

Exploratory Data Analysis (EDA) – Visual Insights Report

Tools Used: Python (Pandas, Matplotlib, Seaborn)

Datasets:

- Netflix Movies and TV Shows
- Iris Dataset

1. Dataset 1: Netflix Movies and TV Shows

Overview:

The Netflix dataset contains details about movies and TV shows including title, type, director, cast, country, release year, rating, duration, and genre.

Distribution of Numerical Features:

- **Release Year:** Most content released after 2015; spike in content after 2018; very few titles before 2000.
- **Duration:** Movies mostly 80–120 minutes; TV shows 1–3 seasons.

Categorical Feature Analysis:

- **Content Type:** 70% movies, 30% TV shows.
- **Rating Distribution:** TV-MA, TV-14, PG-13 most common.
- **Country-wise Content:** USA produces the most, followed by India, UK, and Canada.

Outlier Detection (Box Plot):

- Few movies exceed 200 minutes; most between 90–120 minutes.

Correlation Heatmap:

- Weak correlation between numerical features.
- No strong relationships observed.

Important Features for Prediction:

- Release Year, Duration, Genre, Country, Rating.

```
In [2]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns

df = pd.read_csv("C:/Users/srimulai/Downloads/archive (6)/netflix_titles.csv")

In [3]: df.head()

Out[3]:
   show_id  type      title   director    cast   country  date_added  release_year  rating  duration  listed_in  description
0       s1  Movie  Johnson Is Dead  Kirsten Johnson     NaN  United States  September 25, 2021        2020    PG-13    90 min Documentaries  As her father nears the end of his life, film...
1       s2  TV Show  Blood & Water     NaN          Ama Qamata, Khosi Ngema, Gail Mahalane, Thabani...
2       s3  TV Show      Ganglands  Julien Leclercq     NaN          Sami Bouajila, Tracy Gotoas, Samuel Jouy, Nabi...
3       s4  TV Show  Jailbirds New Orleans     NaN          NaN          NaN  September 24, 2021        2021    TV-MA  1 Season Crime TV Shows, International TV Shows, TV Act...
4       s5  TV Show      Kota Factory     NaN          Mayur More, Jitendra Kumar, Ranjan Raj, Alam K...

```

```
In [4]: df.info()

class 'pandas.core.frame.DataFrame'
RangeIndex: 8887 entries, 0 to 8886
Data columns (total 12 columns):
 #   Column      Non-Null Count  Dtype  
--- 
 0   show_id      8887 non-null   object 
 1   type         8887 non-null   object 
 2   title        8887 non-null   object 
 3   director     6173 non-null   object 
 4   cast          7982 non-null   object 
 5   country      7976 non-null   object 
 6   date_added   8797 non-null   object 
 7   release_year 8887 non-null   int64  
 8   rating        8883 non-null   object 
 9   duration      8884 non-null   object 
 10  listed_in    8887 non-null   object 
 11  description   8887 non-null   object 
dtypes: int64(1), object(11)
memory usage: 825.8+ KB
```

```
In [5]: df.describe()

Out[5]:
      release year
count      8807.000000
mean      2014.180198
std       8.819312
min      1925.000000
25%      2013.000000
50%      2017.000000
75%      2019.000000
max      2021.000000
```

```
In [6]: df.isnull().sum()

Out[6]:
show_id      0
type         0
title        0
director    2624
cast          825
country      831
date_added   10
release_year  0
rating         4
duration       3
listed_in     0
description    0
dtype: int64
```

```
In [7]: df.shape

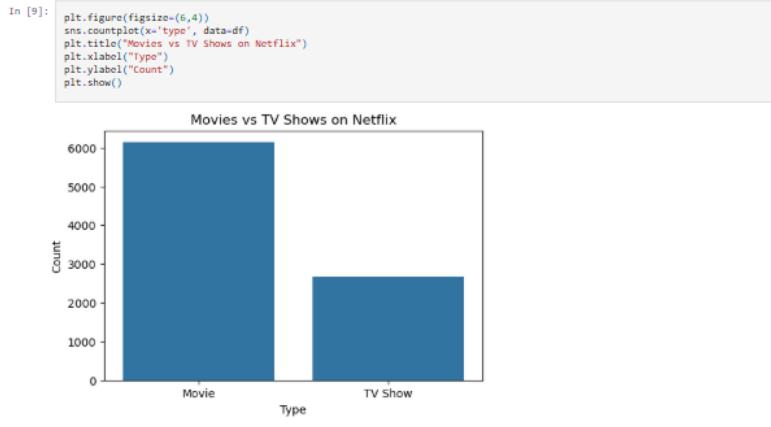
Out[7]: (8887, 12)
```

In [8]:

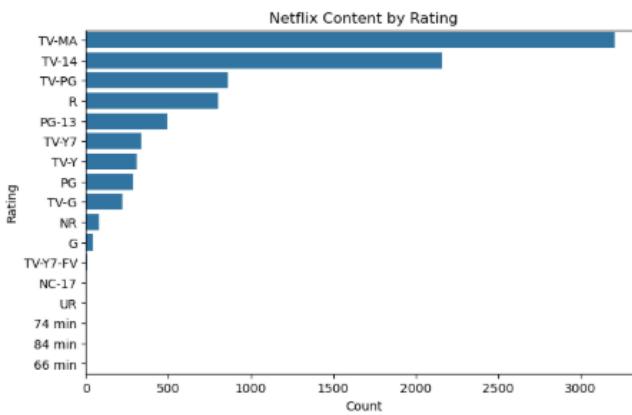
```
df.tail()
```

Out[8]:

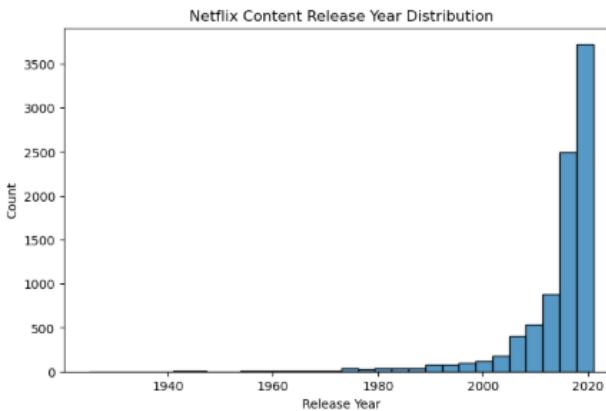
	show_id	type	title	director	cast	country	date_added	release year	rating	duration	listed_in	description
8802	s8803	Movie	Zodiac	David Fincher	Mark Ruffalo, Jake Gyllenhaal, Robert Downey Jr.,...	United States	November 20, 2019	2007	R	158 min	Cult Movies, Dramas, Thrillers	A political cartoonist, a crime reporter and a...
8803	s8804	TV Show	Zombie Dumb	Nan	Nan	Nan	July 1, 2019	2018	TV-Y7	2 Seasons	Kids' TV, Korean TV Shows, TV Comedies	While living alone in a spooky town, a young g...
8804	s8805	Movie	Zombieland	Ruben Fleischer	Jesse Eisenberg, Woody Harrelson, Emma Stone, ...	United States	November 1, 2019	2009	R	88 min	Comedies, Horror Movies	Looking to survive in a world taken over by zo...
8805	s8806	Movie	Zoom	Peter Hewitt	Tim Allen, Courteney Cox, Chevy Chase, Kate M...	United States	January 11, 2020	2006	PG	88 min	Children & Family Movies, Comedies	Dragged from civilian life, a former superhero...
8806	s8807	Movie	Zubaan	Mozz Singh	Vicky Kaushal, Sarah Jane Dias, Raaghav Chanan...	India	March 2, 2019	2015	TV-14	111 min	Dramas, International Movies, Music & Musicals	A scrappy but poor boy worms his way into a ty...



```
In [18]: plt.figure(figsize=(8,5))
sns.countplot(y='rating', data=df, order=df['rating'].value_counts().index)
plt.title("Netflix Content by Rating")
plt.xlabel("Count")
plt.ylabel("Rating")
plt.show()
```

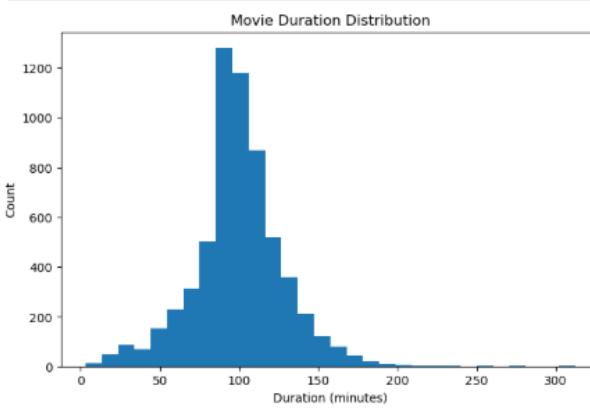


```
In [11]: plt.figure(figsize=(8,5))
sns.histplot(df['release_year'], bins=30)
plt.title("Netflix Content Release Year Distribution")
plt.xlabel("Release Year")
plt.ylabel("Count")
plt.show()
```

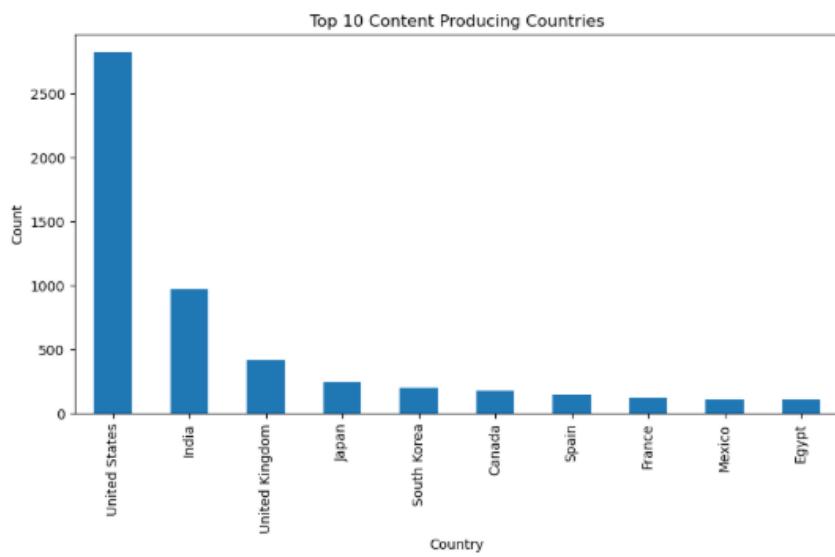


```
In [12]: movies = df[df['type'] == 'Movie'].copy()
movies['duration'] = movies['duration'].str.replace(' min','')
movies['duration'] = pd.to_numeric(movies['duration'])

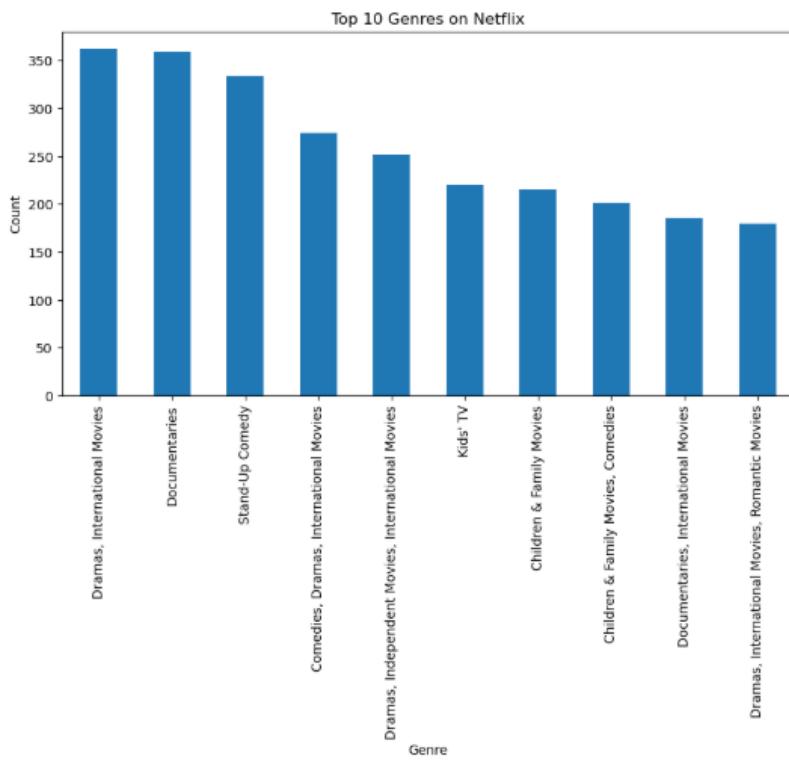
plt.figure(figsize=(8,5))
plt.hist(movies['duration'], bins=30)
plt.title("Movie Duration Distribution")
plt.xlabel("Duration (minutes)")
plt.ylabel("Count")
plt.show()
```



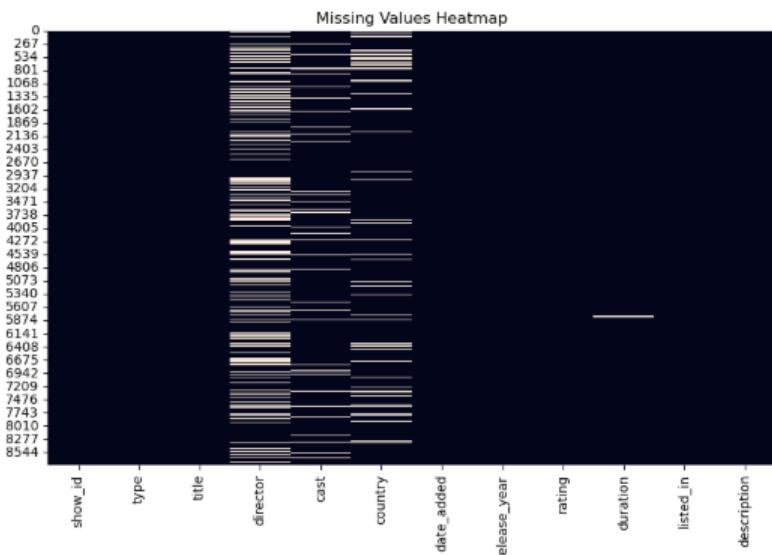
```
In [13]: plt.figure(figsize=(10,5))
df['country'].value_counts().head(10).plot(kind='bar')
plt.title("Top 10 Content Producing Countries")
plt.xlabel("Country")
plt.ylabel("Count")
plt.show()
```



```
In [14]: df['listed_in'].value_counts().head(10).plot(kind='bar', figsize=(10,5))
plt.title("Top 10 Genres on Netflix")
plt.xlabel("Genre")
plt.ylabel("Count")
plt.show()
```



```
In [15]: plt.figure(figsize=(10,6))
sns.heatmap(df.isnull(), cbar=False)
plt.title("Missing Values Heatmap")
plt.show()
```



Conclusion:

- Netflix hosts more movies than TV shows.
- Majority of content is recent (post-2015).
- Teen and adult content dominates.
- USA and India are top content producers.
- Most movies 90–120 minutes; few extremely long movies exist.

2. Dataset 2: Iris Dataset

Overview:

Contains 150 flower samples with 4 features: Sepal Length, Sepal Width, Petal Length, Petal Width, and Species (Setosa, Versicolor, Virginica). Each species has 50 samples.

Distribution of Numerical Features:

- **Sepal Length:** 4.3–7.9 cm; most 5.5–6.5 cm.
- **Sepal Width:** 2.0–4.4 cm; most 2.5–3.5 cm.
- **Petal Length & Width:** Clear separation of species; Setosa has smallest petals.

Categorical Feature Analysis:

- Balanced dataset: 50 samples per species.

Outlier Detection (Box Plot):

- Sepal Width shows few outliers; petal features mostly clean.

Correlation Heatmap:

- Petal Length vs Petal Width: very strong positive correlation
- Sepal Length vs Petal Length: strong positive correlation
- Sepal Width vs Petal Width: weak negative correlation

Important Features for Prediction:

- Petal Length, Petal Width, Sepal Length
- Sepal Width less significant.

```
In [2]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns

df = pd.read_csv(r"C:\Users\srimullai\Downloads\archive (5)\IRIS.csv")

In [3]: df.head()

Out[3]:   sepal_length  sepal_width  petal_length  petal_width  species
0           5.1         3.5          1.4         0.2  Iris-setosa
1           4.9         3.0          1.4         0.2  Iris-setosa
2           4.7         3.2          1.3         0.2  Iris-setosa
3           4.6         3.1          1.5         0.2  Iris-setosa
4           5.0         3.6          1.4         0.2  Iris-setosa

In [4]: df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 150 entries, 0 to 149
Data columns (total 5 columns):
 #   Column      Non-Null Count  Dtype  
 --- 
  0   sepal_length    150 non-null   float64 
  1   sepal_width     150 non-null   float64 
  2   petal_length    150 non-null   float64 
  3   petal_width     150 non-null   float64 
  4   species        150 non-null   object  
 dtypes: float64(4), object(1)
memory usage: 6.0+ KB
```

```
In [5]: df.describe()
```

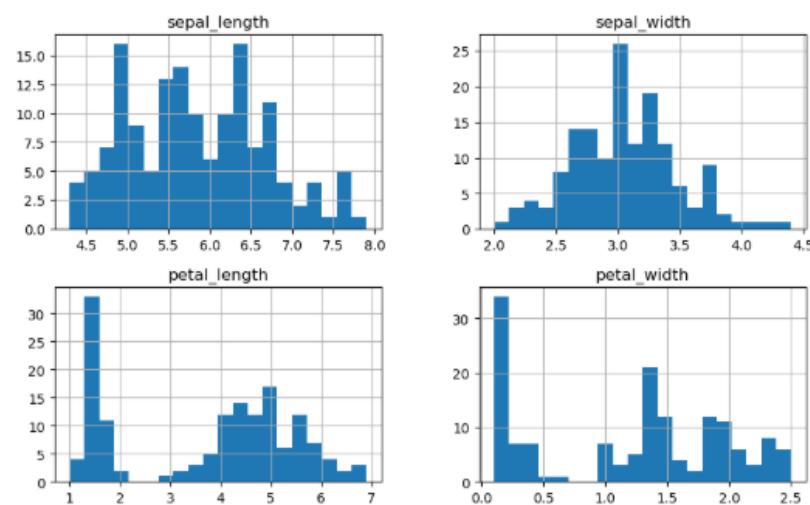
	sepal length	sepal width	petal length	petal width
count	150.000000	150.000000	150.000000	150.000000
mean	5.843333	3.054000	5.758667	1.198667
std	0.828066	0.433594	1.764420	0.763161
min	4.300000	2.000000	1.000000	0.100000
25%	5.100000	2.800000	1.600000	0.300000
50%	5.800000	3.000000	4.350000	1.300000
75%	6.400000	3.300000	5.100000	1.800000
max	7.900000	4.400000	6.900000	2.500000

```
In [8]: df.shape
```

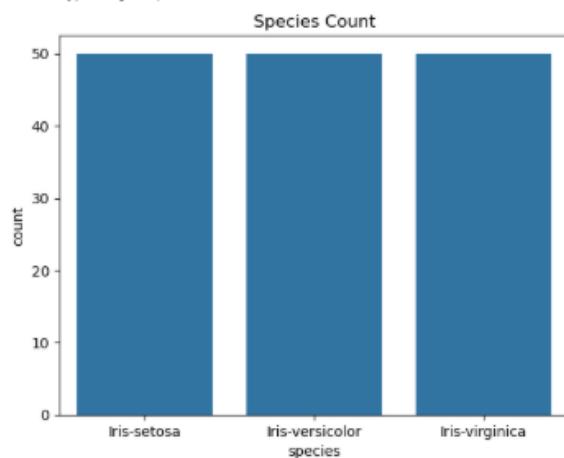
```
Out[8]: (150, 5)
```

```
In [11]: df.hist(figsize=(10,6), bins=20)
plt.suptitle("Distribution of Iris Features")
plt.show()
```

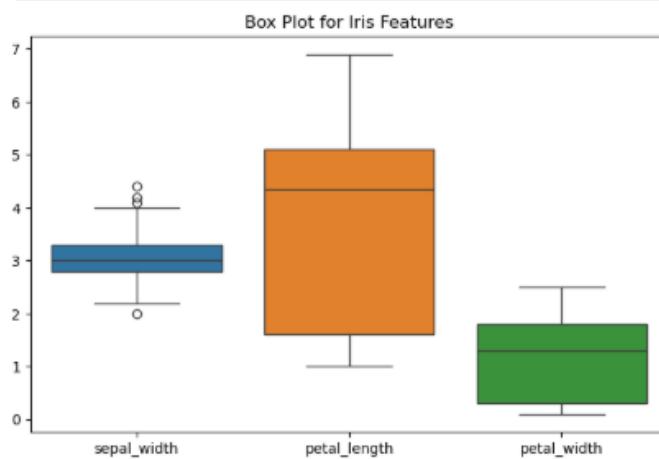
Distribution of Iris Features



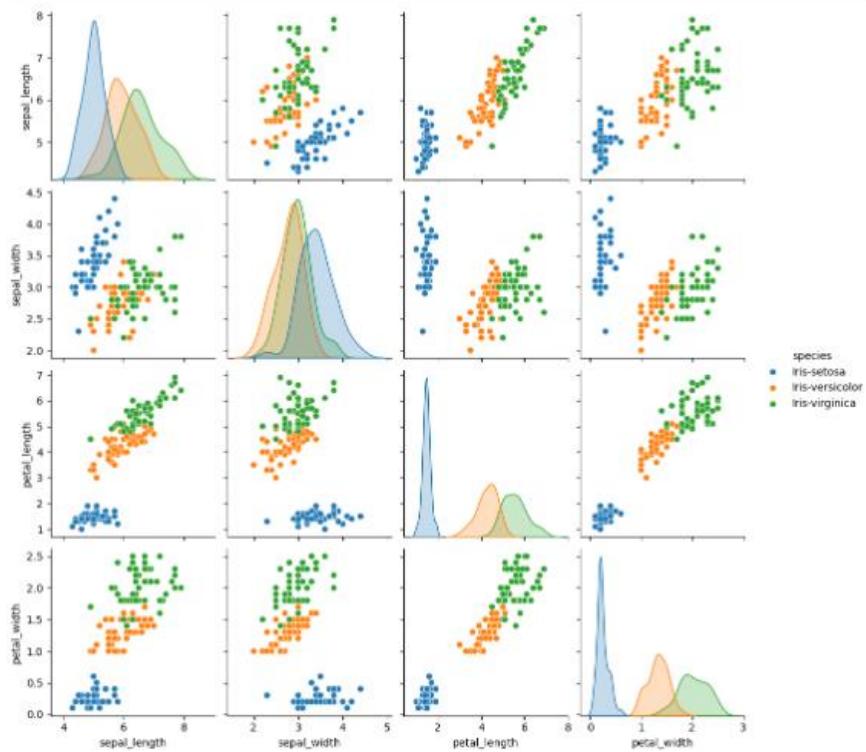
```
In [13]:  
print(df.columns)  
sns.countplot(x='species', data=df)  
plt.title("Species Count")  
plt.show()  
  
Index(['sepal_length', 'sepal_width', 'petal_length', 'petal_width',  
       'species'],  
      dtype='object')
```



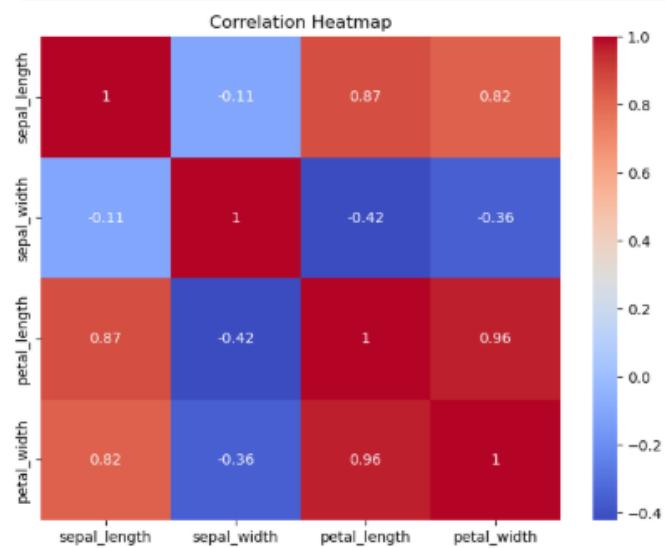
```
In [14]:  
plt.figure(figsize=(8,5))  
sns.boxplot(data=df.iloc[:,1:5])  
plt.title("Box Plot for Iris Features")  
plt.show()
```



```
In [19]: sns.pairplot(df, hue="species")
plt.show()
```



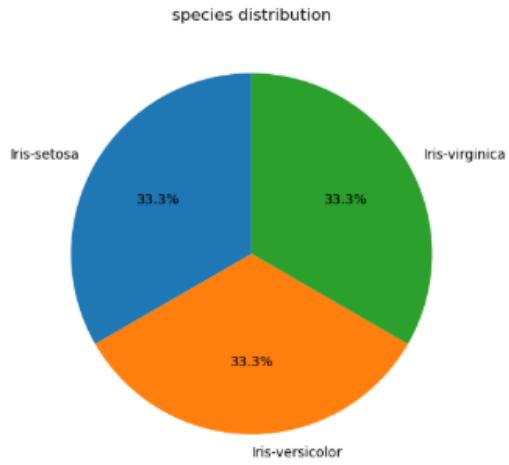
```
In [17]: numerical_cols = df.select_dtypes(include=['float64', 'int64']).columns
plt.figure(figsize=(8,6))
sns.heatmap(df[numerical_cols].corr(), annot=True, cmap="coolwarm")
plt.title("Correlation Heatmap")
plt.show()
```



```
In [20]: df.isnull().sum()
```

```
Out[20]: sepal_length    0
sepal_width     0
petal_length    0
petal_width    0
species        0
dtype: int64
```

```
In [22]: plt.figure(figsize=(6,6))
df['species'].value_counts().plot.pie(autopct='%1.1f%%', startangle=90)
plt.title("species distribution")
plt.ylabel('')
plt.show()
```



Conclusion:

- Clean, balanced dataset
- Petal features clearly separate species
- Strong correlation between petal length and width
- Setosa easiest to classify
- Ideal for ML classification

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