=90, textprops={'fontcolor': 'white'})
plt.title('Percentage of Books by Language', fontsize=20)
plt.figtext(0.5, -0.1, "Insight: Most books are published in English (around 68%")
plt.legend(lang_counts.index, title="Languages", loc="center left", bbox_to_anchor=(1, 0))
plt.tight_layout()
plt.show()
```

Percentage of Books by Language



Insight: Most books are published in English (around 68% as 'eng' and 23% as 'en-US'), while approximately 5% of the records had missing language information, which has been labeled as 'Unknown'.

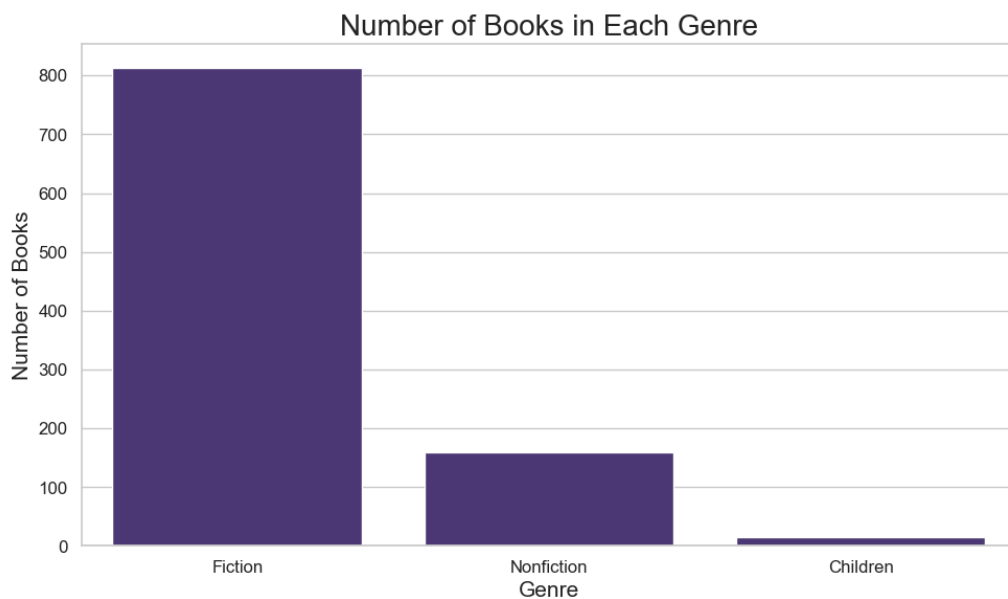
```
In [29]: # Check the Genre of Books
df['genre'].unique()
```

```
Out[29]: array(['genre fiction', 'fiction', 'nonfiction', 'children'], dtype=object)
```

```
In [30]: # Change 'Genre fiction' into 'Fiction' Because both are same
df['genre']=df['genre'].replace({'genre fiction':'fiction'})
df['genre']=df['genre'].str.title()
```

Number of Books in Each Genre

```
In [31]: #Plot
sns.countplot(x='genre',data=df)
plt.title('Number of Books in Each Genre',fontsize=20)
plt.xlabel('Genre',fontsize=15)
plt.xticks(rotation=0)
plt.ylabel('Number of Books',fontsize=15)
plt.figtext(0.5, -0.05, "Insight: Most of the published books belong to the Fict")
plt.tight_layout()
plt.show()
```



Insight: Most of the published books belong to the Fiction category (over 80%), followed by Nonfiction (about 16%), while Children's books make up only a small fraction of the total.

Books Published by Each Publisher

```
In [32]: # Sort the data so the highest values appear at the top.
sort=df['Publisher '].value_counts().index

# Plot
sns.countplot(y='Publisher ',data=df,order=sort)
plt.ylabel("Publisher",fontsize=15)
plt.xlabel("Number of Books",fontsize=15)
plt.title("Number of Books Published by Each Publisher",fontsize=20)
plt.figtext(0.5, -0.1, '''Insight: Amazon Digital Services, Inc. leads in book p

''', ha='center', fontsize=12)
plt.show()
```



Insight: Amazon Digital Services, Inc. leads in book publications, followed by Random House LLC and Penguin Group (USA) LLC, while publishers like HarperCollins have far fewer titles.

Number of Books Across Different Author Ratings

```
In [33]: # Sort the data so the highest values appear at the top.
sort_rating=df['Author_Rating'].value_counts().index

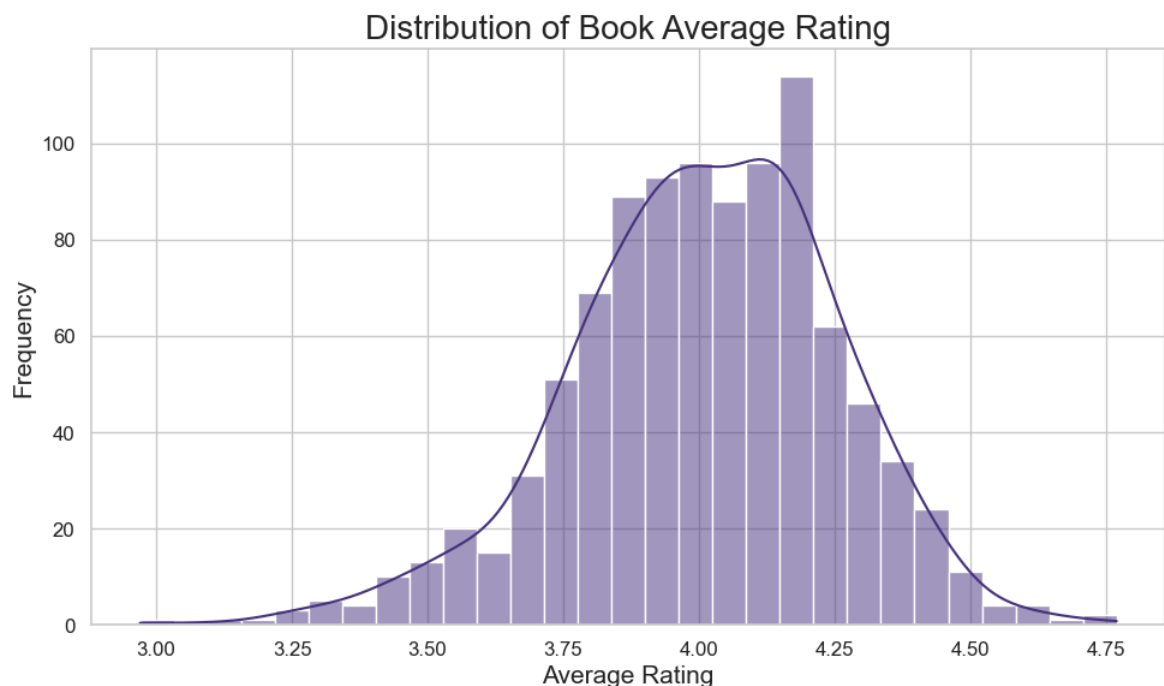
# Plot
sns.countplot(x='Author_Rating',data=df,order=sort_rating)
plt.title('Number of Books by Author Rating',fontsize=20)
plt.xlabel('Author Rating',fontsize=15)
plt.ylabel('Number of Books Published',fontsize=15)
plt.figtext(0.5, -0.05, "Insight: Most books are written by Intermediate-rated a
plt.tight_layout()
plt.show()
```



Insight: Most books are written by Intermediate-rated authors, followed by Excellent-rated ones, while Famous and Novice authors contribute very few titles.

Distribution of Books Based on Average Ratings

```
In [34]: # Histogram
sns.histplot(x='Book_average_rating', data=df, kde=True)
plt.title('Distribution of Book Average Rating', fontsize=20)
plt.xlabel('Average Rating', fontsize=15)
plt.ylabel('Frequency', fontsize=15)
plt.figtext(0.5, -0.05, 'Insight: Most books have an average rating between 3.8',
            ha='center', fontsize=10)
plt.tight_layout()
plt.show()
```



Insight: Most books have an average rating between 3.8 and 4.2, indicating generally positive reviews, with very few books rated below 3.5 or above 4.5.

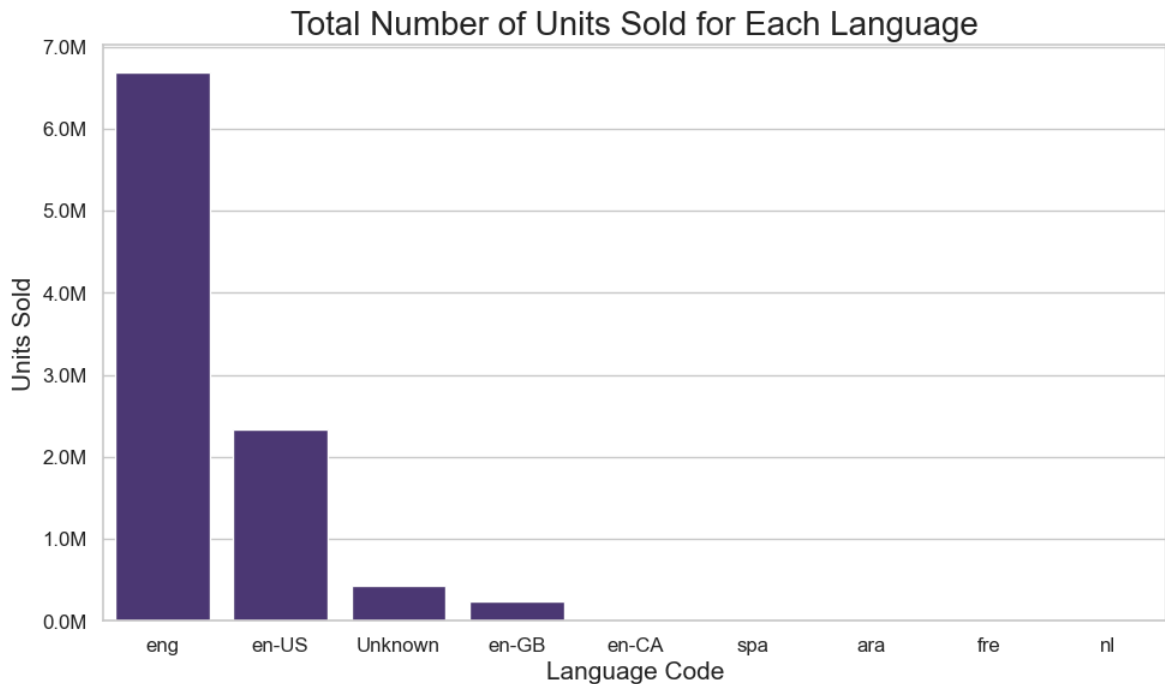
Total Number of Units Sold for Each Language

```
In [35]: # Create a Function to Change a Axis Values in Million
from matplotlib.ticker import FuncFormatter
def millions(x, pos):
    return f'{x/1e6:.1f}M'

# Group & sort the Data
sales_data=df.groupby('language_code', as_index=False)['units sold'].sum().sort_v

# Plot
plot = sns.barplot(x='language_code', y='units sold', data=sales_data, estimator=s
plt.title('Total Number of Units Sold for Each Language', fontsize=20)
plt.xlabel('Language Code', fontsize=15)
plt.ylabel('Units Sold', fontsize=15)
plot.yaxis.set_major_formatter(FuncFormatter(millions))
plt.figtext(0.5, -0.05, 'Insight: Most books were sold in English (especially en
```

```
, ha='center', fontsize=10)
plt.tight_layout()
plt.show()
```

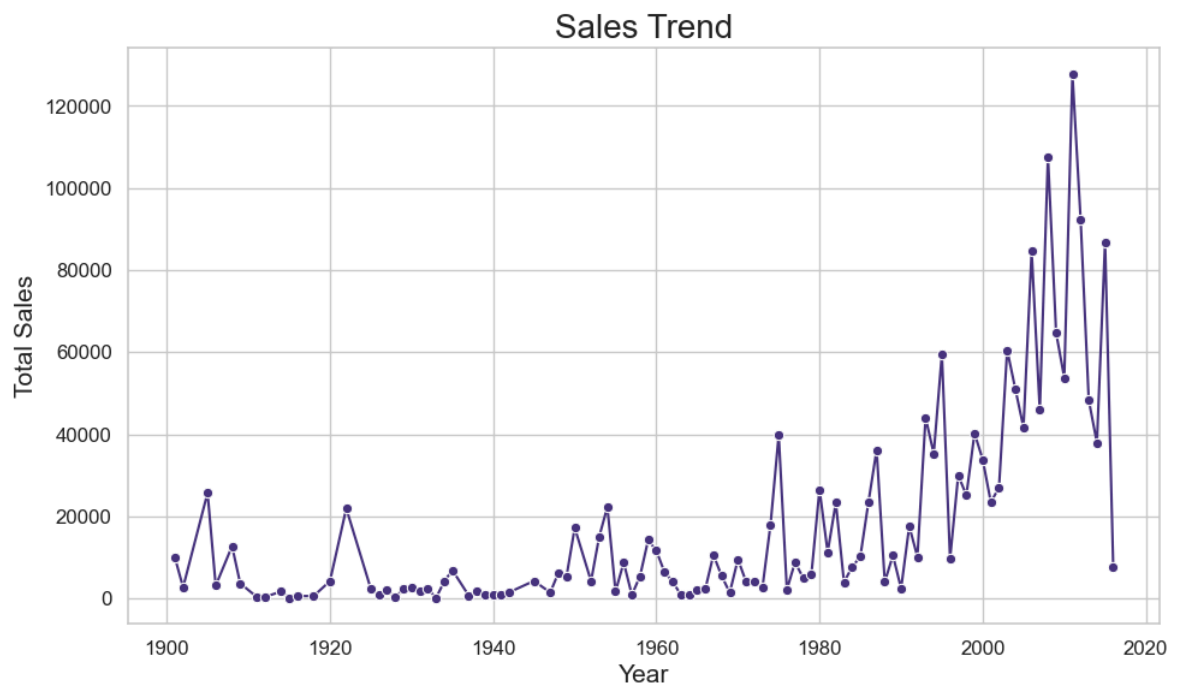


Insight: Most books were sold in English (especially eng and en-US), while sales in other languages were very low.

Sales Trend

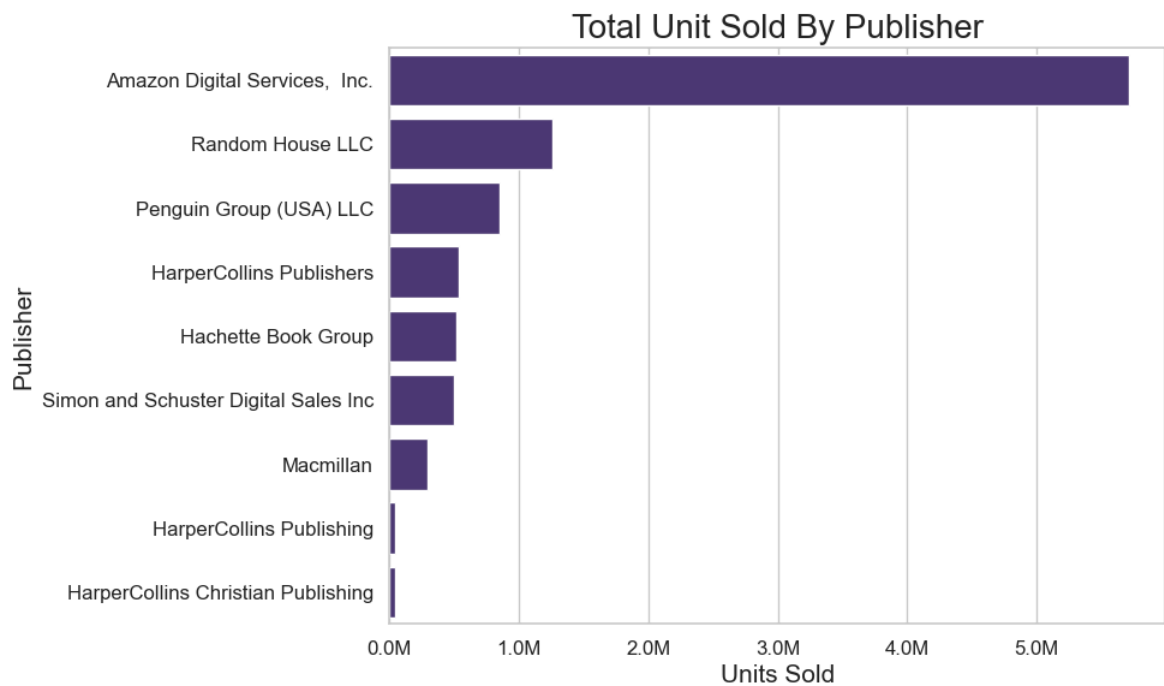
```
In [36]: # Group the data to Year wise
yearly_sales = df.groupby('Publishing Year')['gross sales'].sum().reset_index()

# PLOT
sns.lineplot(x='Publishing Year',y='gross sales',data=yearly_sales,marker='o')
plt.xlabel('Year',fontsize=15)
plt.ylabel('Total Sales',fontsize=15)
plt.title('Sales Trend',fontsize=20)
plt.figtext(0.5, -0.05, 'Insight: Sales remained steady until the 1990s, then gr
, ha='center', fontsize=10)
plt.tight_layout()
plt.show()
```



Total Units Sold By Publisher

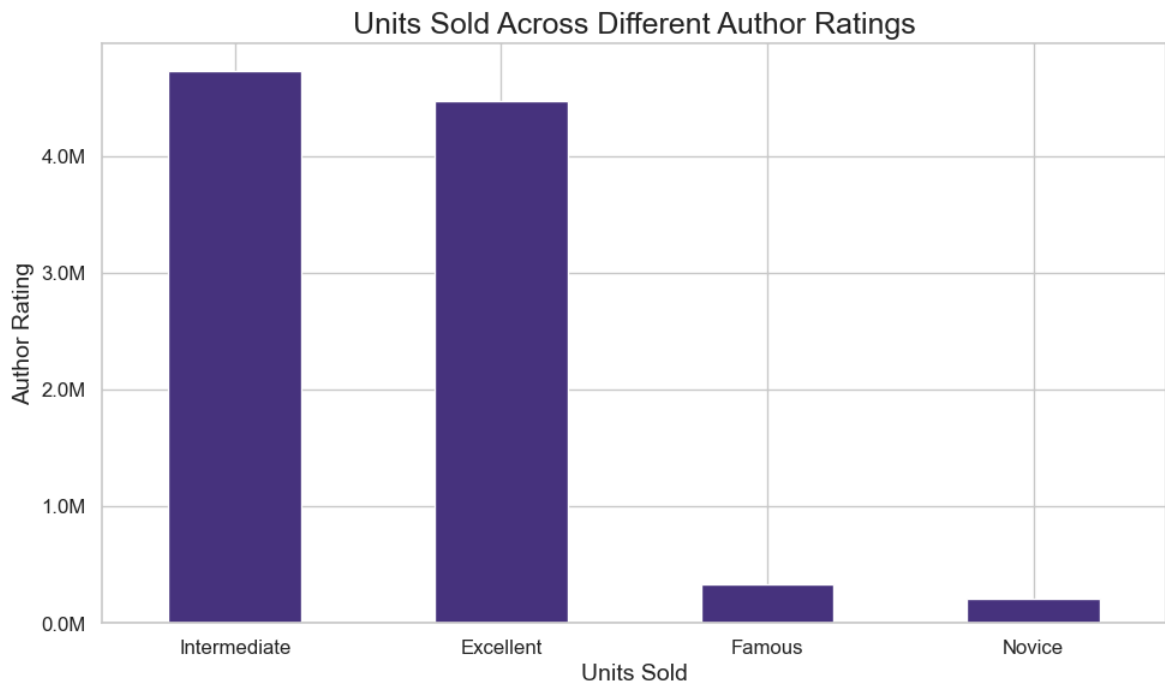
```
In [37]: # Group & sort the Data
sort_by_publisher=df.groupby('Publisher ',as_index=False)['units sold'].sum().so
# Plot
plot1=sns.barplot(y='Publisher ',x='units sold',data=sort_by_publisher,orient='h
plot1.xaxis.set_major_formatter(FuncFormatter(millions))
plt.title('Total Unit Sold By Publisher',fontsize=20)
plt.xlabel('Units Sold',fontsize=15)
plt.ylabel('Publisher',fontsize=15)
plt.figtext(0.5, -0.05, 'Insight: Looks like Amazon Digital Services, Inc. is wa
, ha='center', fontsize=10)
plt.tight_layout()
plt.show()
```



Insight: Looks like Amazon Digital Services, Inc. is way ahead of the competition, selling far more units than any other publisher out there.

Units Sold Across Different Author Rating

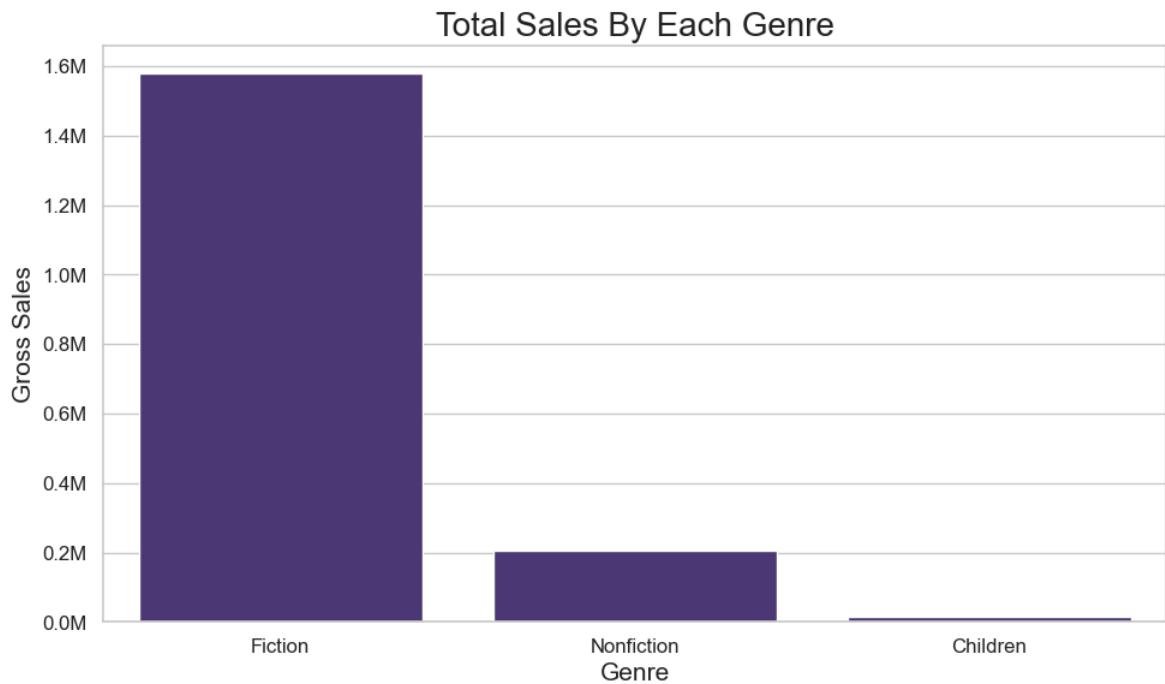
```
In [38]: #Plot
plot2=df.groupby('Author_Rating')['units sold'].sum().sort_values(ascending=False)
plot2.yaxis.set_major_formatter(FuncFormatter(millions))
plt.title('Units Sold Across Different Author Ratings',fontsize=18)
plt.xlabel('Units Sold',fontsize=14)
plt.xticks(rotation=0)
plt.ylabel('Author Rating',fontsize=14)
plt.figtext(0.5, -0.05, 'Insight: Authors with Intermediate and Excellent rating
, ha='center', fontsize=10)
plt.tight_layout()
plt.show()
```



Insight: Authors with Intermediate and Excellent ratings dominate sales, while Novice and Famous authors account for only a small share of units sold.

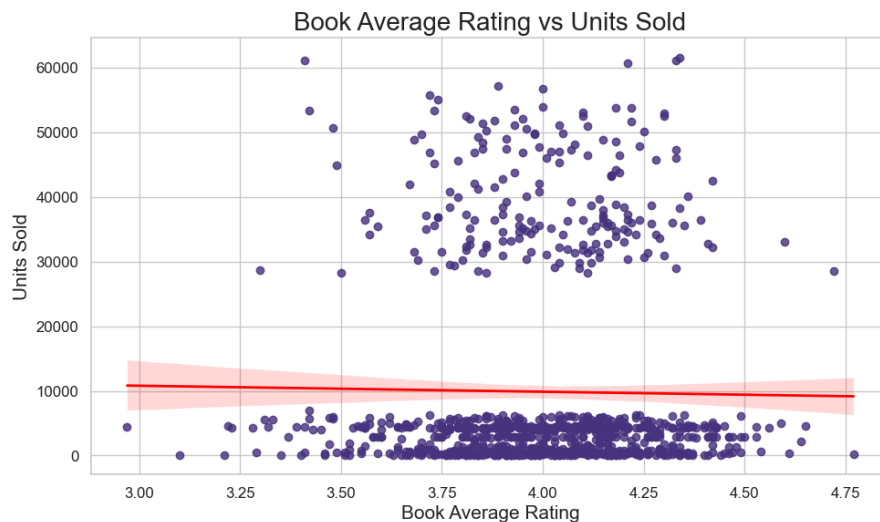
Total Sales By Each Genre

```
In [39]: sort=df.groupby('genre', as_index=False)['gross sales'].sum().sort_values(by='gross sales')
plot3=sns.barplot(x='genre',y='gross sales',data=sort)
plot3.yaxis.set_major_formatter(FuncFormatter(millions))
plt.title('Total Sales By Each Genre',fontsize=20)
plt.xlabel('Genre',fontsize=15)
plt.ylabel('Gross Sales',fontsize=15)
plt.figtext(0.5, -0.05, 'Insight: Fiction is bringing in the highest gross sales', ha='center', fontsize=10)
plt.tight_layout()
plt.show()
```



Insight: Fiction is bringing in the highest gross sales by far, followed by Nonfiction, while Children's books generate the least.

```
In [44]: # Scatter Plot
sns.regplot(x='Book_average_rating',y='units sold',data=df,line_kws={'color':'red'})
plt.title("Book Average Rating vs Units Sold", fontsize=20)
plt.xlabel("Book Average Rating",fontsize=15)
plt.ylabel("Units Sold",fontsize=15)
plt.figtext(0.5, -0.05, 'Insight: The chart shows that book ratings and sales do not really go together. A high rating doesn\'t mean the book will sell more, and a low rating doesn\'t mean it will sell less.', ha='center', fontsize=12)
plt.tight_layout()
plt.show()
```



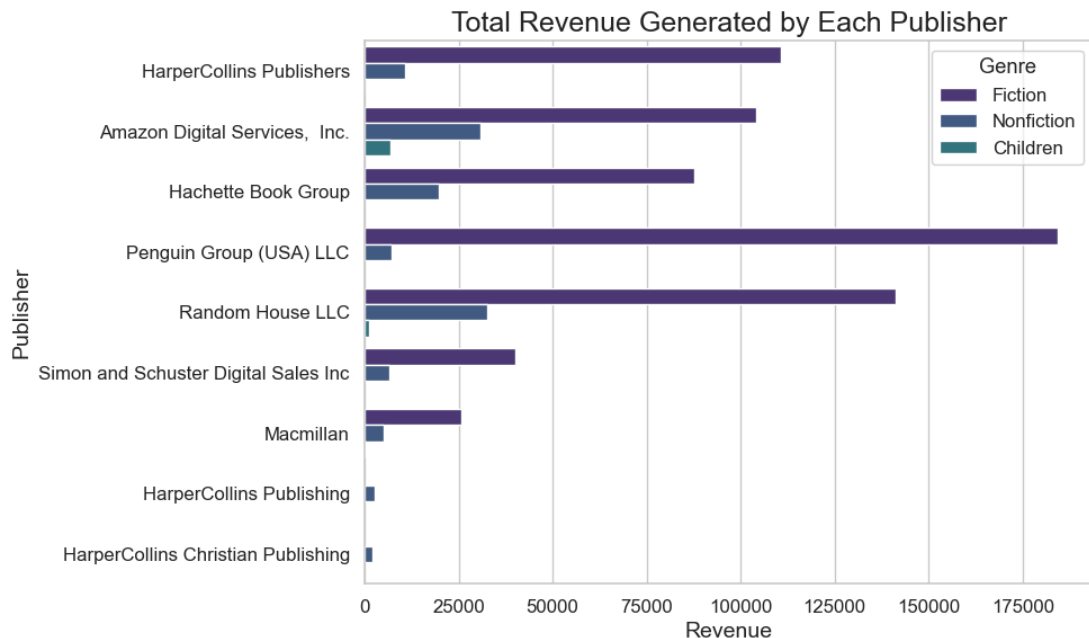
Insight: The chart shows that book ratings and sales don't really go together. A high rating doesn't mean the book will sell more, and a low rating doesn't mean it will sell less.

```
In [41]: # Calculate values to confirm whether there is no relationship between rating and units sold
corr = df['Book_average_rating'].corr(df['units sold'])
print("Correlation Coefficient:", corr)
print(f"The correlation coefficient is {corr:.2f}. Since this value is very close to 0, it suggests there is no meaningful relationship between a book's average rating and the number of units sold.")
```

Correlation Coefficient: -0.014723158455247238

The correlation coefficient is -0.01. Since this value is very close to 0, it suggests there is no meaningful relationship between a book's average rating and the number of units sold.

```
In [42]: sns.barplot(x='publisher revenue',y='Publisher ',data=df,estimator=sum,hue='genre')
plt.title('Total Revenue Generated by Each Publisher',fontsize=18)
plt.xlabel('Revenue',fontsize=14)
plt.ylabel('Publisher',fontsize=14)
plt.figtext(0.5, -0.05, 'Insight: Penguin Group (USA) LLC is making the most mon
, ha='center', fontsize=10)
plt.tight_layout()
plt.legend(title='Genre')
plt.show()
```



Insight: Penguin Group (USA) LLC is making the most money, with Random House LLC not far behind, while HarperCollins Christian Publishing is bringing in the least.

Conclusion

```
In [43]: print(''
Final Conclusion:

1. Amazon Digital Services, Inc. is clearly the top publisher with the highest n
2. Random House LLC and Penguin Group (USA) LLC also have a strong presence but
3. Some big publishers like HarperCollins Publishing and HarperCollins Christian
4. There is no clear link between a book's average rating and its sales – high r
5. Fiction and Non-Fiction are the most popular genres, dominating the market.
6. English is the main publishing language, with other languages having a very s
7. Overall, the analysis shows the leading publishers, popular genres, and how l
''')
```

Final Conclusion:

1. Amazon Digital Services, Inc. is clearly the top publisher with the highest number of books and sales revenue.
2. Random House LLC and Penguin Group (USA) LLC also have a strong presence but are far behind Amazon in sales.
3. Some big publishers like HarperCollins Publishing and HarperCollins Christian Publishing have surprisingly low numbers.
4. There is no clear link between a book's average rating and its sales – high ratings don't always mean more sales.
5. Fiction and Non-Fiction are the most popular genres, dominating the market.
6. English is the main publishing language, with other languages having a very small share.
7. Overall, the analysis shows the leading publishers, popular genres, and how language and ratings affect the market.