

# RAINFALL IN INDIA 1901-2015

## PROBLEM STATEMENT:-

To perform an analytics report on 100 years of Rainfall data  
Which Subdivision getting more rainfall

In [1]:

```
1 import numpy as np
2 import pandas as pd
3 import warnings
4 warnings.simplefilter(action='ignore')
```

In [2]:

```
1 df = pd.read_csv(r"C:\Users\HP\OneDrive\Desktop\revathi\rainfall in india 1901-2015.csv")
2 df
```

Out[2]:

	SUBDIVISION	YEAR	JAN	FEB	MAR	APR	MAY	JUN	JUL	AUG	SEP	OCT	NOV	DEC	ANNUAL	Jan-Feb	Mar-May	Jun-Sep
0	ANDAMAN & NICOBAR ISLANDS	1901	49.2	87.1	29.2	2.3	528.8	517.5	365.1	481.1	332.6	388.5	558.2	33.6	3373.2	136.3	560.3	1696.3
1	ANDAMAN & NICOBAR ISLANDS	1902	0.0	159.8	12.2	0.0	446.1	537.1	228.9	753.7	666.2	197.2	359.0	160.5	3520.7	159.8	458.3	2185.9
2	ANDAMAN & NICOBAR ISLANDS	1903	12.7	144.0	0.0	1.0	235.1	479.9	728.4	326.7	339.0	181.2	284.4	225.0	2957.4	156.7	236.1	1874.0
3	ANDAMAN & NICOBAR ISLANDS	1904	9.4	14.7	0.0	202.4	304.5	495.1	502.0	160.1	820.4	222.2	308.7	40.1	3079.6	24.1	506.9	1977.6
4	ANDAMAN & NICOBAR ISLANDS	1905	1.3	0.0	3.3	26.9	279.5	628.7	368.7	330.5	297.0	260.7	25.4	344.7	2566.7	1.3	309.7	1624.9
...	...	...	...	...	...	...	...	...	...	...	...	...	...	...	...	...	...	...
4111	LAKSHADWEEP	2011	5.1	2.8	3.1	85.9	107.2	153.6	350.2	254.0	255.2	117.4	184.3	14.9	1533.7	7.9	196.2	1013.0
4112	LAKSHADWEEP	2012	19.2	0.1	1.6	76.8	21.2	327.0	231.5	381.2	179.8	145.9	12.4	8.8	1405.5	19.3	99.6	1119.5
4113	LAKSHADWEEP	2013	26.2	34.4	37.5	5.3	88.3	426.2	296.4	154.4	180.0	72.8	78.1	26.7	1426.3	60.6	131.1	1057.0
4114	LAKSHADWEEP	2014	53.2	16.1	4.4	14.9	57.4	244.1	116.1	466.1	132.2	169.2	59.0	62.3	1395.0	69.3	76.7	958.5
4115	LAKSHADWEEP	2015	2.2	0.5	3.7	87.1	133.1	296.6	257.5	146.4	160.4	165.4	231.0	159.0	1642.9	2.7	223.9	860.9

4116 rows × 19 columns

In [3]:

```
1 df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 4116 entries, 0 to 4115
Data columns (total 19 columns):
#   Column      Non-Null Count  Dtype
---  -
0   SUBDIVISION 4116 non-null   object
1   YEAR        4116 non-null   int64
2   JAN         4112 non-null   float64
3   FEB         4113 non-null   float64
4   MAR         4110 non-null   float64
5   APR         4112 non-null   float64
6   MAY         4113 non-null   float64
7   JUN         4111 non-null   float64
8   JUL         4109 non-null   float64
9   AUG         4112 non-null   float64
10  SEP         4110 non-null   float64
11  OCT         4109 non-null   float64
12  NOV         4105 non-null   float64
13  DEC         4106 non-null   float64
14  ANNUAL       4090 non-null   float64
15  Jan-Feb     4110 non-null   float64
16  Mar-May     4107 non-null   float64
17  Jun-Sep     4106 non-null   float64
18  Oct-Dec     4103 non-null   float64
dtypes: float64(17), int64(1), object(1)
memory usage: 611.1+ KB
```

```
In [4]: 1 df.isnull().sum()
```

```
Out[4]: SUBDIVISION    0
YEAR                0
JAN                 4
FEB                 3
MAR                 6
APR                 4
MAY                 3
JUN                 5
JUL                 7
AUG                 4
SEP                 6
OCT                 7
NOV                11
DEC                10
ANNUAL             26
Jan-Feb             6
Mar-May             9
Jun-Sep            10
Oct-Dec            13
dtype: int64
```

```
In [5]: 1 df.fillna(method='ffill',inplace=True)
```

```
In [6]: 1 df.isnull().sum()
```

```
Out[6]: SUBDIVISION    0
YEAR                0
JAN                 0
FEB                 0
MAR                 0
APR                 0
MAY                 0
JUN                 0
JUL                 0
AUG                 0
SEP                 0
OCT                 0
NOV                 0
DEC                 0
ANNUAL              0
Jan-Feb             0
Mar-May             0
Jun-Sep             0
Oct-Dec             0
dtype: int64
```

```
In [7]: df['SUBDIVISION'].value_counts()
```

Out[7]:

SUBDIVISION	
WEST MADHYA PRADESH	115
EAST RAJASTHAN	115
COASTAL KARNATAKA	115
TAMIL NADU	115
RAYALSEEMA	115
TELANGANA	115
COASTAL ANDHRA PRADESH	115
CHHATTISGARH	115
VIDARBHA	115
MATATHWADA	115
MADHYA MAHARASHTRA	115
KONKAN & GOA	115
SAURASHTRA & KUTCH	115
GUJARAT REGION	115
EAST MADHYA PRADESH	115
KERALA	115
WEST RAJASTHAN	115
SOUTH INTERIOR KARNATAKA	115
JAMMU & KASHMIR	115
HIMACHAL PRADESH	115
PUNJAB	115
HARYANA DELHI & CHANDIGARH	115
UTTARAKHAND	115
WEST UTTAR PRADESH	115
EAST UTTAR PRADESH	115
BIHAR	115
JHARKHAND	115
ORISSA	115
GANGETIC WEST BENGAL	115
SUB HIMALAYAN WEST BENGAL & SIKKIM	115
NAGA MANI MIZO TRIPURA	115
ASSAM & MEGHALAYA	115
NORTH INTERIOR KARNATAKA	115
LAKSHADWEEP	114
ANDAMAN & NICOBAR ISLANDS	110
ARUNACHAL PRADESH	97
Name: count, dtype: int64	

In [8]:

```
1 s = {'SUBDIVISION':{'WEST MADHYA PRADESH':1,
2           'EAST RAJASTHAN':2,
3           'COASTAL KARNATAKA':3,
4           'TAMIL NADU':4,
5           'RAYALSEEMA':5,
6           'TELANGANA':6,
7           'COASTAL ANDHRA PRADESH':7,
8           'CHHATTISGARH':8,
9           'VIDARBHA':9,
10          'MATATHWADA':10,
11          'MADHYA MAHARASHTRA':11,
12          'KONKAN & GOA':12,
13          'SAURASHTRA & KUTCH':13,
14          'GUJARAT REGION':14,
15          'EAST MADHYA PRADESH':15,
16          'KERALA':16,
17          'WEST RAJASTHAN':17,
18          'SOUTH INTERIOR KARNATAKA':18,
19          'JAMMU & KASHMIR':19,
20          'HIMACHAL PRADESH':20,
21          'PUNJAB':21,
22          'HARYANA DELHI & CHANDIGARH':22,
23          'UTTARAKHAND':23,
24          'WEST UTTAR PRADESH':24,
25          'EAST UTTAR PRADESH':25,
26          'BIHAR':26,
27          'JHARKHAND':27,
28          'ORISSA':28,
29          'GANGETIC WEST BENGAL':29,
30          'SUB HIMALAYAN WEST BENGAL & SIKKIM':30,
31          'NAGA MANI MIZO TRIPURA':31,
32          'ASSAM & MEGHALAYA':32,
33          'NORTH INTERIOR KARNATAKA':33,
34          'LAKSHADWEEP':34,
35          'ANDAMAN & NICOBAR ISLANDS':35,
36          'ARUNACHAL PRADESH':36}}
37 df = df.replace(s)
38 df
```

Out[8]:

	SUBDIVISION	YEAR	JAN	FEB	MAR	APR	MAY	JUN	JUL	AUG	SEP	OCT	NOV	DEC	ANNUAL	Jan-Feb	Mar-May	Jun-Sep
0	35	1901	49.2	87.1	29.2	2.3	528.8	517.5	365.1	481.1	332.6	388.5	558.2	33.6	3373.2	136.3	560.3	1696.3
1	35	1902	0.0	159.8	12.2	0.0	446.1	537.1	228.9	753.7	666.2	197.2	359.0	160.5	3520.7	159.8	458.3	2185.9
2	35	1903	12.7	144.0	0.0	1.0	235.1	479.9	728.4	326.7	339.0	181.2	284.4	225.0	2957.4	156.7	236.1	1874.0
3	35	1904	9.4	14.7	0.0	202.4	304.5	495.1	502.0	160.1	820.4	222.2	308.7	40.1	3079.6	24.1	506.9	1977.6
4	35	1905	1.3	0.0	3.3	26.9	279.5	628.7	368.7	330.5	297.0	260.7	25.4	344.7	2566.7	1.3	309.7	1624.9
...	...	...	...	...	...	...	...	...	...	...	...	...	...	...	...	...	...	...
4111	34	2011	5.1	2.8	3.1	85.9	107.2	153.6	350.2	254.0	255.2	117.4	184.3	14.9	1533.7	7.9	196.2	1013.0
4112	34	2012	19.2	0.1	1.6	76.8	21.2	327.0	231.5	381.2	179.8	145.9	12.4	8.8	1405.5	19.3	99.6	1119.5
4113	34	2013	26.2	34.4	37.5	5.3	88.3	426.2	296.4	154.4	180.0	72.8	78.1	26.7	1426.3	60.6	131.1	1057.0
4114	34	2014	53.2	16.1	4.4	14.9	57.4	244.1	116.1	466.1	132.2	169.2	59.0	62.3	1395.0	69.3	76.7	958.5
4115	34	2015	2.2	0.5	3.7	87.1	133.1	296.6	257.5	146.4	160.4	165.4	231.0	159.0	1642.9	2.7	223.9	860.9

4116 rows × 19 columns

In [9]:

```
1 df.columns
```

Out[9]:

Index(['SUBDIVISION', 'YEAR', 'JAN', 'FEB', 'MAR', 'APR', 'MAY', 'JUN', 'JUL',  
 'AUG', 'SEP', 'OCT', 'NOV', 'DEC', 'ANNUAL', 'Jan-Feb', 'Mar-May',  
 'Jun-Sep', 'Oct-Dec'],  
 dtype='object')

In [10]:

```
1 x=df[['YEAR', 'JAN', 'FEB', 'MAR', 'APR', 'MAY', 'JUN', 'JUL',  
2       'AUG', 'SEP', 'OCT', 'NOV', 'DEC', 'Jan-Feb', 'Mar-May',  
3       'Jun-Sep', 'Oct-Dec', 'ANNUAL']]  
4 y=df[['SUBDIVISION']]
```

In [11]:

```
1 from sklearn.model_selection import train_test_split
```

In [12]:

```
1 x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.3)  
2 x_train.shape,x_test.shape
```

Out[12]:

((2881, 18), (1235, 18))

In [13]: 1 **from** sklearn.linear\_model **import** LinearRegression

In [14]: 1 lr = LinearRegression()  
2 lr.fit(x\_train,y\_train)  
3 **print**(lr.score(x\_test,y\_test))  
4 **print**(lr.score(x\_train,y\_train))

0.3350430369928301  
0.3477307578496588

In [15]: 1 **from** sklearn.linear\_model **import** Lasso,Ridge,LassoCV,RidgeCV

In [16]: 1 lasso = Lasso(alpha=10)  
2 lasso.fit(x\_train,y\_train)  
3 **print**(lasso.score(x\_test,y\_test))  
4 **print**(lasso.score(x\_train,y\_train))

0.3363056150705481  
0.3430519877004349

In [17]: 1 ridge = Ridge(alpha=10)  
2 ridge.fit(x\_train,y\_train)  
3 **print**(ridge.score(x\_test,y\_test))  
4 **print**(ridge.score(x\_train,y\_train))

0.3350439248441428  
0.34773074194053677

In [18]: 1 lasso\_cv = LassoCV(alphas=[10,20,30,40,50])  
2 lasso\_cv.fit(x\_train,y\_train)  
3 **print**(lasso\_cv.score(x\_test,y\_test))  
4 **print**(lasso\_cv.score(x\_train,y\_train))

0.3363056150705481  
0.3430519877004349

In [19]: 1 ridge\_cv = RidgeCV(alphas=[10,20,30,40,50])  
2 ridge\_cv.fit(x\_train,y\_train)  
3 **print**(ridge\_cv.score(x\_test,y\_test))  
4 **print**(ridge\_cv.score(x\_train,y\_train))

0.33504733213082327  
0.34773038438019777

In [20]: 1 **from** sklearn.linear\_model **import** ElasticNet

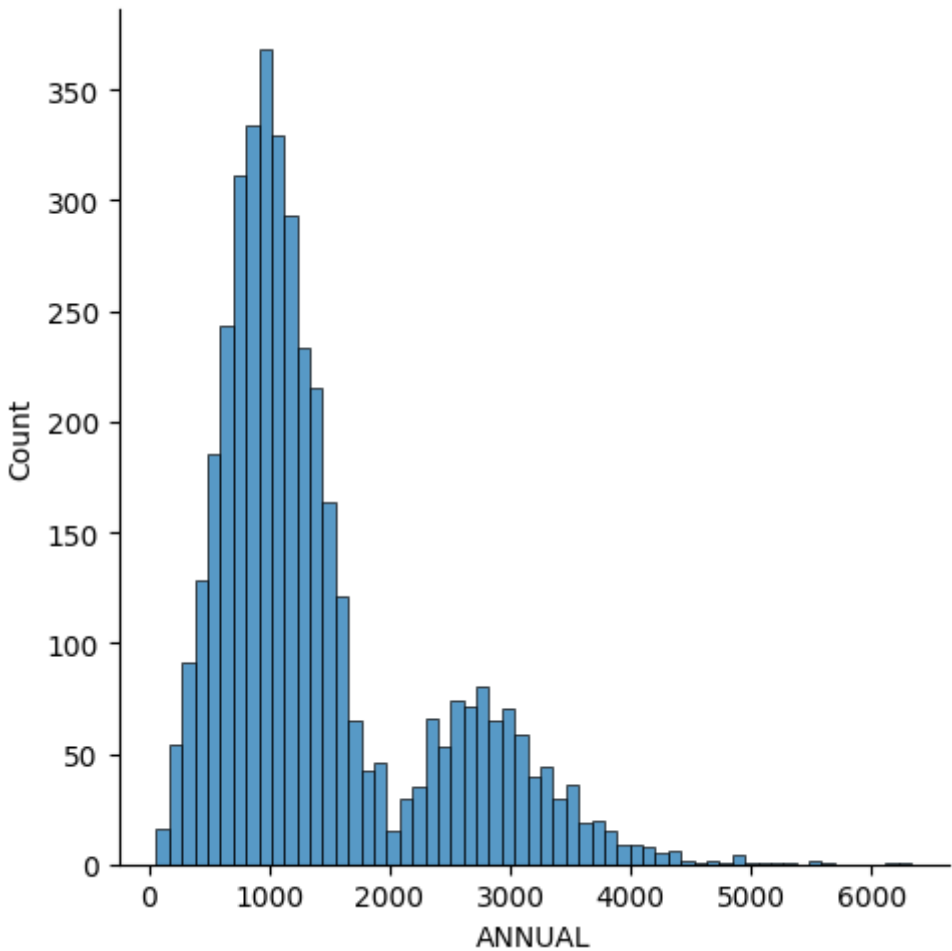
In [21]: 1 en = ElasticNet()  
2 en.fit(x\_train,y\_train)  
3 **print**(en.score(x\_train,y\_train))  
4 **print**(en.score(x\_test,y\_test))

0.34738681609723476  
0.33550251703102685

In [22]:

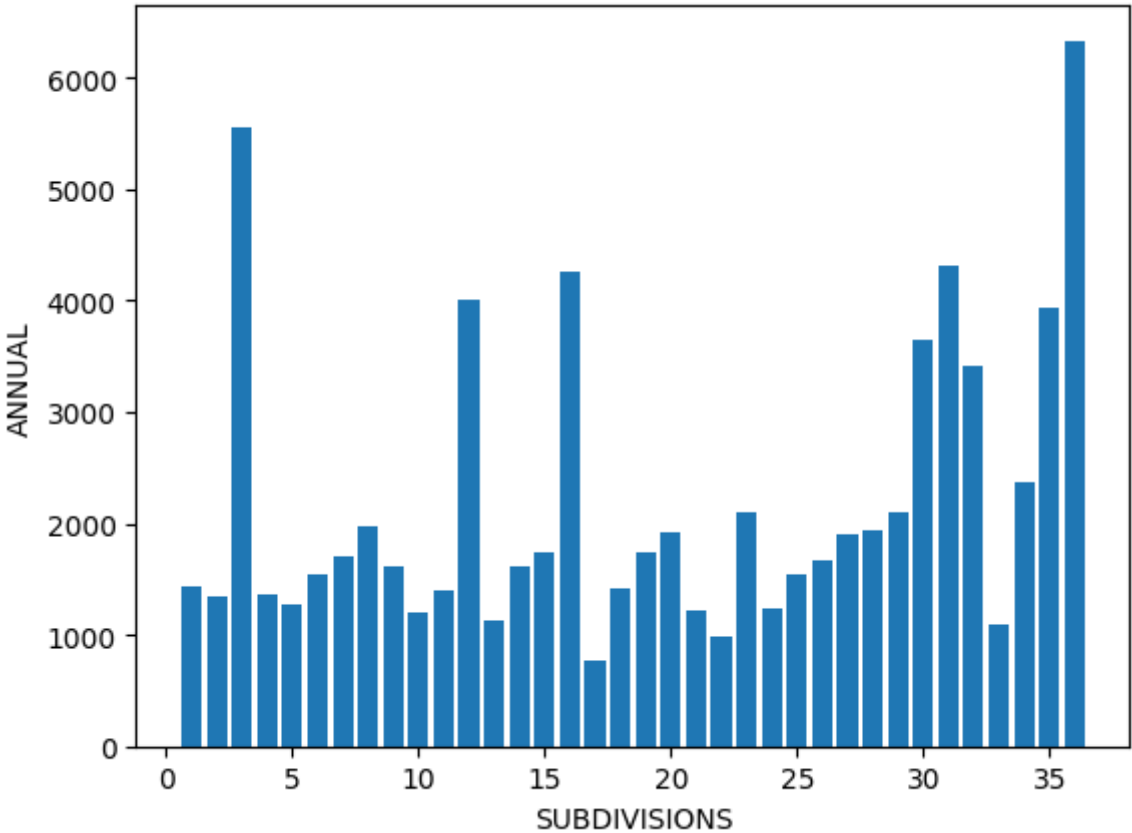
```
1 import seaborn as sns
2 sns.displot(df[ 'ANNUAL '])
```

Out[22]: <seaborn.axisgrid.FacetGrid at 0x16f4ca29e70>



In [23]:

```
1 import matplotlib.pyplot as plt
2 import seaborn as sns
3 plt.bar(df[ 'SUBDIVISION'],df[ 'ANNUAL '])
4 plt.xlabel( 'SUBDIVISIONS')
5 plt.ylabel( 'ANNUAL')
6 plt.show()
```



CONCLUSION:-

According to this dataset Andaman & nicobar was noted highest rain fall in annually

District wise RainFall Normal

Problem Statement:-

Which state has more rainfall in annually

In [38]:

1 import pandas as pd  
2 import numpy as np

In [39]:

1 data =pd.read\_csv(r"C:\Users\HP\OneDrive\Desktop\revathi\district wise rainfall normal.csv")  
2 data

Out[39]:

NAME	DISTRICT	JAN	FEB	MAR	APR	MAY	JUN	JUL	AUG	SEP	OCT	NOV	DEC	ANNUAL	Jan-Feb	Mar-May	Jun-Sep	Oct-Dec
Andaman and Nicobar Islands	NICOBAR	107.3	57.9	65.2	117.0	358.5	295.5	285.0	271.9	354.8	326.0	315.2	250.9	2805.2	165.2	540.7	1207.2	85.0
Andaman and Nicobar Islands	SOUTH ANDAMAN	43.7	26.0	18.6	90.5	374.4	457.2	421.3	423.1	455.6	301.2	275.8	128.3	3015.7	69.7	483.5	1757.2	70.0
Andaman and Nicobar Islands	N & MANDAMAN	32.7	15.9	8.6	53.4	343.6	503.3	465.4	460.9	454.8	276.1	198.6	100.0	2913.3	48.6	405.6	1884.4	57.0
ARUNACHAL PRADESH	LOHIT	42.2	80.8	176.4	358.5	306.4	447.0	660.1	427.8	313.6	167.1	34.1	29.8	3043.8	123.0	841.3	1848.5	23.0
ARUNACHAL PRADESH	EAST SIANG	33.3	79.5	105.9	216.5	323.0	738.3	990.9	711.2	568.0	206.9	29.5	31.7	4034.7	112.8	645.4	3008.4	26.0
...	...	...	...	...	...	...	...	...	...	...	...	...	...	...	...	...	...	...
KERALA	IDUKKI	13.4	22.1	43.6	150.4	232.6	651.6	788.9	527.3	308.4	343.2	172.9	48.1	3302.5	35.5	426.6	2276.2	56.0
KERALA	KASARGOD	2.3	1.0	8.4	46.9	217.6	999.6	1108.5	636.3	263.1	234.9	84.6	18.4	3621.6	3.3	272.9	3007.5	33.0
KERALA	PATHANAMTHITTA	19.8	45.2	73.9	184.9	294.7	556.9	539.9	352.7	266.2	359.4	213.5	51.3	2958.4	65.0	553.5	1715.7	62.0
KERALA	WAYANAD	4.8	8.3	17.5	83.3	174.6	698.1	1110.4	592.9	230.7	213.1	93.6	25.8	3253.1	13.1	275.4	2632.1	33.0
MIZORAM	LAKSHADWEEP	20.8	14.7	11.8	48.9	171.7	330.2	287.7	217.5	163.1	157.1	117.7	58.8	1600.0	35.5	232.4	998.5	33.0

Columns

In [40]:

1 data.info()

```
<class 'pandas.core.frame.DataFrame'>  
RangeIndex: 641 entries, 0 to 640  
Data columns (total 19 columns):  
#   Column          Non-Null Count  Dtype  
---  -  
0   STATE_UT_NAME   641 non-null    object  
1   DISTRICT        641 non-null    object  
2   JAN             641 non-null    float64  
3   FEB             641 non-null    float64  
4   MAR             641 non-null    float64  
5   APR             641 non-null    float64  
6   MAY             641 non-null    float64  
7   JUN             641 non-null    float64  
8   JUL             641 non-null    float64  
9   AUG             641 non-null    float64  
10  SEP             641 non-null    float64  
11  OCT             641 non-null    float64  
12  NOV             641 non-null    float64  
13  DEC             641 non-null    float64  
14  ANNUAL          641 non-null    float64  
15  Jan-Feb        641 non-null    float64  
16  Mar-May        641 non-null    float64  
17  Jun-Sep        641 non-null    float64  
18  Oct-Dec        641 non-null    float64  
dtypes: float64(17), object(2)  
memory usage: 95.3+ KB
```

In [41]:

▶

1 data.isnull().sum()

Out[41]: STATE\_UT\_NAME 0  
DISTRICT 0  
JAN 0  
FEB 0  
MAR 0  
APR 0  
MAY 0  
JUN 0  
JUL 0  
AUG 0  
SEP 0  
OCT 0  
NOV 0  
DEC 0  
ANNUAL 0  
Jan-Feb 0  
Mar-May 0  
Jun-Sep 0  
Oct-Dec 0  
dtype: int64

In [42]:

▶

1 data.columns

Out[42]: Index(['STATE\_UT\_NAME', 'DISTRICT', 'JAN', 'FEB', 'MAR', 'APR', 'MAY', 'JUN',  
'JUL', 'AUG', 'SEP', 'OCT', 'NOV', 'DEC', 'ANNUAL', 'Jan-Feb',  
'Mar-May', 'Jun-Sep', 'Oct-Dec'],  
dtype='object')

In [43]:

▶

1 data['STATE\_UT\_NAME'].value\_counts()

Out[43]: STATE\_UT\_NAME  
UTTAR PRADESH 71  
MADHYA PRADESH 50  
BIHAR 38  
MAHARASHTRA 35  
RAJASTHAN 33  
TAMIL NADU 32  
KARNATAKA 30  
ORISSA 30  
ASSAM 27  
GUJARAT 26  
JHARKHAND 24  
ANDHRA PRADESH 23  
JAMMU AND KASHMIR 22  
HARYANA 21  
PUNJAB 20  
WEST BENGAL 19  
CHATISGARH 18  
ARUNACHAL PRADESH 16  
KERALA 14  
UTTARANCHAL 13  
HIMACHAL 12  
NAGALAND 11  
MIZORAM 9  
MANIPUR 9  
DELHI 9  
MEGHALAYA 7  
SIKKIM 4  
TRIPURA 4  
PONDICHERRY 4  
ANDAMAN And NICOBAR ISLANDS 3  
GOA 2  
DAMAN AND DUI 2  
DADAR NAGAR HAVELI 1  
CHANDIGARH 1  
LAKSHADWEEP 1  
Name: count, dtype: int64



In [44]:

```
1 t = {'STATE_UT_NAME':{'UTTAR PRADESH':1,
2                               'MADHYA PRADESH':2,
3                               'BIHAR':3,
4                               'MAHARASHTRA':4,
5                               'RAJASTHAN':5,
6                               'TAMIL NADU':6,
7                               'KARNATAKA':7,
8                               'ORISSA':8,
9                               'ASSAM':9,
10                              'GUJARAT':10,
11                              'JHARKHAND':11,
12                              'ANDHRA PRADESH':12,
13                              'JAMMU AND KASHMIR':13,
14                              'HARYANA':14,
15                              'PUNJAB':15,
16                              'WEST BENGAL':16,
17                              'CHATISGARH':17,
18                              'ARUNACHAL PRADESH':18,
19                              'KERALA':19,
20                              'UTTARANCHAL':20,
21                              'HIMACHAL':21,
22                              'NAGALAND':22,
23                              'MIZORAM':23,
24                              'MANIPUR':24,
25                              'DELHI':25,
26                              'MEGHALAYA':26,
27                              'SIKKIM':27,
28                              'TRIPURA':28,
29                              'PONDICHERRY':29,
30                              'ANDAMAN And NICOBAR ISLANDS':30,
31                              'GOA':31,
32                              'DAMAN AND DIU':32,
33                              'DADAR NAGAR HAVELI':33,
34                              'CHANDIGARH':34,
35                              'LAKSHADWEEP':35}}
36 data = data.replace(t)
37 data
```

Out[44]:

	STATE_UT_NAME	DISTRICT	JAN	FEB	MAR	APR	MAY	JUN	JUL	AUG	SEP	OCT	NOV	DEC	ANNUAL	Jan-Feb
0	30	NICOBAR	107.3	57.9	65.2	117.0	358.5	295.5	285.0	271.9	354.8	326.0	315.2	250.9	2805.2	165.2
1	30	SOUTH ANDAMAN	43.7	26.0	18.6	90.5	374.4	457.2	421.3	423.1	455.6	301.2	275.8	128.3	3015.7	69.7
2	30	N & M ANDAMAN	32.7	15.9	8.6	53.4	343.6	503.3	465.4	460.9	454.8	276.1	198.6	100.0	2913.3	48.6
3	18	LOHIT	42.2	80.8	176.4	358.5	306.4	447.0	660.1	427.8	313.6	167.1	34.1	29.8	3043.8	123.0
4	18	EAST SIANG	33.3	79.5	105.9	216.5	323.0	738.3	990.9	711.2	568.0	206.9	29.5	31.7	4034.7	112.8
...	...	...	...	...	...	...	...	...	...	...	...	...	...	...	...	...
636	19	IDUKKI	13.4	22.1	43.6	150.4	232.6	651.6	788.9	527.3	308.4	343.2	172.9	48.1	3302.5	35.5
637	19	KASARGOD	2.3	1.0	8.4	46.9	217.6	999.6	1108.5	636.3	263.1	234.9	84.6	18.4	3621.6	3.3
638	19	PATHANAMTHITTA	19.8	45.2	73.9	184.9	294.7	556.9	539.9	352.7	266.2	359.4	213.5	51.3	2958.4	65.0
639	19	WAYANAD	4.8	8.3	17.5	83.3	174.6	698.1	1110.4	592.9	230.7	213.1	93.6	25.8	3253.1	13.1
640	35	LAKSHADWEEP	20.8	14.7	11.8	48.9	171.7	330.2	287.7	217.5	163.1	157.1	117.7	58.8	1600.0	35.5

641 rows × 19 columns

In [45]:

```
1 data['DISTRICT'].value_counts()
```

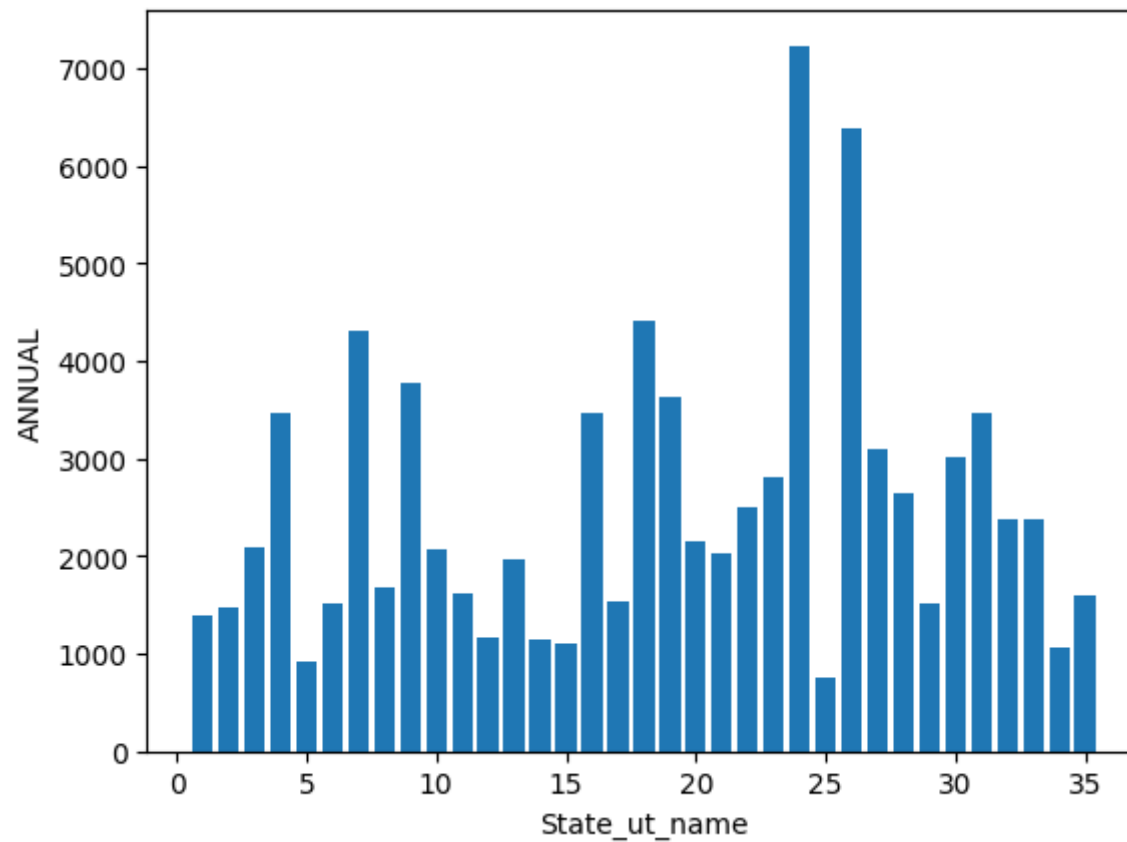
Out[45]:

```
DISTRICT
BIJAPUR      2
BILASPUR     2
AURANGABAD   2
HAMIRPUR     2
NICOBAR       1
..
GONDA        1
GORAKHPUR    1
HARDOI       1
JAUNPUR      1
LAKSHADWEEP  1
Name: count, Length: 637, dtype: int64
```

In [46]:

```
1 import matplotlib.pyplot as plt
2 import seaborn as sns
```

```
In [47]: 1 plt.bar(data['STATE_UT_NAME'],data['ANNUAL'])
2 plt.xlabel("State_ut_name")
3 plt.ylabel("ANNUAL")
4 plt.show()
```



```
In [59]: 1 x=data[['STATE_UT_NAME', 'JAN', 'FEB', 'MAR', 'APR', 'MAY', 'JUN',
2         'JUL', 'AUG', 'SEP', 'OCT', 'NOV', 'DEC', 'Jan-Feb',
3         'Mar-May', 'Jun-Sep', 'Oct-Dec']]
4 y=data[['ANNUAL']]
```

```
In [60]: 1 from sklearn.model_selection import train_test_split
2 x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.3)
```

## LinearRegression

```
In [61]: 1 from sklearn.linear_model import LinearRegression
2 lr=LinearRegression()
3 lr.fit(x_train,y_train)
4 print(lr.score(x_test,y_test))
5 print(lr.score(x_train,y_train))
```

```
1.0
1.0
```

## Ridge, Lasso,RidgeCV and LassoCV

```
In [62]: 1 from sklearn.linear_model import Lasso,Ridge,LassoCV,RidgeCV
```

```
In [63]: 1 ridge=Ridge(alpha=20)
2 ridge.fit(x_train,y_train)
3 print(ridge.score(x_test,y_test))
4 print(ridge.score(x_train,y_train))
```

```
0.9999999999983411
0.9999999999983038
```

```
In [64]: 1 lasso=Lasso(alpha=20)
2 lasso.fit(x_train,y_train)
3 print(lasso.score(x_test,y_test))
4 print(lasso.score(x_train,y_train))
```

```
0.9999966231147088
0.9999979052675522
```

In [65]:

```
1 ridge_cv=RidgeCV(alphas=[20,30,40,50])
2 ridge_cv.fit(x_train,y_train)
3 print(ridge_cv.score(x_test,y_test))
4 print(ridge_cv.score(x_train,y_train))
```

0.9999999999983404

0.9999999999983021

In [66]:

```
1 lasso_cv=LassoCV(alphas=[20,40,60,80,100])
2 lasso_cv.fit(x_train,y_train)
3 print(lasso_cv.score(x_test,y_test))
4 print(lasso_cv.score(x_train,y_train))
```

0.9999966231147088

0.9999979052675522

ElasticNet

In [67]:

```
1 from sklearn.linear_model import ElasticNet
2 en = ElasticNet()
3 en.fit(x_train,y_train)
4 print(en.score(x_test,y_test))
5 print(en.score(x_train,y_train))
```

0.999997350849389

0.999997754503391

In [68]:

```
1 print(en.coef_)
2 print(en.intercept_)
```

[ 0. 3.98877046 4.06401905 2.85788025 2.84060681 2.8482631

0.95094543 0.95238712 0.95190578 0.95118891 -0.19115325 -0.20630732

-0.16499712 -3.03943597 -1.84734926 0.04822758 1.19302125]

[0.1604596]

In [71]:

```
1 sns.pairplot(data,x_vars=['JAN', 'FEB', 'MAR', 'APR', 'MAY', 'JUN',
2                        'JUL', 'AUG', 'SEP', 'OCT', 'NOV', 'DEC', 'ANNUAL', 'Jan-Feb',
3                        'Mar-May', 'Jun-Sep', 'Oct-Dec'],y_vars=['STATE_UT_NAME'],height=5,aspect=0.5,kind='reg')
```

Out[71]: <seaborn.axisgrid.PairGrid at 0x16f5a554430>

A pairplot generated by Seaborn showing the relationship between rainfall (STATE\_UT\_NAME) and various months (JAN to DEC) and annual averages (ANNUAL, Jan-Feb, Mar-May, Jun-Sep, Oct-Dec). The y-axis for all plots is 'STATE\_UT\_NAME' with a scale from 0 to 40. The x-axis for each plot represents a different time period. Each plot includes a blue regression line and a light blue shaded area representing the confidence interval. The plots are arranged in a grid, with the diagonal showing the distribution of rainfall for each month/period.

Conclusion:-

Annually Manipur got more rainfall compared to all other states

In [ ]:

```
1
```

localhost:8888/notebooks/Mini project - 3.ipynb

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