

```
In [1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

```
In [2]: data=pd.read_csv(r"C:\Users\user\Downloads\sector_dataset_revised.csv")
data
```

Out[2]:

	country	country_code	sector	sector_number	year	Nominal_value_adde
0	Australia	AUS	Total	0	1975	
1	Australia	AUS	Agriculture	1	1975	
2	Australia	AUS	Mining	2	1975	
3	Australia	AUS	Manufacturing	3	1975	
4	Australia	AUS	Utilities	4	1975	
...	...	...	...	...	...	
35149	Uganda	UGA	Utilities	4	2018	4
35150	Uganda	UGA	Construction	5	2018	8
35151	Uganda	UGA	Whole sale, Accommodation and food service act...	6	2018	18
35152	Uganda	UGA	Transportation,information and communication, ...	7	2018	18
35153	Uganda	UGA	Government services, Community, social and per...	8	2018	16

35154 rows × 11 columns

```
In [3]: da=data.head(100)
da
```

Out[3]:

_number	year	Nominal_value_added_LCU	Nominal_value_added_USD	Real_value_added_LCU	Em
0	1975	76723	100440	427474	
1	1975	4291	5617	18379	
2	1975	3801	4976	16013	
3	1975	15253	19968	65863	
4	1975	2479	3246	15131	
...	...	...	...	...	
5	1985	16856	11772	42433	
6	1985	28813	20122	69052	
7	1985	63330	44228	177417	
8	1985	52131	36407	149734	
0	1986	263135	175896	602233	

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

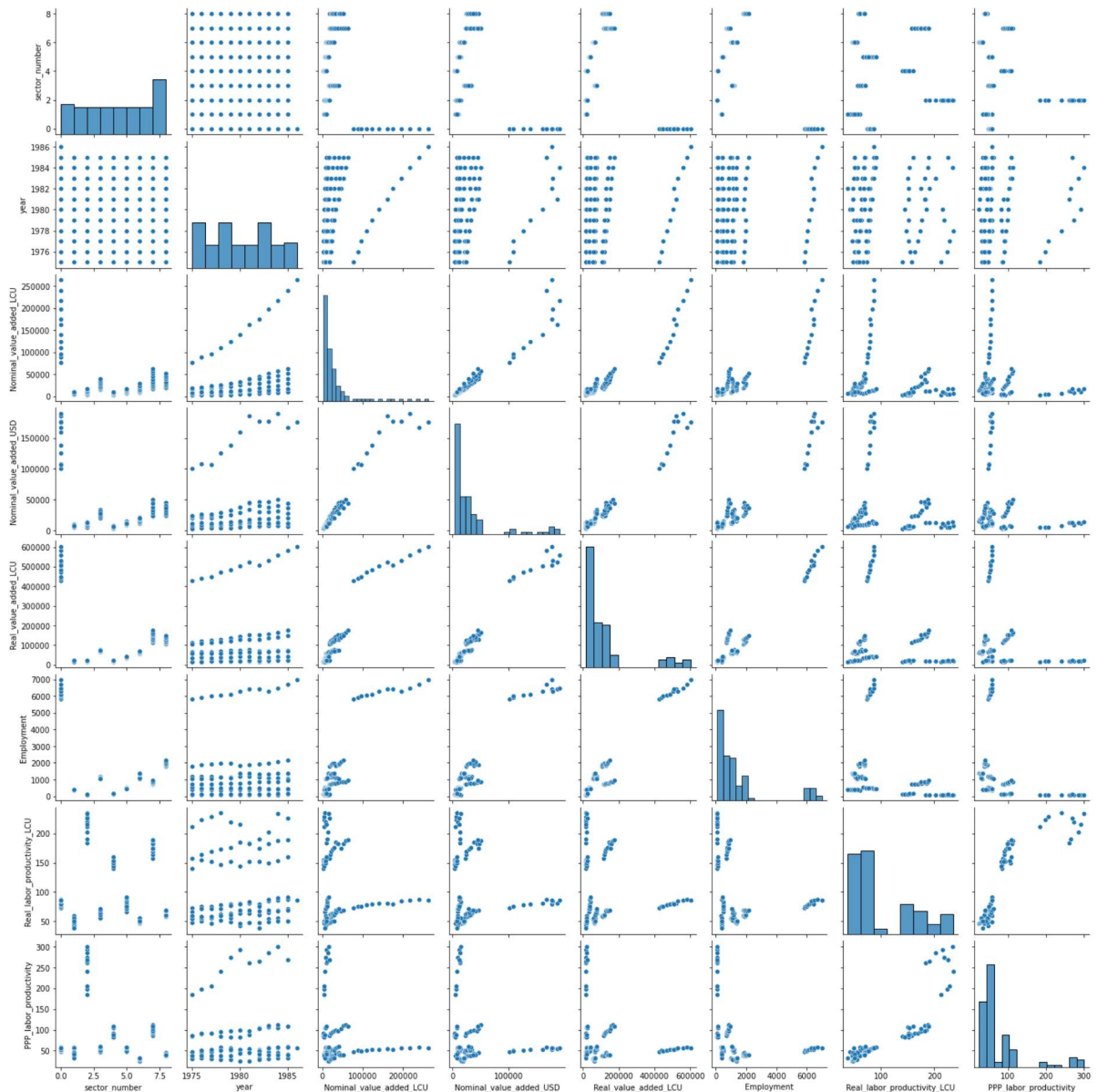
```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 35154 entries, 0 to 35153
Data columns (total 11 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   country                               35154 non-null  object
1   country_code                           35154 non-null  object
2   sector                                 35154 non-null  object
3   sector_number                           35154 non-null  int64
4   year                                   35154 non-null  int64
5   Nominal_value_added_LCU                35154 non-null  int64
6   Nominal_value_added_USD                35154 non-null  int64
7   Real_value_added_LCU                    35154 non-null  int64
8   Employment                             35154 non-null  int64
9   Real_labor_productivity_LCU            35154 non-null  int64
10  PPP_labor_productivity                  35154 non-null  int64
dtypes: int64(8), object(3)
memory usage: 3.0+ MB
```

```
In [5]: da.columns
```

```
Out[5]: Index(['country', 'country_code', 'sector', 'sector_number', 'year',
              'Nominal_value_added_LCU', 'Nominal_value_added_USD',
              'Real_value_added_LCU', 'Employment', 'Real_labor_productivity_LCU',
              'PPP_labor_productivity'],
             dtype='object')
```

```
In [6]: sns.pairplot(da)
```

```
Out[6]: <seaborn.axisgrid.PairGrid at 0x167017f2ca0>
```

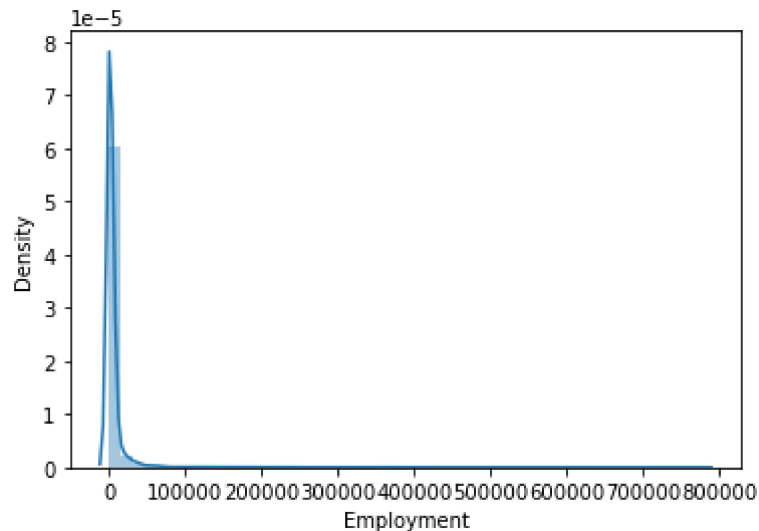


```
In [9]: sns.distplot(data["Employment"])
```

C:\Users\USER\anaconda3\lib\site-packages\seaborn\distributions.py:2619: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

```
warnings.warn(msg, FutureWarning)
```

```
Out[9]: <AxesSubplot:xlabel='Employment', ylabel='Density'>
```



## Linear Regression

```
In [10]: df=da[['sector_number', 'year',
               'Nominal_value_added_LCU', 'Nominal_value_added_USD',
               'Real_value_added_LCU', 'Employment', 'Real_labor_productivity_LCU',
               'PPP_labor_productivity']]
```

```
In [11]: x=df[['sector_number', 'year',
               'Nominal_value_added_LCU', 'Nominal_value_added_USD',
               'Real_value_added_LCU', 'Employment', 'Real_labor_productivity_LCU']]
y=df[['PPP_labor_productivity']]
```

```
In [13]: from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.3)
```

```
In [16]: from sklearn.linear_model import LinearRegression
lr=LinearRegression()
lr.fit(x_train,y_train)
```

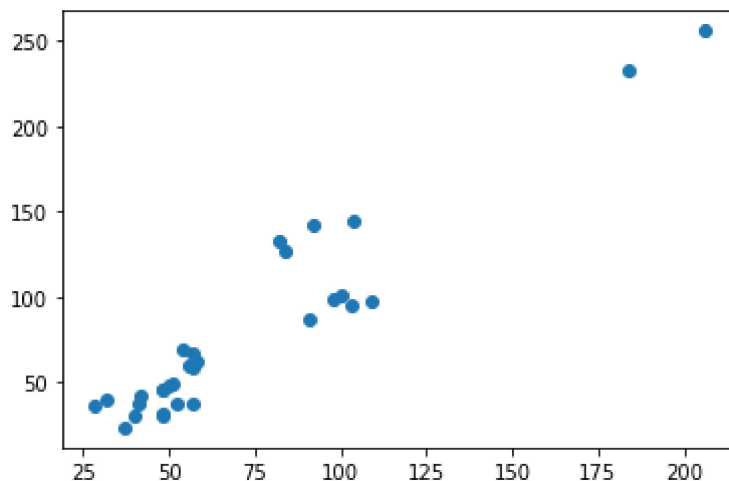
```
Out[16]: LinearRegression()
```

```
In [17]: print(lr.intercept_)
```

```
[-838.75566248]
```

```
In [18]: prediction=lr.predict(x_test)  
plt.scatter(y_test,prediction)
```

```
Out[18]: <matplotlib.collections.PathCollection at 0x1670a7bbe80>
```



```
In [19]: print(lr.score(x_test,y_test))
```

```
0.6817567228468826
```

## Ridge Regression

```
In [21]: from sklearn.linear_model import Ridge
```

```
In [22]: rr=Ridge(alpha=0)  
rr.fit(x_train,y_train)
```

```
Out[22]: Ridge(alpha=0)
```

```
In [23]: rr.score(x_test,y_test)
```

```
Out[23]: 0.6817567228468757
```

## Lasso Regression

```
In [24]: from sklearn.linear_model import Lasso
```

```
In [25]: la=Lasso(alpha=10)
         la.fit(x_train,y_train)
```

```
Out[25]: Lasso(alpha=10)
```

```
In [26]: la.score(x_test,y_test)
```

```
Out[26]: 0.6646662695472622
```

## Elastic regression

```
In [27]: from sklearn.linear_model import ElasticNet
         en=ElasticNet()
         en.fit(x_train,y_train)
```

```
Out[27]: ElasticNet()
```

```
In [28]: print(en.coef_)
```

```
[-4.20980301e+00  2.25776857e-01  4.12207102e-04  5.81578865e-04
 -1.18897592e-03  6.93498212e-02  1.38020131e+00]
```

```
In [29]: print(en.score(x_test,y_test))
```

```
0.6776894621477989
```

## Logistic Regression

```
In [55]: from sklearn.linear_model import LogisticRegression
```

```
In [56]: df=da[['sector_number', 'year',
               'Nominal_value_added_LCU', 'Nominal_value_added_USD',
               'Real_value_added_LCU', 'Employment', 'Real_labor_productivity_LCU',
               'PPP_labor_productivity']]
```

```
In [57]: feature_matrix=df.iloc[:,0:11]
         target_vector=df.iloc[:,-1]
```

```
In [58]: feature_matrix.shape
```

```
Out[58]: (100, 8)
```

```
In [59]: target_vector.shape
```

```
Out[59]: (100,)
```

```
In [60]: from sklearn.preprocessing import StandardScaler
```

```
In [61]: fs=StandardScaler().fit_transform(feature_matrix)
```

```
In [65]: logr=LogisticRegression()  
logr.fit(fs,target_vector)
```

```
Out[65]: LogisticRegression()
```

```
In [68]: observation=[[1,2,3,4,5,6,7,8]]
```

```
In [69]: predicton=logr.predict(observation)  
print(observation)
```

```
[[1, 2, 3, 4, 5, 6, 7, 8]]
```

```
In [71]: logr.classes_
```

```
Out[71]: array([ 24,  25,  28,  29,  30,  31,  32,  33,  37,  38,  39,  40,  41,  
                42,  43,  45,  46,  47,  48,  49,  50,  51,  52,  53,  54,  55,  
                56,  57,  58,  61,  82,  83,  84,  85,  86,  88,  90,  91,  92,  
                95,  98, 100, 103, 104, 106, 108, 109, 111, 112, 184, 198, 206,  
                241, 262, 265, 269, 274, 285, 293, 301], dtype=int64)
```

```
In [72]: logr.score(fs,target_vector)
```

```
Out[72]: 0.31
```

## Random Forest

```
In [74]: from sklearn.ensemble import RandomForestClassifier  
from sklearn.tree import plot_tree
```

```
In [76]: from sklearn.model_selection import train_test_split  
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.30)
```

```
In [77]: rfc=RandomForestClassifier()  
rfc.fit(x_train,y_train)
```

```
C:\Users\USER\AppData\Local\Temp\ipykernel_2068\2068597404.py:2: DataConversionWarning: A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n_samples,), for example using ravel().  
    rfc.fit(x_train,y_train)
```

```
Out[77]: RandomForestClassifier()
```

```
In [82]: parameters={'max_depth':[1,2,3,4,5],
                    'min_samples_leaf':[5,6,7,8,9,10],
                    'n_estimators':[10,20,30,40,50]}
```

```
In [83]: from sklearn.model_selection import GridSearchCV
grid_search=GridSearchCV(estimator=rfc,param_grid=parameters,cv=2,scoring='accu
grid_search.fit(x_train,y_train)
```

C:\Users\USER\anaconda3\lib\site-packages\sklearn\model\_selection\\_split.p  
y:676: UserWarning: The least populated class in y has only 1 members, whic  
h is less than n\_splits=2.

warnings.warn(

C:\Users\USER\anaconda3\lib\site-packages\sklearn\model\_selection\\_validati  
on.py:680: DataConversionWarning: A column-vector y was passed when a 1d ar  
ray was expected. Please change the shape of y to (n\_samples,), for example  
using ravel().

estimator.fit(X\_train, y\_train, \*\*fit\_params)

C:\Users\USER\anaconda3\lib\site-packages\sklearn\model\_selection\\_validati  
on.py:680: DataConversionWarning: A column-vector y was passed when a 1d ar  
ray was expected. Please change the shape of y to (n\_samples,), for example  
using ravel().

estimator.fit(X\_train, y\_train, \*\*fit\_params)

C:\Users\USER\anaconda3\lib\site-packages\sklearn\model\_selection\\_validati  
on.py:680: DataConversionWarning: A column-vector y was passed when a 1d ar  
ray was expected. Please change the shape of y to (n\_samples,), for example  
using ravel().

estimator.fit(X\_train, y\_train, \*\*fit\_params)

```
In [84]: grid_search.best_score_
```

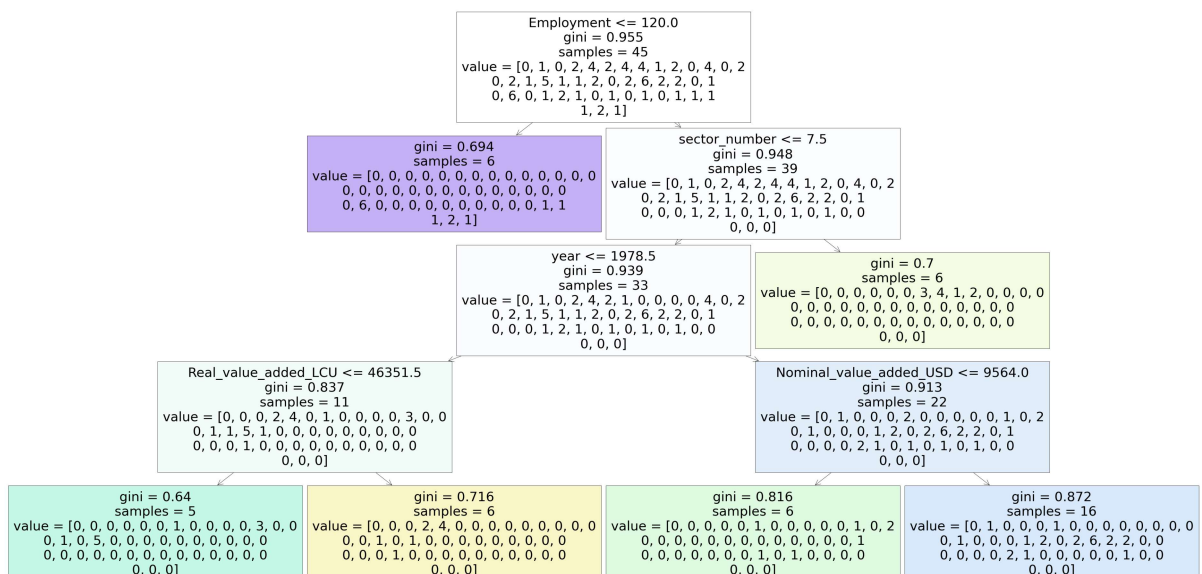
```
Out[84]: 0.12857142857142856
```

```
In [85]: rfc_best=grid_search.best_estimator_
```



```
In [86]: from sklearn.tree import plot_tree
plt.figure(figsize=(80,40))
plot_tree(rfc_best.estimators_[5],feature_names=x.columns,filled=True)
```

```
Out[86]: [Text(0.5, 0.9, 'Employment <= 120.0\ngini = 0.955\nsamples = 45\nvalue = [0,
1, 0, 2, 4, 2, 4, 4, 1, 2, 0, 4, 0, 2\n0, 2, 1, 5, 1, 1, 2, 0, 2, 6, 2, 2, 0,
1\n0, 6, 0, 1, 2, 1, 0, 1, 0, 1, 0, 1, 1, 1\n1, 2, 1]'),
Text(0.375, 0.7, 'gini = 0.694\nsamples = 6\nvalue = [0, 0, 0, 0, 0, 0, 0,
0, 0, 0, 0, 0, 0\n0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0\n0, 6, 0, 0, 0,
0, 0, 0, 0, 0, 0, 0, 1, 1\n1, 2, 1]'),
Text(0.625, 0.7, 'sector_number <= 7.5\ngini = 0.948\nsamples = 39\nvalue =
[0, 1, 0, 2, 4, 2, 4, 4, 1, 2, 0, 4, 0, 2\n0, 2, 1, 5, 1, 1, 2, 0, 2, 6, 2,
2, 0, 1\n0, 0, 0, 1, 2, 1, 0, 1, 0, 1, 0, 1, 0, 0\n0, 0, 0]'),
Text(0.5, 0.5, 'year <= 1978.5\ngini = 0.939\nsamples = 33\nvalue = [0, 1,
0, 2, 4, 2, 1, 0, 0, 0, 0, 4, 0, 2\n0, 2, 1, 5, 1, 1, 2, 0, 2, 6, 2, 2, 0, 1
\n0, 0, 0, 1, 2, 1, 0, 1, 0, 1, 0, 1, 0, 0\n0, 0, 0]'),
Text(0.25, 0.3, 'Real_value_added_LCU <= 46351.5\ngini = 0.837\nsamples = 11
\nvalue = [0, 0, 0, 2, 4, 0, 1, 0, 0, 0, 0, 3, 0, 0\n0, 1, 1, 5, 1, 0, 0, 0,
0, 0, 0, 0, 0, 0\n0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0\n0, 0, 0]'),
Text(0.125, 0.1, 'gini = 0.64\nsamples = 5\nvalue = [0, 0, 0, 0, 0, 0, 1, 0,
0, 0, 0, 3, 0, 0\n0, 1, 0, 5, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0\n0, 0, 0, 0, 0, 0,
0, 0, 0, 0, 0, 0, 0\n0, 0, 0]'),
Text(0.375, 0.1, 'gini = 0.716\nsamples = 6\nvalue = [0, 0, 0, 2, 4, 0, 0,
0, 0, 0, 0, 0, 0\n0, 0, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0\n0, 0, 0, 1, 0,
0, 0, 0, 0, 0, 0, 0, 0\n0, 0, 0]'),
Text(0.75, 0.3, 'Nominal_value_added_USD <= 9564.0\ngini = 0.913\nsamples =
22\nvalue = [0, 1, 0, 0, 0, 2, 0, 0, 0, 0, 0, 1, 0, 2\n0, 1, 0, 0, 0, 1, 2,
0, 2, 6, 2, 2, 0, 1\n0, 0, 0, 0, 2, 1, 0, 1, 0, 1, 0, 1, 0, 0\n0, 0, 0]'),
Text(0.625, 0.1, 'gini = 0.816\nsamples = 6\nvalue = [0, 0, 0, 0, 0, 1, 0,
0, 0, 0, 0, 1, 0, 2\n0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1\n0, 0, 0, 0, 0,
0, 0, 1, 0, 1, 0, 0, 0, 0\n0, 0, 0]'),
Text(0.875, 0.1, 'gini = 0.872\nsamples = 16\nvalue = [0, 1, 0, 0, 0, 1, 0,
0, 0, 0, 0, 0, 0\n0, 1, 0, 0, 0, 1, 2, 0, 2, 6, 2, 2, 0, 0\n0, 0, 0, 0, 2,
1, 0, 0, 0, 0, 0, 1, 0, 0\n0, 0, 0]'),
Text(0.75, 0.5, 'gini = 0.7\nsamples = 6\nvalue = [0, 0, 0, 0, 0, 0, 3, 4,
1, 2, 0, 0, 0, 0\n0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0\n0, 0, 0, 0, 0, 0, 0,
0, 0, 0, 0, 0, 0\n0, 0, 0]')]
```



In [ ]: