In [802	<pre>import pandas as pd import matplotlib.pyplot as plt import seaborn as sns import numpy as np import plotly.express as px from sklearn.neighbors import train_test_split from sklearn.model_selection import train_test_split from sklearn.model_selection import train_test_split from sklearn.model_selection import KFold from sklearn.model_selection import StratifiedKFold from sklearn.model_selection import trassification_report from sklearn.model_selection import stratifiedKFold from sklearn.model_selection import cross_val_score from sklearn.model_selection import cross_val_score from sklearn.ensemble import RandomForestClassifier import warnings</pre>
Out[803	
In [804	# checking for nulls chip.info() <class 'pandas.core.frame.dataframe'=""> RangeIndex: 4854 entries, 0 to 4853 Data columns (total 14 columns): # Column Non-Null Count Otype Discount Of State Of Sta</class>
In [806	<pre>chip = chip.drop(['FP16 GFLOPS', 'FP32 GFLOPS', 'FP64 GFLOPS'], axis=1) # Replace null values with column-wise average chip = chip.fillna(chip.mean()) chip = chip.found(decimals=0).astype(object) warnings.filterwarnings("ignore") # Creating a pie chart values = chip['Foundry'].value_counts() names = values.index plt.figure(figsize=(20, 10)) patches, = plt.pie(values, labels=None, autopct='%1.1F%', startangle=90) plt.title('Foundry Distribution') plt.legend(patches, names, loc='center left', bbox_to_anchor=(1, 0.5)) plt.show()</pre> Foundry Distribution
	17.8% 17.8% 17.8% 18.60 18.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19.60 19
In [808	# Creating a bar chart plt.figure(figsize = (12,8)) ex = sns.countplot(data-chip, x="vendor", hue="Type") ex = sns.countplot(data-chip, x="vendor and Type", fontsize= 25) plt.show() Count of Chips by Vendor and Type Count of Chips by Vendor and Type Gov 1000 600 400 200 600
In [809	# Creating a scatterplot chart plt.figure(figsize=(15, 10)) ax = sns.scatterplot(data=chip, x="Transistors (million)", y="Freq (MHz)", hue="Vendor", style="Type") ax.set_title("Freq per Transistors by vendor and type", fontsize= 25) plt.show() Freq per Transistors by vendor and type Vendor
	4000
In [810	# Creating a line graph plt.figure(figsize=(12, 8)) my_chip = chip.query("Type == 'CPU'") sns.lineplot(data=my_chip, x="Process Size (nm)", y="Transistors (million)") plt.title("Transistor Count by Process Size for CPUs", fontsize= 20) plt.show() Transistor Count by Process Size for CPUs
	7000 - 6000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 7000 - 70
In [811	<pre># Creating a scatterplot chart chip['Die Size (mm^2)'] = chip['Die Size (mm^2)'].astype(float) chip['Transistors (million)'] = chip['Transistors (million)'].astype(float) plt.figure(figsize=(20, 5)) sns.regplot(x='Die Size (mm^2)', y='Transistors (million)', data=chip) plt.xlabel('Die Size (mm^2)') plt.ylabel('Transistors (million)') plt.title('Transistor Count vs Die Size', fontsize=24) plt.show()</pre>
	Transistor Count vs Die Size 10000
In [812	plt.figure(figsize(12, 6)) sns.scatterplot(data-chip.sort_values(by='Release Date'), x='Release Date', y='Transistors (million)', hue='Type') plt.title("Moore's Law by CPU/GPU", fontsize= 20) plt.show() Moore's Law by CPU/GPU ype
In [813 In [814 In [815	Release Date # Splitting the dataset into features and target X = chip[['Die Size (mm^2)', 'Transistors (million)']] y = chip["Vendor"] # Splitting the dataset into training and test sets in the ratio 70/30 X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.30, random_state = 10) # Feature Scaling \$\text{sc = StandardScaler()} \text{X_train = sc.fit_transform(X_train)} \text{X_train = sc.fit_transform(X_test)}
In [816 Out[816 In [817	<pre># Training a Random Forests classifier rf = RandomForestClassifier(n_estimators=10, criterion='gini',random_state=1) rf.fit(X_train,y_train) RandomForestClassifier(n_estimators=10, random_state=1) # Performing 5-Fold Cross Validation k = 5 kf = KFold(n_splits=k) #for other scoring parameters see here https://scikit-learn.org/stable/modules/model_evaluation.html result = cross_val_score(rf, X_train, y_train.ravel(), cv = kf, scoring='accuracy') print(f' Avg accuracy:{result.mean()}') Avg accuracy:0.9537810794420862</pre>
In [818 In [819	# Making predictions using your test data (X_test)- Random Forests y_pred_rf = rf.predict(X_test) # Random Forests - Generating Classification Report print(classification_report(y_test, y_pred_rf, target_names = ['AMD', 'NVIDIA', 'ATI', 'intel', 'Other Vnedor'])) precision recall fi-score support