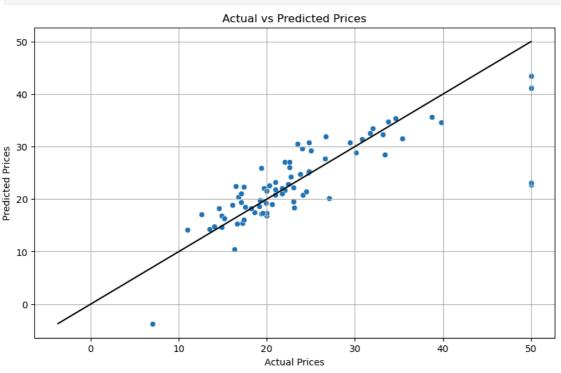
```
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
#Step 2: Load Dataset from URL
url- "https://raw.githubusercontent.com/datasciencedojo/datasets/master/titanic.csv" # Titanic
dataset
# Step 3: Initial Exploration of Data
print("In====- Dataset Overview -======\n")
print(df.head().to string())
df.info()
print("In=== Basic Statistics
\n")
print(df.describe().to_string())
#Step 4: Data Preprocessing - Handling Missing Values
print("In=====Missing Values=====\n")
print(df.isnull().sum())
# Filling missing values
#Age: Filling with median value
df["Age"] = df["Age"].fillna(df["Age"].median())
Embarked: Filling with most frequent value
df["Embarked"] = df["Embarked"].fillna(df["Embarked"].mode()[0])
# Drop Cabin column due to too many missing values
if "Cabin" in df.columns:
df.drop(columns=["Cabin"], inplace=True)
print(df.isnull().sum())
# Convert categorical variables to appropriate types
df['sex'] = df['sex'].astype("category')
df['Embarked'] = dfr'Embarked ].astype("category')
print(df.dtypes)
2.
# pract2
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
from scipy import stats
# Create a sample "Academic performance" dataset
data
```

```
Student_ID': range(1, 11),
Maths_Score': [85, 92, np.nan, 78, 88, 85, 79, 93, 105, 74].
English_Score': [78, 85, 90, 82, np.nan, 88, 74, 91, 86, 80],
Science_Score': [89, 91, 94, 76, 88, np.nan, 72, 85, 98, 74]
df= pd.DataFrame(data)
# Display the initial dataset
print("Initial Dataset:")
print(df)
#1. Scan all variables for missing values and inconsistencies
# Check for missing values
missing values = df.isnull().sum()
print("InMissing values in each column:")
print(missing_values)
# Handling missing values by filling them with the mean of respective columns
df['Maths_score'].fillna(df['Maths_score'].mean(), inplace=True)
df['English_score'].fillna(df['English_Score'].mean(), inplace=True)
df['Science_Score'].fillna(df['Science_Score'].mean(), inplace=True)
# Display dataset after handling missing values
print("InDataset after handling missing values:")
print(df)
#2.Scan all numeric variables for outliers
# Detect outliers using z-score
z scores =np.abs(stats.zscore(df[[ 'Maths score'Science Score, 'English Score', 'Science Score']]))
print("nz-scores of the dataset:")
print(z_scores)
# Set a threshold for Z-score, consider any z-score greater than 3 as an outlier
outliers = (z scores > 3)
print("Inoutliers detected (True means outlier):")
print(outliers)
#3. Apply data transformations on at least one variable
# Applying a log transformation to 'Maths _Score" to decrease skewness
df[ 'Maths_score_log
np.l0g(df[ 'Maths_Score'] + 1)
# visualizing the transformation effect
plt.figure(figsize=(10, 6))
sns.histplot(df[ 'Maths_Score'], kde=True, color='blue', label='Original Maths_Score")
sns.histplot (df[ 'Maths_Score_108
kde=True, color='red'
label='Log-transformed Maths_Score')
plt.legend()
plt.title("Effect of Log Transformation on Maths_Score")
```

```
plt.show()
3.
import pandas as pd
import numpy as np
# Create a sample employee dataset
data = {
Employee ID: [1,2,3,4,5,6,7,8,9,10]
Employee ID: range(1,11),
Age: [25, 30, 35, 40, 28, 45, 50, 60, 30, 38]
Salary': [45000, 55000, 60000, 75000, 48000, 80000]
}
# Create a DataFrame
# dataframe will use dictionary (data) and will convert,
df = pd.DataFrame(data)
# Display the whole dataset
# df.head() - first S rows will be displayed
# df.tail() - last S rows will be displayed
# Define a function to categorize age groups
def categorize_age(age):
if 20 <= age < 30:
return '20-30'
elif 30 <= age
< 40:
return '30-40'
elif 40 < age < 50:
return '40-50'
elif 50 <= age < 60:
return '50-60'
elif 60 <= age < 70:
return '60-70
else:
return 'Unknown:
#In case there are ages outside the range
# Apply the function to the 'Age' column to create a new 'Age Group
df["Age Group'] = df['Age'].apply(categorize_age)
# Group the data
by Age Group and calculate summary statistics for
summary_stats = df.groupby('Age Group') ["Salary'].describe()
# Display the summary statistics
print(summary_stats)
```

```
In [17]: ## 1. Import Libraries
         import pandas as pd
         import matplotlib.pyplot as plt
         import seaborn as sns
         from sklearn.model_selection import train_test_split
         from sklearn.linear_model import LinearRegression
         from sklearn.metrics import mean_squared_error, r2_score
In [23]: ## 2. Load Dataset
         df = pd.read_csv("HousingData.csv")
         df = df.dropna()
In [24]: ## 3. Train-Test Split
         X = df.drop('MEDV', axis=1)
         y = df['MEDV']
         X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
In [26]: ## 4. Train the Model
         model = LinearRegression()
         model.fit(X_train, y_train)
Out[26]: ▼ LinearRegression ①
         LinearRegression()
In [27]: ## 5. Evaluate the Model
         y_pred = model.predict(X_test)
         mse = mean_squared_error(y_test, y_pred)
         r2 = r2_score(y_test, y_pred)
         print(f"Mean Squared Error: {mse}")
         print(f"R^2 Score: {r2}")
        Mean Squared Error: 31.45404766495081
        R^2 Score: 0.6270849941673199
In [34]: plt.figure(figsize=(10,6))
         sns.scatterplot(x=y_test, y=y_pred)
         min_val = min(y_test.min(), y_pred.min())
         max_val = max(y_test.max(), y_pred.max())
         plt.plot([min_val, max_val],
                  [min_val, max_val],
                  linewidth=1.5,
                  color='black')
                                    # solid line
         plt.xlabel("Actual Prices")
         plt.ylabel("Predicted Prices")
         plt.title("Actual vs Predicted Prices")
         plt.grid(True)
         plt.show()
```



```
In [1]: # Import necessary libraries
        import pandas as pd
        import numpy as np
        import matplotlib.pyplot as plt
        import seaborn as sns
        from sklearn.model_selection import train_test_split
        from sklearn.preprocessing import StandardScaler
        from sklearn.linear_model import LogisticRegression
        from sklearn.metrics import confusion_matrix, accuracy_score, classification_report
In [2]: # Load dataset
        df = pd.read_csv('Social_Network_Ads.csv')
        df.head()
Out[2]:
             User ID Gender Age EstimatedSalary Purchased
        0 15624510
                       Male
                              19
                                           19000
                                                          0
        1 15810944
                       Male
                              35
                                           20000
                                                          0
                                           43000
        2 15668575 Female
                              26
                                                          0
        3 15603246 Female
                              27
                                           57000
                                                          0
        4 15804002
                                           76000
                                                          0
                              19
                       Male
In [3]: # Data preprocessing
        # Assuming columns: 'Age', 'EstimatedSalary', 'Purchased'
        X = df[['Age', 'EstimatedSalary']]
        y = df['Purchased']
        # Train-test split
        X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.25, random_state=0)
        # Feature scaling
        sc = StandardScaler()
        X_train = sc.fit_transform(X_train)
        X_test = sc.transform(X_test)
In [4]: # Logistic Regression Model
        classifier = LogisticRegression(random_state=0)
        classifier.fit(X_train, y_train)
Out[4]: 🔻
               LogisticRegression
        LogisticRegression(random_state=0)
In [5]: # Predictions
        y_pred = classifier.predict(X_test)
        # Evaluation
        print(confusion_matrix(y_test, y_pred))
        print(accuracy_score(y_test, y_pred))
        print(classification_report(y_test, y_pred))
       [[65 3]
        [ 8 24]]
       0.89
                                 recall f1-score
                     precision
                                                     support
                          0.89
                                    0.96
                                              0.92
                                                          68
                          0.89
                                    0.75
                                              0.81
                                                          32
                  1
                                              0.89
                                                         100
           accuracy
                          0.89
                                    0.85
                                              0.87
                                                         100
          macro avg
       weighted avg
                          0.89
                                    0.89
                                              0.89
                                                         100
```

Naive Bayes Classification on Iris Dataset

This notebook demonstrates how to implement a simple Naive Bayes classification algorithm using Python on the iris.csv dataset.

```
In [1]: # Import necessary libraries
        import pandas as pd
        from sklearn.model_selection import train_test_split
        from sklearn.naive_bayes import GaussianNB
        from sklearn.metrics import accuracy score, confusion matrix, classification report
In [6]: # Load dataset
        df = pd.read_csv('iris.csv')
       Index(['Id', 'SepalLengthCm', 'SepalWidthCm', 'PetalLengthCm', 'PetalWidthCm',
              'Species'],
            dtype='object')
In [7]: # Split dataset into features and target
        X = df.drop('Species', axis=1)
        y = df['Species']
        # Train-test split
        X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=42)
In [8]: # Train Naive Bayes model
        model = GaussianNB()
        model.fit(X_train, y_train)
Out[8]: ▼ GaussianNB 🗓 🖰
        GaussianNB()
In [9]: # Predict and evaluate
        y_pred = model.predict(X_test)
        print('Accuracy:', accuracy_score(y_test, y_pred))
        print('Confusion Matrix:\n', confusion_matrix(y_test, y_pred))
        print('Classification Report:\n', classification_report(y_test, y_pred))
       Accuracy: 1.0
       Confusion Matrix:
       [[19 0 0]
       [ 0 13 0]
       [ 0 0 13]]
       Classification Report:
                       precision recall f1-score support
          Iris-setosa
                                    1.00
                                              1.00
                                                           19
                          1.00
                          1.00
                                    1.00
                                              1.00
       Iris-versicolor
                                                           13
                          1.00
       Iris-virginica
                                    1.00
                                               1.00
                                                           13
                                               1.00
                                                          45
             accuracy
                      1.00
1.00
            macro avg
                                     1.00
                                               1.00
                                                           45
         weighted avg
                                     1.00
                                               1.00
                                                           45
```

```
In [3]: import os
         import nltk
         from nltk.tokenize import word_tokenize, sent_tokenize
         from nltk.tag import pos_tag
         from nltk.corpus import stopwords
         from nltk.stem import PorterStemmer, WordNetLemmatizer
         from nltk.probability import FreqDist
         import math
         import matplotlib.pyplot as plt
         import pandas as pd
        WARNING: Skipping C:\Users\Adnan\anaconda3\Lib\site-packages\numpy-2.2.4.dist-inf
        o due to invalid metadata entry 'name'
        WARNING: Skipping C:\Users\Adnan\anaconda3\Lib\site-packages\numpy-2.2.4.dist-inf
        o due to invalid metadata entry 'name'
        WARNING: Skipping C:\Users\Adnan\anaconda3\Lib\site-packages\numpy-2.2.4.dist-inf
        o due to invalid metadata entry 'name'
        WARNING: Skipping C:\Users\Adnan\anaconda3\Lib\site-packages\numpy-2.2.4.dist-inf
        o due to invalid metadata entry 'name'
        WARNING: Skipping C:\Users\Adnan\anaconda3\Lib\site-packages\numpy-2.2.4.dist-inf
        o due to invalid metadata entry 'name'
        WARNING: Skipping C:\Users\Adnan\anaconda3\Lib\site-packages\numpy-2.2.4.dist-inf
        o due to invalid metadata entry 'name'
        Requirement already satisfied: nltk in c:\users\adnan\anaconda3\lib\site-packages
        (3.8.1)
        Requirement already satisfied: click in c:\users\adnan\anaconda3\lib\site-package
        s (from nltk) (8.1.7)
        Requirement already satisfied: joblib in c:\users\adnan\anaconda3\lib\site-packag
        es (from nltk) (1.4.2)
        Requirement already satisfied: regex>=2021.8.3 in c:\users\adnan\anaconda3\lib\si
        te-packages (from nltk) (2023.10.3)
        Requirement already satisfied: tqdm in c:\users\adnan\anaconda3\lib\site-packages
        (from nltk) (4.66.4)
        Requirement already satisfied: colorama in c:\users\adnan\anaconda3\lib\site-pack
        ages (from click->nltk) (0.4.6)
In [45]: sent = "Sachin is considered to be one of the greatest cricket players. Virat is
         print(word tokenize(sent))
         print(sent_tokenize(sent))
        ['Sachin', 'is', 'considered', 'to', 'be', 'one', 'of', 'the', 'greatest', 'crick
        et', 'players', '.', 'Virat', 'is', 'the', 'captain', 'of', 'the', 'Indian', 'cri
        cket', 'team']
        ['Sachin is considered to be one of the greatest cricket players.', 'Virat is the
        captain of the Indian cricket team']
In [47]: nltk.download('stopwords')
         stop_words = stopwords.words('english')
```

print(stop_words)

```
['a', 'about', 'above', 'after', 'again', 'against', 'ain', 'all', 'am', 'an', 'a
nd', 'any', 'are', 'aren', "aren't", 'as', 'at', 'be', 'because', 'been', 'befor
e', 'being', 'below', 'between', 'both', 'but', 'by', 'can', 'couldn', "could
n't", 'd', 'did', 'didn', "didn't", 'do', 'does', 'doesn', "doesn't", 'doing', 'd
on', "don't", 'down', 'during', 'each', 'few', 'for', 'from', 'further', 'had',
'hadn', "hadn't", 'has', 'hasn', "hasn't", 'have', 'haven', "haven't", 'having',
'he', "he'd", "he'll", 'her', 'here', 'hers', 'herself', "he's", 'him', 'himsel
f', 'his', 'how', 'i', "i'd", 'if', "i'll", "i'm", 'in', 'into', 'is', 'isn', "is
n't", 'it', "it'd", "it'll", "it's", 'its', 'itself', "i've", 'just', 'll', 'm',
'ma', 'me', 'mightn', "mightn't", 'more', 'most', 'mustn', "mustn't", 'my', 'myse
lf', 'needn', "needn't", 'no', 'nor', 'not', 'now', 'o', 'of', 'off', 'on', 'onc
e', 'only', 'or', 'other', 'our', 'ours', 'ourselves', 'out', 'over', 'own', 'r
e', 's', 'same', 'shan', "shan't", 'she', "she'd", "she'll", "she's", 'should',
'shouldn', "shouldn't", "should've", 'so', 'some', 'such', 't', 'than', 'that',
"that'll", 'the', 'their', 'theirs', 'them', 'themselves', 'then', 'there', 'thes
e', 'they', "they'd", "they'll", "they're", "they've", 'this', 'those', 'throug
h', 'to', 'too', 'under', 'until', 'up', 've', 'very', 'was', 'wasn', "wasn't",
'we', "we'd", "we'll", "we're", 'weren', "weren't", "we've", 'what', 'whe
n', 'where', 'which', 'while', 'who', 'whom', 'why', 'will', 'with', 'won', "wo
n't", 'wouldn', "wouldn't", 'y', 'you', "you'd", "you'll", 'your', "you're", 'you
rs', 'yourself', 'yourselves', "you've"]
[nltk data] Downloading package stopwords to
[nltk_data] C:\Users\Adnan\nltk_data...
[nltk_data] Package stopwords is already up-to-date!
```

```
In [49]: token = word_tokenize(sent)
  cleaned_token = []
  for word in token:
    if word not in stop_words:
       cleaned_token.append(word)

print("This is the unclean version : ",token)
  print("This is the cleaned version : ",cleaned_token)
```

This is the unclean version: ['Sachin', 'is', 'considered', 'to', 'be', 'one', 'of', 'the', 'greatest', 'cricket', 'players', '.', 'Virat', 'is', 'the', 'captain', 'of', 'the', 'Indian', 'cricket', 'team']

This is the cleaned version: ['Sachin', 'considered', 'one', 'greatest', 'cricket', 'players', '.', 'Virat', 'captain', 'Indian', 'cricket', 'team']

In [53]: words = [cleaned_token.lower() for cleaned_token in cleaned_token if cleaned_tok
print(words)

['sachin', 'considered', 'one', 'greatest', 'cricket', 'players', 'virat', 'capta
in', 'indian', 'cricket', 'team']

```
In [57]: stemmer = PorterStemmer()
    port_stemmer_output = [stemmer.stem(words) for words in words]
    print(port_stemmer_output)
```

['sachin', 'consid', 'one', 'greatest', 'cricket', 'player', 'virat', 'captain',
'indian', 'cricket', 'team']

```
In [59]: nltk.download('wordnet')
  lemmatizer = WordNetLemmatizer()
  lemmatizer_output = [lemmatizer.lemmatize(words) for words in words]
  print(lemmatizer_output)
```

['sachin', 'considered', 'one', 'greatest', 'cricket', 'player', 'virat', 'captai
n', 'indian', 'cricket', 'team']

```
[nltk_data] Downloading package wordnet to C:\Users\Adnan\nltk_data...
       [nltk_data] Package wordnet is already up-to-date!
In [61]: nltk.download('averaged perceptron tagger')
         token = word_tokenize(sent)
         cleaned token = []
         for word in token:
          if word not in stop_words:
             cleaned_token.append(word)
         tagged = pos_tag(cleaned_token)
         print(tagged)
        [('Sachin', 'NNP'), ('considered', 'VBD'), ('one', 'CD'), ('greatest', 'JJS'),
        ('cricket', 'NN'), ('players', 'NNS'), ('.', '.'), ('Virat', 'NNP'), ('captain',
        'NN'), ('Indian', 'JJ'), ('cricket', 'NN'), ('team', 'NN')]
        [nltk data] Downloading package averaged perceptron tagger to
        [nltk_data] C:\Users\Adnan\nltk_data...
        [nltk_data]
                      Package averaged_perceptron_tagger is already up-to-
        [nltk_data]
                          date!
In [63]: from sklearn.feature_extraction.text import TfidfVectorizer
         from sklearn.metrics.pairwise import cosine_similarity
         import pandas as pd
In [65]: docs = [ "Sachin is considered to be one of the greatest cricket players",
          "Federer is considered one of the greatest tennis players",
          "Nadal is considered one of the greatest tennis players",
          "Virat is the captain of the Indian cricket team"]
In [67]: vectorizer = TfidfVectorizer(analyzer = "word", norm = None , use_idf = True , s
         Mat = vectorizer.fit(docs)
         print(Mat.vocabulary_)
        {'sachin': 12, 'is': 7, 'considered': 2, 'to': 16, 'be': 0, 'one': 10, 'of': 9,
        'the': 15, 'greatest': 5, 'cricket': 3, 'players': 11, 'federer': 4, 'tennis': 1
        4, 'nadal': 8, 'virat': 17, 'captain': 1, 'indian': 6, 'team': 13}
In [69]: tfidfMat = vectorizer.fit_transform(docs)
In [71]: features_names = vectorizer.get_feature_names_out()
         print(features_names)
        ['be' 'captain' 'considered' 'cricket' 'federer' 'greatest' 'indian' 'is'
         'nadal' 'of' 'one' 'players' 'sachin' 'team' 'tennis' 'the' 'to' 'virat']
In [73]: dense = tfidfMat.todense()
         denselist = dense.tolist()
         df = pd.DataFrame(denselist , columns = features_names)
In [75]: df
```

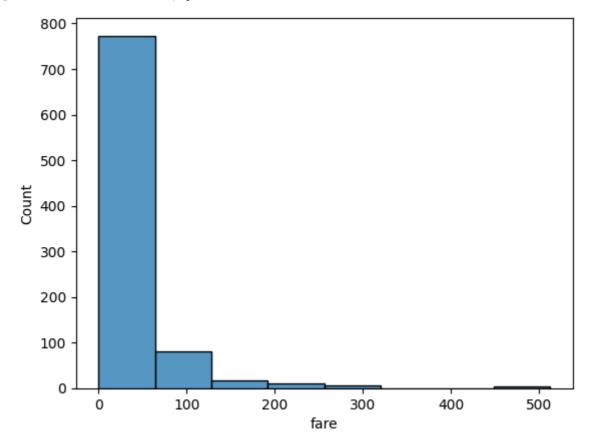
```
Out[75]:
                 be
                   captain considered
                                         cricket
                                                 federer greatest
                                                                    indian is
                                                                                 nadal
         0 1.916291 0.000000
                               1.223144 1.510826 0.000000 1.223144 0.000000 1.0 0.000000
         1 0.000000 0.000000
                               1.223144 0.000000 1.916291 1.223144 0.000000 1.0
                                                                              0.000000
         2 0.000000 0.000000
                               1.223144 0.000000 0.000000 1.223144 0.000000
                                                                          1.0
                                                                              1.916291
         3 0.000000 1.916291
                               0.000000 1.510826 0.000000 0.000000 1.916291 1.0
                                                                             0.000000
        features_names = sorted(vectorizer.get_feature_names_out())
In [79]:
         docList = ['Doc 1','Doc 2','Doc 3','Doc 4']
In [81]:
         skDocsIfIdfdf = pd.DataFrame(tfidfMat.todense(),index = sorted(docList), columns
         print(skDocsIfIdfdf)
                    be
                        captain considered
                                            cricket federer greatest
                                                                           indian
       Doc 1 1.916291 0.000000
                                   1.223144 1.510826 0.000000 1.223144
                                                                         0.000000
       Doc 2 0.000000 0.000000
                                 1.223144 0.000000 1.916291 1.223144 0.000000
       Doc 3 0.000000 0.000000 1.223144 0.000000 0.000000 1.223144
                                                                         0.000000
       Doc 4 0.000000 1.916291
                                   0.000000 1.510826 0.000000 0.000000 1.916291
                             of
               is
                      nadal
                                      one
                                            players
                                                      sachin
                                                                  team
                                                                         tennis \
       Doc 1 1.0 0.000000 1.0 1.223144 1.223144 1.916291 0.000000 0.0000000
       Doc 2 1.0 0.000000 1.0 1.223144 1.223144 0.000000
                                                              0.000000 1.510826
       Doc 3 1.0 1.916291 1.0 1.223144 1.223144 0.000000
                                                              0.000000 1.510826
       Doc 4 1.0 0.000000 1.0 0.000000 0.000000 0.000000 1.916291 0.000000
              the
                               virat
       Doc 1 1.0 1.916291 0.000000
       Doc 2 1.0 0.000000 0.000000
       Doc 3 1.0 0.000000 0.000000
       Doc 4 2.0 0.000000 1.916291
In [83]: csim = cosine similarity(tfidfMat,tfidfMat)
In [85]: csimDf = pd.DataFrame(csim,index=sorted(docList),columns=sorted(docList))
In [87]:
         print(csimDf)
                          Doc 2
                                    Doc 3
                                             Doc 4
                 Doc 1
       Doc 1 1.000000 0.492416 0.492416
                                           0.277687
       Doc 2 0.492416 1.000000 0.754190
                                           0.215926
       Doc 3 0.492416 0.754190 1.000000
                                           0.215926
       Doc 4 0.277687 0.215926 0.215926 1.000000
```

```
In [5]: #!pip install seaborn
        import seaborn as sns
        import pandas as pd
        # Load the titanic dataset
        titanic = sns.load_dataset("titanic")
        # Verify it worked with a simple plot
        titanic.head(1)
Out[5]:
                           sex age sibsp parch fare embarked class who adult_male
           survived pclass
                       3 male 22.0
                                              0 7.25
                                                            S Third
                                                                     man
                                                                                True
In [7]: titanic.info()
      <class 'pandas.core.frame.DataFrame'>
      RangeIndex: 891 entries, 0 to 890
      Data columns (total 15 columns):
          Column
                       Non-Null Count Dtype
                       -----
       0
           survived
                       891 non-null
                                       int64
       1
                       891 non-null int64
           pclass
       2
                      891 non-null
                                      object
           sex
                       714 non-null
                                      float64
       3
           age
                      891 non-null
       4
           sibsp
                                      int64
       5
                      891 non-null int64
          parch
          fare
                      891 non-null float64
       7
           embarked 889 non-null
                                      object
                      891 non-null category
       8
           class
       9
           who
                      891 non-null
                                      object
       10 adult_male 891 non-null
                                      bool
       11 deck
                       203 non-null
                                      category
       12 embark_town 889 non-null
                                      object
       13 alive
                       891 non-null
                                       object
                       891 non-null
                                       bool
       14 alone
      dtypes: bool(2), category(2), float64(2), int64(4), object(5)
      memory usage: 80.7+ KB
In [9]: titanic.describe()
```

```
Out[9]:
                   survived
                                 pclass
                                               age
                                                         sibsp
                                                                    parch
                                                                                 fare
          count 891.000000 891.000000 714.000000 891.000000
                                                               891.000000
                                                                           891.000000
                   0.383838
                               2.308642
                                         29.699118
                                                      0.523008
                                                                  0.381594
                                                                            32.204208
          mean
            std
                   0.486592
                               0.836071
                                         14.526497
                                                      1.102743
                                                                  0.806057
                                                                            49.693429
                   0.000000
                               1.000000
                                          0.420000
                                                      0.000000
                                                                  0.000000
                                                                             0.000000
            min
           25%
                   0.000000
                               2.000000
                                         20.125000
                                                      0.000000
                                                                  0.000000
                                                                             7.910400
           50%
                   0.000000
                               3.000000
                                         28.000000
                                                      0.000000
                                                                  0.000000
                                                                            14.454200
           75%
                   1.000000
                                         38.000000
                               3.000000
                                                      1.000000
                                                                  0.000000
                                                                            31.000000
                   1.000000
                               3.000000
                                         80.000000
                                                      8.000000
                                                                  6.000000 512.329200
           max
In [11]:
          df=titanic.drop(['pclass','embarked','deck','embark_town'],axis=1)
          df.head(2)
Out[11]:
             survived
                              age sibsp
                                                         class
                                                                  who adult_male alive
                                                                                          alon
                         sex
                                          parch
                                                    fare
          0
                    0
                              22.0
                                              0
                                                  7.2500 Third
                                                                                           Fals
                        male
                                                                  man
                                                                              True
                                                                                      no
                    1 female 38.0
                                              0 71.2833
                                                          First woman
                                                                              False
                                                                                     yes
                                                                                           Fals
In [15]: df.info()
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 891 entries, 0 to 890
        Data columns (total 11 columns):
             Column
                          Non-Null Count Dtype
             -----
                          -----
        ---
                                           ----
         0
             survived
                          891 non-null
                                           int64
                          891 non-null
                                           object
         1
             sex
         2
             age
                          714 non-null
                                           float64
         3
             sibsp
                          891 non-null
                                           int64
                          891 non-null
                                           int64
         4
             parch
         5
             fare
                          891 non-null
                                           float64
                          891 non-null
         6
             class
                                           category
         7
             who
                          891 non-null
                                           object
         8
             adult_male 891 non-null
                                           bool
         9
             alive
                          891 non-null
                                           object
         10
             alone
                          891 non-null
                                           bool
        dtypes: bool(2), category(1), float64(2), int64(3), object(3)
        memory usage: 58.6+ KB
In [17]: df.isnull().sum()
```

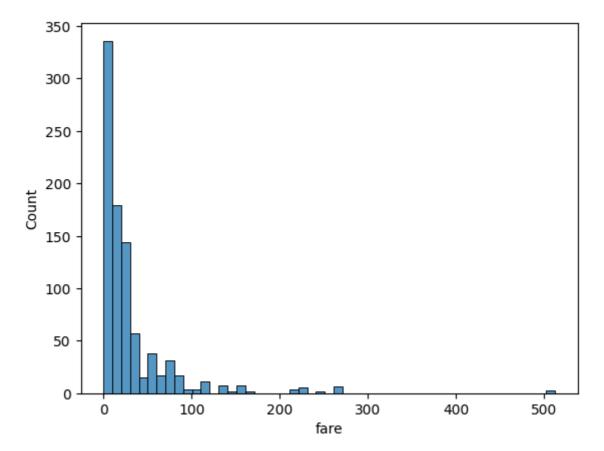
```
Out[17]: survived
                          0
                          0
          sex
                        177
          age
          sibsp
                          0
          parch
                          0
          fare
                          0
          class
                          0
          who
                          0
          adult_male
          alive
                          0
          alone
          dtype: int64
In [21]:
         numeric_df = df.select_dtypes(include=['float64', 'int64'])
          correlation = numeric_df.corr(method='pearson')
In [23]: sns.histplot(data=titanic,x="fare",bins=8)
```

Out[23]: <Axes: xlabel='fare', ylabel='Count'>



```
In [30]: sns.histplot(data=titanic,x="fare",bins=20,binwidth=10)
```

Out[30]: <Axes: xlabel='fare', ylabel='Count'>



In [34]: sns.distplot(df['fare'])

C:\Users\Adnan\AppData\Local\Temp\ipykernel_23152\1195996103.py:1: UserWarning:

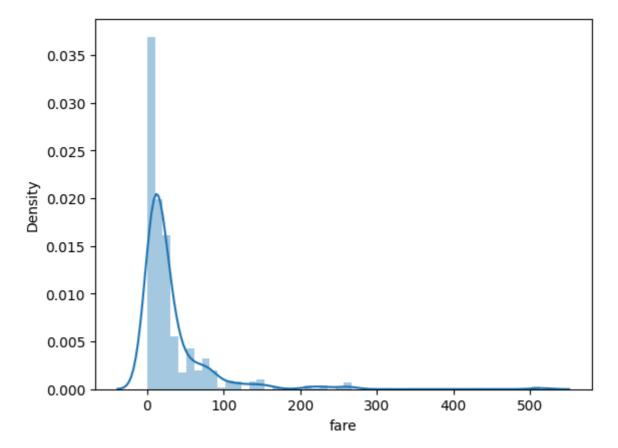
`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751

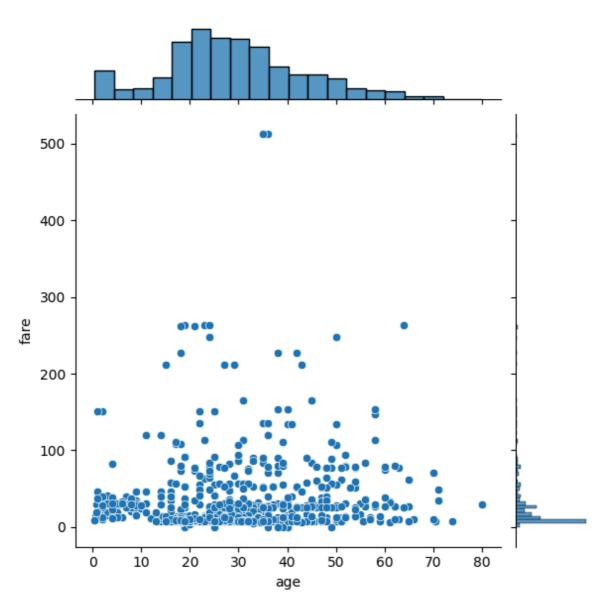
sns.distplot(df['fare'])

Out[34]: <Axes: xlabel='fare', ylabel='Density'>



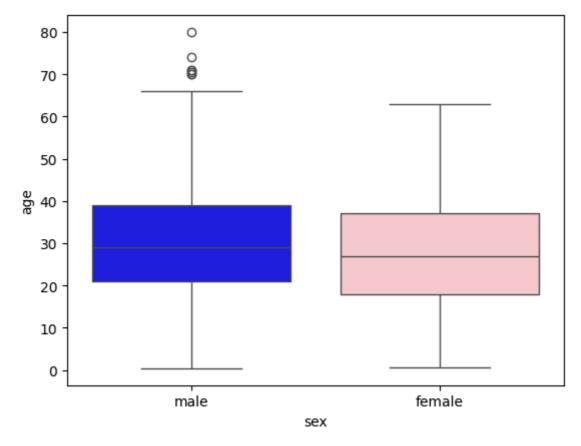
In [44]: sns.jointplot(x='age', y='fare', data=df)

Out[44]: <seaborn.axisgrid.JointGrid at 0x1bf37616990>



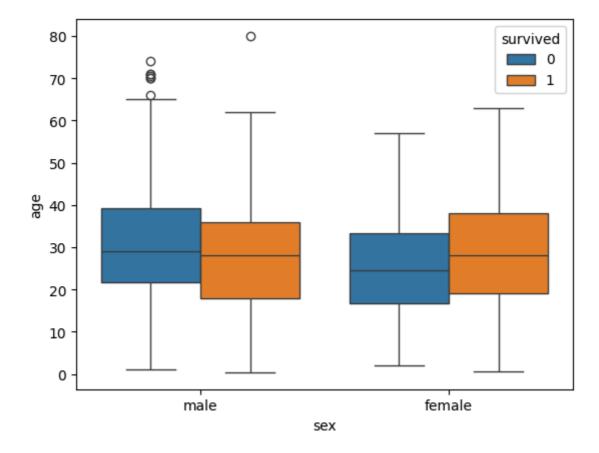
```
In [4]:
         import seaborn as sns
         import pandas as pd
         # Load the titanic dataset
         titanic = sns.load_dataset("titanic")
         # Verify it worked with a simple plot
         titanic.head(1)
Out[4]:
            survived pclass
                                   age sibsp parch fare embarked
                                                                      class who adult_male
                              sex
         0
                  0
                                 22.0
                                                     7.25
                                                                     Third
                          3 male
                                            1
                                                   0
                                                                            man
                                                                                         True
In [6]: titanic.info()
       <class 'pandas.core.frame.DataFrame'>
       RangeIndex: 891 entries, 0 to 890
       Data columns (total 15 columns):
            Column
                          Non-Null Count Dtype
        0
            survived
                          891 non-null
                                           int64
        1
            pclass
                          891 non-null
                                           int64
                          891 non-null
        2
            sex
                                           object
                                           float64
        3
            age
                          714 non-null
                          891 non-null
                                           int64
        4
            sibsp
        5
            parch
                          891 non-null
                                           int64
                          891 non-null
                                           float64
        6
            fare
        7
            embarked
                          889 non-null
                                           object
        8
            class
                          891 non-null
                                           category
                                           object
        9
            who
                          891 non-null
        10 adult_male
                          891 non-null
                                           bool
        11 deck
                          203 non-null
                                           category
        12
            embark_town 889 non-null
                                           object
        13
                          891 non-null
                                           object
           alive
        14 alone
                          891 non-null
                                           bool
       dtypes: bool(2), category(2), float64(2), int64(4), object(5)
       memory usage: 80.7+ KB
In [8]: titanic.describe()
Out[8]:
                  survived
                                pclass
                                                        sibsp
                                              age
                                                                   parch
                                                                                fare
         count
               891.000000
                           891.000000
                                       714.000000
                                                   891.000000
                                                              891.000000
                                                                          891.000000
                  0.383838
                              2.308642
                                        29.699118
                                                     0.523008
                                                                 0.381594
                                                                           32.204208
         mean
           std
                  0.486592
                              0.836071
                                        14.526497
                                                     1.102743
                                                                 0.806057
                                                                           49.693429
          min
                  0.000000
                              1.000000
                                         0.420000
                                                     0.000000
                                                                 0.000000
                                                                            0.000000
          25%
                  0.000000
                              2.000000
                                        20.125000
                                                     0.000000
                                                                 0.000000
                                                                            7.910400
          50%
                  0.000000
                              3.000000
                                        28.000000
                                                     0.000000
                                                                 0.000000
                                                                           14.454200
          75%
                  1.000000
                              3.000000
                                        38.000000
                                                     1.000000
                                                                 0.000000
                                                                           31.000000
                  1.000000
                              3.000000
                                        80.000000
                                                     8.000000
                                                                 6.000000
                                                                          512.329200
          max
```

Out[16]: <Axes: xlabel='sex', ylabel='age'>

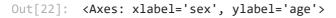


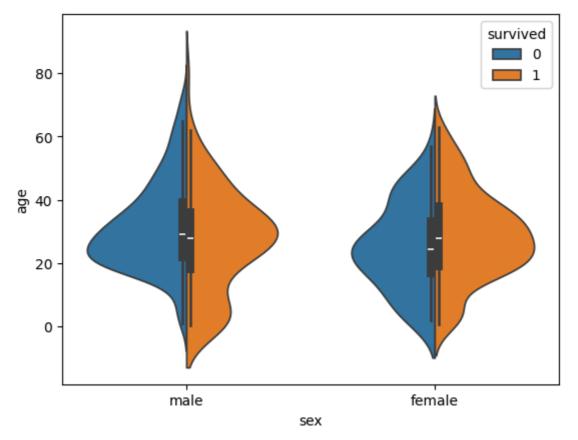
In [18]: sns.boxplot(x="sex",y="age",data=titanic,hue="survived")

Out[18]: <Axes: xlabel='sex', ylabel='age'>



In [22]: sns.violinplot(x='sex', y='age', data=titanic, hue='survived', split=True)

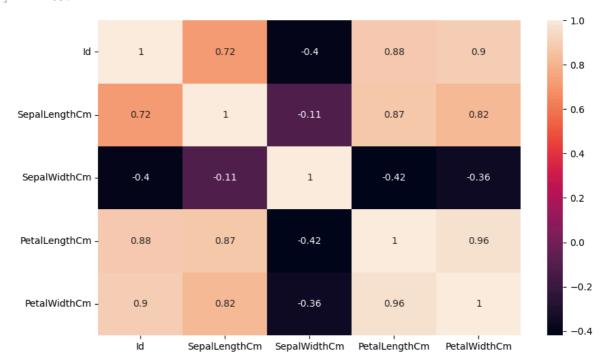




```
In [1]:
          import matplotlib.pyplot as plt
          import seaborn as sns
          import pandas as pd
 In [3]: df = pd.read_csv("https://raw.githubusercontent.com/shrikant-temburwar/Iris-Data
          df.head(3)
 Out[3]:
             Id SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm
                                                                                 Species
          0
              1
                            5.1
                                            3.5
                                                            1.4
                                                                           0.2 Iris-setosa
              2
          1
                             4.9
                                            3.0
                                                            1.4
                                                                           0.2 Iris-setosa
          2
              3
                            4.7
                                            3.2
                                                            1.3
                                                                           0.2 Iris-setosa
         df.describe()
 In [5]:
 Out[5]:
                         Id SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm
          count 150.000000
                                 150.000000
                                                 150.000000
                                                                 150.000000
                                                                                150.000000
                  75.500000
                                   5.843333
                                                   3.054000
                                                                   3.758667
                                                                                  1.198667
          mean
                  43.445368
                                   0.828066
                                                   0.433594
                                                                   1.764420
                                                                                  0.763161
            std
                   1.000000
                                   4.300000
                                                   2.000000
                                                                   1.000000
                                                                                  0.100000
            min
           25%
                  38.250000
                                   5.100000
                                                   2.800000
                                                                   1.600000
                                                                                  0.300000
           50%
                  75.500000
                                   5.800000
                                                   3.000000
                                                                                  1.300000
                                                                  4.350000
                112.750000
                                   6.400000
                                                                                  1.800000
           75%
                                                   3.300000
                                                                   5.100000
                150.000000
                                   7.900000
                                                                   6.900000
                                                                                  2.500000
                                                   4.400000
           max
 In [7]:
         df.shape
 Out[7]: (150, 6)
 In [9]: df["Species"].unique()
 Out[9]: array(['Iris-setosa', 'Iris-versicolor', 'Iris-virginica'], dtype=object)
In [11]: df.groupby("Species").size()
Out[11]: Species
          Iris-setosa
                              50
          Iris-versicolor
                              50
          Iris-virginica
                              50
          dtype: int64
In [15]: numeric_df = df.select_dtypes(include=['float64', 'int64'])
          # Now calculate correlation only on numerical columns
          corr = numeric df.corr()
          # Plot the heatmap
```

```
plt.subplots(figsize=(10,6))
sns.heatmap(corr, annot=True)
```

Out[15]: <Axes: >



```
In [21]: def graph(y, position, title, palette):
             plt.subplot(2, 2, position)
             sns.boxplot(x="Species", y=y, data=df, palette=palette)
             plt.title(title)
             plt.xlabel("")
             plt.ylabel(y)
         # Features, titles, and custom palettes for variety
         features = ['SepalLengthCm', 'SepalWidthCm', 'PetalLengthCm', 'PetalWidthCm']
         titles = ['Sepal Length by Species', 'Sepal Width by Species',
                   'Petal Length by Species', 'Petal Width by Species']
         palettes = ['Set2', 'Pastel1', 'coolwarm', 'husl']
         plt.figure(figsize=(12, 10))
         for i, (feature, title, palette) in enumerate(zip(features, titles, palettes), 1
             graph(feature, i, title, palette)
         plt.tight_layout()
         plt.show()
```

C:\Users\Adnan\AppData\Local\Temp\ipykernel_21992\2327543419.py:3: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v 0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.

sns.boxplot(x="Species", y=y, data=df, palette=palette)

C:\Users\Adnan\AppData\Local\Temp\ipykernel_21992\2327543419.py:3: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v 0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.

sns.boxplot(x="Species", y=y, data=df, palette=palette)

C:\Users\Adnan\AppData\Local\Temp\ipykernel_21992\2327543419.py:3: FutureWarning:

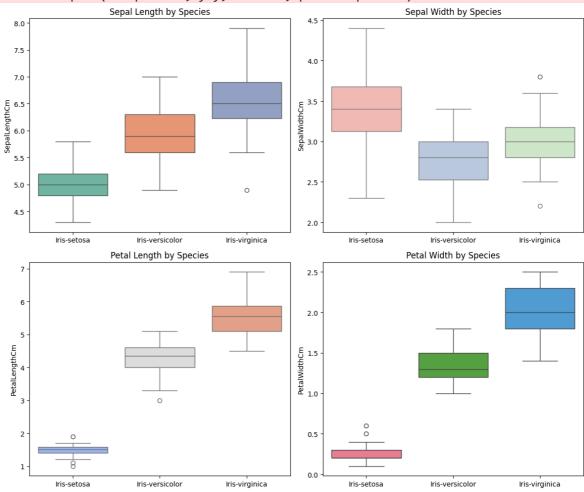
Passing `palette` without assigning `hue` is deprecated and will be removed in v 0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.

sns.boxplot(x="Species", y=y, data=df, palette=palette)

C:\Users\Adnan\AppData\Local\Temp\ipykernel_21992\2327543419.py:3: FutureWarning:

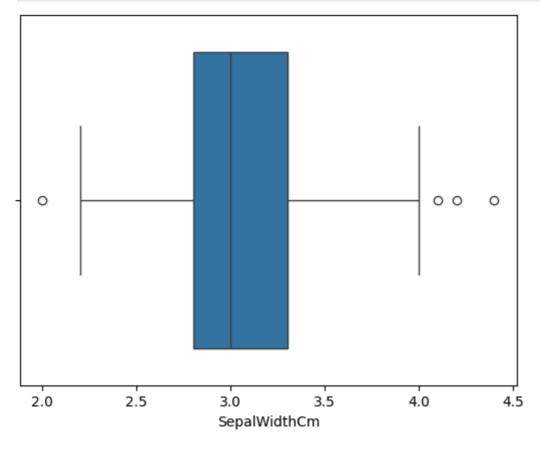
Passing `palette` without assigning `hue` is deprecated and will be removed in v 0.14.0. Assign the `x` variable to `hue` and set `legend=False` for the same effect.

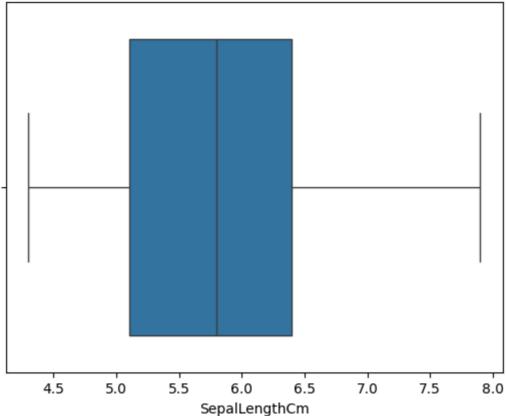
sns.boxplot(x="Species", y=y, data=df, palette=palette)

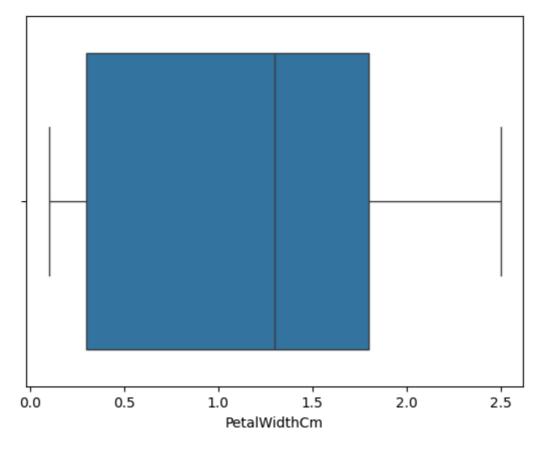


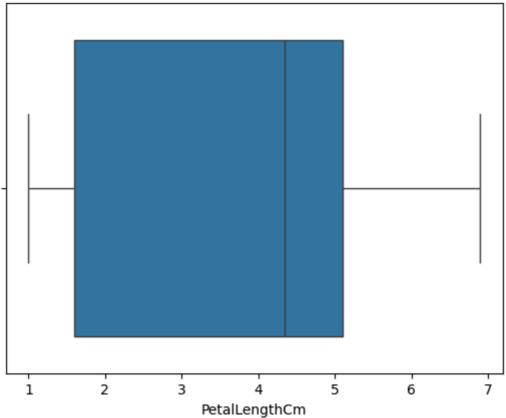
In [23]: sns.boxplot(x='SepalWidthCm', data=df)
plt.show()

```
sns.boxplot(x='SepalLengthCm', data=df)
plt.show()
sns.boxplot(x='PetalWidthCm', data=df)
plt.show()
sns.boxplot(x='PetalLengthCm', data=df)
plt.show()
```









In []:

```
import java.io.IOException;
import java.util.StringTokenizer;
import org.apache.hadoop.conf.Configuration;
import org.apache.hadoop.fs.Path;
import org.apache.hadoop.io.IntWritable;
import org.apache.hadoop.io.Text;
import org.apache.hadoop.mapreduce.Job;
import org.apache.hadoop.mapreduce.Mapper;
import org.apache.hadoop.mapreduce.Reducer;
import org.apache.hadoop.mapreduce.lib.input.FileInputFormat;
import org.apache.hadoop.mapreduce.lib.output.FileOutputFormat;
public class WordCount {
public static class TokenizerMapper
extends Mapper<Object, Text, Text, IntWritable>{
private final static IntWritable one = new IntWritable(1);
private Text word = new Text();
/*
The Mapper implementation, via the map method, processes one line at a time, as
provided by the specified TextInputFormat. It then splits the line into tokens separated by
whitespaces,
via the StringTokenizer, and emits a key-value pair of < <word>, 1>.
For the given sample input the first map emits:
< Hello, 1>
< World, 1>
< Bye, 1>
< World, 1>
*/
public void map(Object key, Text value, Context context
```

```
) throws IOException, InterruptedException {
StringTokenizer itr = new StringTokenizer(value.toString());
while (itr.hasMoreTokens()) {
word.set(itr.nextToken());
context.write(word, one);
}
}
public static class IntSumReducer
extends Reducer<Text,IntWritable,Text,IntWritable> {
private IntWritable result = new IntWritable();
public void reduce(Text key, Iterable<IntWritable> values,
Context context
) throws IOException, InterruptedException {
int sum = 0;
for (IntWritable val : values) {
sum += val.get();
}
result.set(sum);
context.write(key, result);
}
}
public static void main(String[] args) throws Exception {
Configuration conf = new Configuration();
Job job = Job.getInstance(conf, "word count");
job.setJarByClass(WordCount.class);
job.setMapperClass(TokenizerMapper.class);
job.setCombinerClass(IntSumReducer.class);
```

```
job.setReducerClass(IntSumReducer.class);
job.setOutputKeyClass(Text.class);
job.setOutputValueClass(IntWritable.class);
FileInputFormat.addInputPath(job, new Path(args[0]));
FileOutputFormat.setOutputPath(job, new Path(args[1]));
System.exit(job.waitForCompletion(true) ? 0 : 1);
}
Output:
Hello 1
Java 2
```

World 1

```
import java.io.IOException;
import org.apache.hadoop.conf.Configuration;
import org.apache.hadoop.fs.Path;
import org.apache.hadoop.io.FloatWritable;
import org.apache.hadoop.io.Text;
import org.apache.hadoop.mapreduce.Job;
import org.apache.hadoop.mapreduce.Mapper;
import org.apache.hadoop.mapreduce.Reducer;
import org.apache.hadoop.mapreduce.lib.input.TextInputFormat;
import org.apache.hadoop.mapreduce.lib.output.TextOutputFormat;
public class WeatherDataAverage {
public static class TokenizerMapper extends Mapper<Object, Text, Text, FloatWritable> {
private Text category = new Text();
private FloatWritable temperature = new FloatWritable();
private FloatWritable windSpeed = new FloatWritable();
private FloatWritable dewPoint = new FloatWritable();
public void map(Object key, Text value, Context context) throws IOException,
InterruptedException {
String[] cols = value.toString().split(" ");
float temp = Float.parseFloat(cols[0]);
float wind = Float.parseFloat(cols[1]);
float dew = Float.parseFloat(cols[2]);
category.set("Temperature");
temperature.set(temp);
context.write(category, temperature);
category.set("WindSpeed");
windSpeed.set(wind);
context.write(category, windSpeed);
```

```
category.set("DewPoint");
dewPoint.set(dew);
context.write(category, dewPoint);
}
public static class FloatAverageReducer extends Reducer<Text, FloatWritable, Text,
FloatWritable> {
private FloatWritable result = new FloatWritable();
public void reduce(Text key, Iterable<FloatWritable> values, Context context)
throws IOException, InterruptedException {
float sum = 0;
int count = 0;
for (FloatWritable a : values){sum += a.get();count++;}
float avg = sum / count;
result.set(avg);
context.write(key, result);
}
}
public static void main(String[] args) throws Exception {
Configuration conf = new Configuration();
Job job = Job.getInstance(conf, "weather data average");
job.setJarByClass(WeatherDataAverage.class);
job.setMapperClass(TokenizerMapper.class);
job.setReducerClass(FloatAverageReducer.class);
job.setMapOutputKeyClass(Text.class);
job.setMapOutputValueClass(FloatWritable.class);
job.setOutputKeyClass(Text.class);
job.setOutputValueClass(FloatWritable.class);
```

```
job.setInputFormatClass(TextInputFormat.class);
job.setOutputFormatClass(TextOutputFormat.class);
TextInputFormat.addInputPath(job, new Path(args[0]));
TextOutputFormat.setOutputPath(job, new Path(args[1]));
System.exit(job.waitForCompletion(true) ? 0 : 1);
}
Output:
DewPoint 61.94638
Temperature 54.65593
```

WindSpeed 1007.7881

```
// This opens up an interactive shell
// Create a new input1.txt file
val input = sc.textFile("passage.txt")
// Create a new RDD by splitting the input RDD on the basis of space
val words = input.flatMap(x => x.split(" "))
// Create a new RDD by mapping each word to a tuple of (word, 1)
val counts = words.map(x => (x, 1))
// Create a new RDD by reducing the tuples by key
val reducedCounts = counts.reduceByKey((x, y) => x + y)
// Save the RDD to a file
reducedCounts.saveAsTextFile("output.txt")
// Print the contents of the file
reducedCounts.foreach(println)
Output:
(Hello,1)
(Shubham,1)
```