

# **Institute of Technical Education and Research [ITER] , Bhubaneswar**

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**TRAINING REPORT ON COAL PRICE ANALYSIS AND COAL CONSUMPTION**

**FOR BCCL**



**BHARAT COKING COAL LIMITED**

(A SUBSIDIARY OF COAL INDIA LIMITED)

DHANBAD, JHARKHAND

**B.TECH IN COMPUTER SCIENCE AND ENGINEERING**  
**(2022 – 2026)**

UNDERTAKEN AT :

**SYSTEM DEPARTMENT BCCL , DHANBAD**

## **CERTIFICATE**

This is to certify that I am a student of 2<sup>nd</sup> year  
B.TECH at **SIKSHA 'O' ANUSANDHAN**  
**,Bhubaneswar** have successfully completed my  
project report entitled "**COAL PRICES ANALYSIS**  
**AND COAL CONSUMPTION**". It's Industrial  
training starts from **1<sup>st</sup> of July to 28<sup>th</sup> of July (4 weeks)** at **System Department BCCL , Headquarter ,Dhanbad .**

The matter embodied in this project is entirely  
genuine and authentic work done by the students .  
This project has not been submitted to any other  
institute for fulfilment of the requirements of any  
course of studying to the best of my knowledge .

HOD (SYSTEM )  
SYSTEM DEPARTMENT  
B.C.C.L ,DHANBAD

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course of studying to the best of my knowledge .

**MR. AVINASH KUMAR NANDAN**

**DEPUTY MANAGER SYSTEM**

**SYSTEM DEPARTMENT**

**B.C.C.L ,DHANBAD**

## **ACKNOWLEDGEMENT**

I would like to extend our sincere thanks to our project manager **MR. AVINASH KUMAR NANDAN, Deputy Manager System .**

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I am also thankful to my parents ,seniors and batchmates for showing their support , co-operation and continuous motivation .

**PRAKSHI PRIYA**

**SIKSHA 'O' ANUSANDHAN ,BHUBANESWAR**

**REG NO. – 2241016464**

**BRANCH – CSE [DATA SCIENCE]**

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## **ABSTRACT**

Government have been striving since the late 1990's to find better ways to connect with their constituents via the Web . By putting government information online ,making it easily accessible ,findable, readably, understandable , people can now interact with their government in ways never imagined . Current web technology allows government to share with public a variety of information is unlimited quantities on demand . Technology is always available to allow citizens to bring issues of concern to the attention of local regional governments .

However, exploiting these capabilities within government system is a challenge that encompasses environmental policy , legal and cultural issues . Establishing effective eGovernment requires openness transparency , collaboration , and skills in taking advantage of the capabilities of the World Wide Web . The goal of this project is to ensure that the government services are made available to the common people electronically by improved online infrastructure .

## COAL INDIA LIMITED

### ➤ **INTRODUCTION TO COAL INDIA LIMITED**



Coal India Limited (CIL) is an Indian state-controlled coal mining company headquartered in Kolkata ,West Bengal . India and the largest coal producer company in the world and a Maharatna company .

The company contributes to around 82% of the coal production in India . It produced 554.14 million tons of raw coal in 2016-17 ,an increase from its earlier production of 494.24 million tons of coal during FY 2014-15 and earned a revenue of 05,435 crore (USS 14 billion) from sale of coal in the same financial year .

As on 14 October 2015 , Union Government of India owns CIL and controls the operations of CIL through Ministry of Coal .

In April 2011 , CIL was conferred the Maharatna statue by the union govt. of India , making it one of the seven Maharatnas .

As on 14 October 2015, its market capitalization was 2.11 Lakh crore .

(USS 14 Bullion ) making it India's 8<sup>th</sup> most Valuable company by market value .

Coal India limited is one of the largest coal producing companies in the world with the total production of 431.32 million metric tons in 2010-2012 . CIL has 8 holly subsidiaries.

As of March 31<sup>st</sup> 2011 , it operated 470 mines in 21 major coal fields across 8 states in India including 164 opencast mines 275 underground mines and 31 mixed mines .

# BBCL | BHARAT COKING COAL LIMITED

## ➤ INTRODUCTION TO BCCL



**BHARAT COKING COAL LIMITED [ BCCL ]** is a subsidiary of Coal India Limited with its Headquarter in Dhanbad , India . It was incorporated in January 1972 to operate coking coal mines (214 in number) operating in Jharia and Raniganj Coalfields and was taken over by the government of India on 16 October 1971 .

The company operates 81 coal mines which include 40 underground , 18 opencast and 23 mixed mines as of April 2010 . The company also runs 6 Coking coal washeries ,two non-coking coal washeries , one captive power plant (20 MW) and five by -product coke plants . The mines are grouped into 12 areas for administration purposes .

BCCL is the major producer of prime coking coal (raw and washed ) in India . Medium Coking Coal is produced in its mines in MOHUDA and BARAKAR areas . In addition to production of hard coke , BCCL operates washeries , sand gathering plants , a network of aerial ropeways for transport of sand , and coal bed methane – based power plant in Moonidih .

## ➤ **CURRENT SITUATION**

**BHARAT COKING COAL LIMITED** gave an annual coal production of around 37.04 million tons in 2016-17 with a turnover of 11,505 crore (USS 1.7 billion ) . It has a manpower of about 49,901 as cm 01.10 – 2017 .

## ADMINISTRATIVE AREAS

There are 12 areas in the BCCL :

<u>ADMINISTRATIVE AREA</u>	<u>NAME</u>
1. Area No 1	Barora Area
2. Area No 2	Block II Area
3. Area No 3	Govindpur Area
4. Area No 4	Katras Area
5. Area No 5	Sijua Area
6. Area No 6	Kusunda Area
7. Area No 7	Putki Balihari Area
8. Area No 8	Bastacola Area
9. Area No 9	Lodna Area
10. Area No 10	Eastern Jharia Area
11. Area No 11	Chanch Victoria Area
12. Area No 12	Western Jharia Area

## INTRODUCTION

Traditional methods of Web services by BCCL , ever since known , have relied on pen and paper with several layers of substantiation which not only slows down the process but also gives power in not so worthy hands to flout the rules laid by the government to successfully execute this searching procedure . This portal has been developed to tackle these two major issues along with other minor problems which one may be able to see on a deeper analysis .

Understanding coal price dynamics can help BCCL optimize its production, marketing strategies, and financial performance. It is responsible for producing prime coking coal for the steel sector and is a major player in the Indian coal industry.

Data visualization is a powerful tool for enhancing the analysis and understanding of coal consumption at BCCL. By transforming raw data into meaningful visual insights, BCCL can improve operational efficiency, make informed decisions, and effectively communicate findings.

Data visualization transforms raw data into meaningful insights, aiding BCCL in making data-driven decisions, optimizing operations, ensuring compliance, and effectively communicating with stakeholders. By leveraging advanced visualization tools, BCCL can enhance its overall efficiency, reduce costs, and improve its environmental and safety standards.

## MOTIVATION

As mentioned earlier , traditional methods to physically return items that are found or find an item that is just too difficult . This is a real issue faced by all of us in our day – to – day life . In today's world almost everyone is occupied with a smartphone and internet connectivity . So , this gave us an idea to develop this Online portal for the convenience of people at large . Currently it is being developed for BCCL as an intra organizational portal but later it can be scaled up to serve a larger set of people .

## AIM AND SCOPE OF THE PROJECT

The major aim of this project to provide a fast ,easy , secure and optimized platform to the common people Who are in need of our company's facilities and services . The traditional methods are time consuming as well as physically tiring . This project aims to solve basic problems faced by people in their daily lives .

Currently this portal is being developed by keeping in mind Employees of BCCL exclusively as and when deployed . Later on , it can be scaled up for larger public .

# TOOLS AND TECHNOLOGIES USED

## ➤ DATA VISUALIZATION

**Data visualization** is the art and science of transforming raw data into meaningful and easily understandable visual representations. It involves using charts, graphs, maps, and other visual elements to reveal patterns, trends, and insights that might be hidden within the data.

### Why is Data Visualization Important?

- **Enhanced Understanding:** Visuals process information faster than text, making complex data accessible to everyone.
- **Effective Communication:** Visuals tell a story, making it easier to convey insights and persuade audiences.
- **Improved Decision Making:** By identifying trends and patterns, data visualization aids in informed decision-making.
- **Data Exploration:** Visualizations help uncover hidden relationships and correlations within the data.

### Key Principles of Effective Data Visualization

- **Clarity:** The visualization should be easy to understand.
- **Relevance:** The visual should convey the intended message.
- **Simplicity:** Avoid clutter and unnecessary elements.
- **Consistency:** Use consistent colors, fonts, and scales.
- **Interactivity:** Allow users to explore the data further.

## ➤ *Tools for Data Visualization*

There are numerous tools available for creating data visualizations, ranging from basic spreadsheet software to specialized data visualization platforms. Some popular options include:

- **Microsoft Excel:** Offers basic charting capabilities.
- **Python libraries:** Matplotlib, Seaborn, Plotly, and more.

## ➤ Microsoft Excel

**Microsoft Excel** is a powerful spreadsheet software that's become an indispensable tool for businesses, students, and individuals alike. Think of it as a digital version of a paper spreadsheet, but with immense capabilities.

### ADVANTAGES :

- **Organize data:** Create tables with rows and columns to structure your information.
- **Perform calculations:** Use formulas and functions to perform mathematical operations, from simple addition to complex financial analysis.
- **Analyse data:** Discover trends, patterns, and insights using tools like pivot tables and charts.
- **Create visualizations:** Transform data into visual representations (charts, graphs) for better understanding.
- **Manage databases:** Store and manage large datasets efficiently.
- **Automate tasks:** Use macros (recorded sequences of actions) to streamline repetitive tasks.

## ➤ **PYTHON**

**Python** is a high-level, general-purpose programming language renowned for its readability and simplicity. Its emphasis on code clarity makes it an excellent choice for beginners and experienced programmers alike .

### ADVANTAGES :

- **Object-Oriented Programming:** Supports OOP concepts for structured code.
- **Embedding Capabilities:** Can be embedded in other applications.

- **High-Level Language:** Abstracts low-level details, making it easier to focus on problem-solving.

## ➤ **PYTHON IN EXCEL**

Python in Excel integrates the power of Python directly into Microsoft Excel, allowing users to leverage Python's robust data analysis and visualization libraries within the familiar Excel environment. This feature is available in Microsoft 365 and is particularly useful for those who want to enhance their Excel spreadsheets with Python's capabilities.

### **Advantages of Python in Excel**

The integration of Python into Excel brings a powerful combination of capabilities. Here are some key advantages:

1. Enhanced Data Analysis
2. Advanced Data Visualization
3. Automation of Tasks
4. Improved Data Handling
5. Integration with Jupyter Notebooks
6. Machine Learning and Predictive Analytics.

## **DESIGN AND IMPLEMENTATION**

In this project , the portal has been focusing on two divisions that is coal prices and coal consumption . In this project , we use data visualization to create effective and visually appealing charts in Excel to analyse and present various aspects such as production statistics, financial performance, safety records, environmental impact . We created a line chart for the purpose of monthly production volumes or financial performance. We create bar graph for comparing different categories, like production by mine or safety incidents by type. We also created pie charts for showing the composition of a whole, like revenue sources in both divisions . We also plot scatter graph for showing relationships between variables, like production volume vs. revenue. We also

customize the graphs in different forms such as titles and labels , legends , colour and styles and data labels by using matplotlib in python . We have applied library like numpy , pandas , matplotlib and seaborn to design different forms of graphs with dataset installed . In Coal Price Analysis project , there are some benefits for BCCL are revenue optimization , risk management , operational efficiency , strategic planning , price analysis , competitive analysis . Hence , BCCL can enhance its decision-making, improve profitability, and strengthen its market position. from coal prices analysis project what is beneficial for bccl . A comprehensive coal consumption analysis can provide invaluable insights for Bharat Coking Coal Limited (BCCL) to optimize its operations, enhance market position, and contribute to strategic decision-making . The focus of this implementation is on efficiency and providing transparency to the users as well as the system .

## SCREENSHOTS

## > COAL PRICES ANALYSIS

The screenshot shows a Microsoft Excel interface with the following details:

- File Tab:** COAL PRICES ANALYSIS (F).xlsx • Saved to this PC
- Formulas Bar:** Calculation, Save, Insert Python
- Message Bar:** SECURITY WARNING Some active content has been disabled. Click for more details. Enable Content
- Code in Formula Bar:** PY import pandas as pd  
df = xl("A1:D148", headers=True)
- Data View:** A table titled "147x4 DataFrame" is displayed, showing the data from the formula bar.

	Entity	Code	Year	Coal - Prices
1	Asian marker price (BP)		2001	36.89459991455078
2	Asian marker price (BP)		2002	30.406665802001953
3	Asian marker price (BP)		2003	36.529197692871094
4	Asian marker price (BP)		2004	72.41666412353516
5	Asian marker price (BP)		2005	61.84000015258789
6	Asian marker price (BP)		2006	56.47361373901367
7	Asian marker price (BP)		2007	84.56500244140625
8	Asian marker price (BP)		2008	148.05833435058594
9	Asian marker price (BP)		2009	78.80666351
10	Asian marker price (BP)		2010	105.43083190917969
11	Asian marker price (BP)		2011	125.73666381835938
12	Asian marker price (BP)		2012	105.50416564941406
13	Asian marker price (BP)		2013	90.89666748046875
14	Asian marker price (BP)		2014	77.88749694824219
15	Asian marker price (BP)		2015	63.516685485839844
16	Asian marker price (BP)		2016	71.12264251708984
17	Asian marker price (BP)		2017	99.57555389404297
18	Asian marker price (BP)		2018	111.69308471679688
19	Asian marker price (BP)		2019	88.98251342773438
20	Asian marker price (BP)		2020	71.33166504
21	Asian marker price (BP)		2021	45.162505010251562

COAL PRICES ANALYSIS (F).xlsx • Saved to this PC

File Home Insert Draw Page Layout Formulas Data Review View Automate Developer Help

AutoSave Off Undo Redo Calculation Save Insert Python

**SECURITY WARNING** Some active content has been disabled. Click for more details. [Enable Content](#)

F3 : PY df.head()

	A	B	C	D	E	F	G	H	I	J	K	L	M
1	Entity	Code	Year	Coal - Prices									[+] Image
2	Asian marker price (BP)		2001	36.89459991455078	[+] DataFrame								[+] Image
3	Asian marker price (BP)		2002	30.406665802001953	[+] DataFrame								[+] Image
4	Asian marker price (BP)		2003	36.529197692871094	[+] DataFrame								[+] Image
5	Asian marker price (BP)		2004	72.41666412353516									[+] Image
6	Asian marker price (BP)		2005	61.84000015258789	[+] tuple								[+] Image
7	Asian marker price (BP)		2006	56.47361373901367	[+] 588								[+] Image
8	Asian marker price (BP)		2007	84.56500244140625	[+] method								[+] Image
9	Asian marker price (BP)		2008	148.05833435058594	[+] Series								[+] Image
10	Asian marker price (BP)		2009	78.80666351	[+] Series								[+] Image
11	Asian marker price (BP)		2010	105.43083190917969	[+] Series								[+] Image
12	Asian marker price (BP)		2011	125.73666381835938	[+] DataFrame								[+] Image
13	Asian marker price (BP)		2012	105.50416564941406	[+] Series								[+] Image

**5x4 DataFrame**

	Entity	Code	Year	Coal - Prices
0	Asian marker price (BP)	None	2001	36.89459991455078
1	Asian marker price (BP)	None	2002	30.406665802001953
2	Asian marker price (BP)	None	2003	36.529197692871094
3	Asian marker price (BP)	None	2004	72.41666412353516
4	Asian marker price (BP)	None	2005	61.84000015258789

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F4 : PY df.tail()

	A	B	C	D	E	F	G	H	I	J	K	L	M	N	O
1	Entity	Code	Year	Coal - Prices										[+] Image	
2	Asian marker price (BP)		2001	36.89459991455078	[+] DataFrame									[+] Image	
3	Asian marker price (BP)		2002	30.406665802001953	[+] DataFrame									[+] Image	
4	Asian marker price (BP)		2003	36.529197692871094	[+] DataFrame									[+] Image	
5	Asian marker price (BP)		2004	72.41666412353516										[+] Image	
6	Asian marker price (BP)		2005	61.84000015258789	[+] tuple									[+] Image	
7	Asian marker price (BP)		2006	56.47361373901367	[+] 588									[+] Image	
8	Asian marker price (BP)		2007	84.56500244140625	[+] method									[+] Image	
9	Asian marker price (BP)		2008	148.05833435058594	[+] Series									[+] Image	
10	Asian marker price (BP)		2009	78.80666351	[+] Series									[+] Image	
11	Asian marker price (BP)		2010	105.43083190917969	[+] Series									[+] Image	
12	Asian marker price (BP)		2011	125.73666381835938	[+] DataFrame									[+] Image	
13	Asian marker price (BP)		2012	105.50416564941406	[+] Series									[+] Image	
14	Asian marker price (BP)		2013	90.89666748046875	[+] Series									[+] Image	
15	Asian marker price (BP)		2014	77.88749694824219	[+] DataFrame									[+] Image	
16	Asian marker price (BP)		2015	63.516685485839844	[+] Series									[+] Image	
17	Asian marker price (BP)		2016	71.12264251708984	[+] Series									[+] Image	

**5x4 DataFrame**

	Entity	Code	Year	Coal - Prices
142	US Central Appalachian coal spot price index (BP)	None	2017	63.82529067993164
143	US Central Appalachian coal spot price index (BP)	None	2018	72.84400939941406
144	US Central Appalachian coal spot price index (BP)	None	2019	57.16398239135742
145	US Central Appalachian coal spot price index (BP)	None	2020	42.76636887
146	US Central Appalachian coal spot price index (BP)	None	2021	68.53813934326172

COAL PRICES ANALYSIS (F).xlsx • Saved to this PC

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G5 : PY df.describe()

	A	B	C	D	E	F	G	H	I	J	K	L	M
1	Entity	Code	Year	Coal - Prices									[+] Image
2	Asian marker price (BP)		2001	36.89459991455078	[+] DataFrame								[+] Image
3	Asian marker price (BP)		2002	30.406665802001953	[+] DataFrame								[+] Image
4	Asian marker price (BP)		2003	36.529197692871094	[+] DataFrame								[+] Image
5	Asian marker price (BP)		2004	72.41666412353516									[+] Image
6	Asian marker price (BP)		2005	61.84000015258789	[+] tuple								[+] Image
7	Asian marker price (BP)		2006	56.47361373901367	[+] 588								[+] Image
8	Asian marker price (BP)		2007	84.56500244140625	[+] method								[+] Image
9	Asian marker price (BP)		2008	148.05833435058594	[+] Series								[+] Image
10	Asian marker price (BP)		2009	78.80666351	[+] Series								[+] Image
11	Asian marker price (BP)		2010	105.43083190917969	[+] Series								[+] Image
12	Asian marker price (BP)		2011	125.73666381835938	[+] DataFrame								[+] Image
13	Asian marker price (BP)		2012	105.50416564941406	[+] Series								[+] Image
14	Asian marker price (BP)		2013	90.89666748046875	[+] Series								[+] Image
15	Asian marker price (BP)		2014	77.88749694824219	[+] DataFrame								[+] Image
16	Asian marker price (BP)		2015	63.516685485839844	[+] Series								[+] Image
17	Asian marker price (BP)		2016	71.12264251708984	[+] Series								[+] Image

**8x1 DataFrame**

	Year
count	147
mean	2011
std	6.076002651
min	2001
25%	2006
50%	2011
75%	2016
max	2021

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F9 : PY df.isnull().sum()

4x1 Series

Entity	Code	Year	Coal - Prices
Asian marker price (BP)		2001	36.89459991455078
Asian marker price (BP)		2002	30.406665802001953
Asian marker price (BP)		2003	36.529197692871094
Asian marker price (BP)		2004	72.41666412353516
Asian marker price (BP)		2005	61.84000015258789
Asian marker price (BP)		2006	56.47361373901367
Asian marker price (BP)		2007	84.56500244140625
Asian marker price (BP)		2008	148.05833435058594

ANACONDA.

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F11 : PY df['Entity'].value\_counts()

7x1 Series

Entity	count
Asian marker price (BP)	21
China Qinhuangdao spot price (BP)	21
Japan coking coal import CIF price (BP)	21
Japan steam coal import CIF price (BP)	21
Japan steam spot CIF price (BP)	21
Northwest Europe marker price (BP)	21
US Central Appalachian coal spot price index (BP)	21

ANACONDA.

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**SECURITY WARNING** Some active content has been disabled. Click for more details. Enable C

F13 : PY dx['Year'].value\_counts().sort\_index()

21x1 Series

Year	count
2001	7
2002	7
2003	7
2004	7
2005	7
...	...
2017	7
2018	7
2019	7
2020	7
2021	7

ANACONDA.

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F12 : PY df['Code'].value\_counts()  
dx = df.drop(columns=['Code'])

A B C D E F

Entity Code Year Coal - Prices

1 Asian marker price (BP) 2001 36.89459991455078 [r] DataFrame  
2 Asian marker price (BP) 2002 30.406665802001953 [r] DataFrame  
3 Asian marker price (BP) 2003 36.529197692871094 [r] DataFrame  
4 Asian marker price (BP) 2004 72.41666412353516 [r] tuple  
5 Asian marker price (BP) 2005 61.84000015258789 [r] 588  
6 Asian marker price (BP) 2006 56.47361373901367 [r] 588  
7 Asian marker price (BP) 2007 84.56500244140625 [r] method  
8 Asian marker price (BP) 2008 148.05833435058594 [r] Series  
9 Asian marker price (BP) 2009 78.80666351 [r] Series  
10 Asian marker price (BP) 2010 105.43083190917969 [r] Series  
11 Asian marker price (BP) 2011 125.73666381835938 [r] DataFrame

**147x3 DataFrame**

	Entity	Year	Coal - Prices
0	Asian marker price (BP)	2001	36.89459991455078
1	Asian marker price (BP)	2002	30.406665802001953
2	Asian marker price (BP)	2003	36.529197692871094
3	Asian marker price (BP)	2004	72.41666412353516
4	Asian marker price (BP)	2005	61.84000015258789
...	...	...	...
142	US Central Appalachian coal spot price index (BP)	2017	63.82529067993164
143	US Central Appalachian coal spot price index (BP)	2018	72.84400939941406
144	US Central Appalachian coal spot price index (BP)	2019	57.16398239135742
145	US Central Appalachian coal spot price index (BP)	2020	42.76636887
146	US Central Appalachian coal spot price index (BP)	2021	68.53813934326172

ANACONDA.

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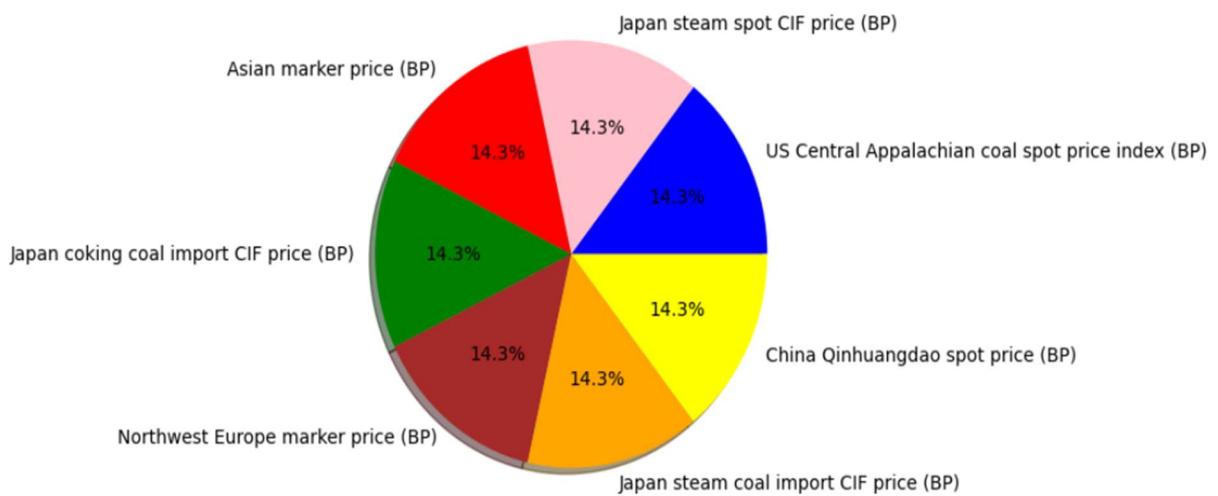
H15 : PY d\_0 = {'y\_0': [2001, 2002, 2003, 2004, 2005, 2006, 2007, 2008, 2009, 2010, 2011, 2012, 2013, 2014, 2015, 2016, 2021], 'p\_0': [36.89459991455078, 30.406665802001953, 36.529197692871094, 72.41666412353516, 61.84000015258789, 56.47361373901367, 84.56500244140625, 148.05833435058594, 78.8066635131836, 105.43083190917969, 125.73666381835938, 105.50416564941406, 90.89666748046876, 77.88749694824219, 63.51668548583984, 71.12264251708984, 99.57555389404295, 111.69308471679688, 88.98251342773438, 71.3316650390625, 145.16250610351562]}  
df\_0 = pd.DataFrame(data=d\_0)  
df\_0['p\_0'].describe()

**8x1 Series**

	p_0
count	21
mean	83.94434375
std	32.50224972
min	30.4066658
25%	63.51668549
50%	78.80666351
75%	105.43083190917969
max	148.05833435058594

```
PY entity = ('US Central Appalachian coal spot price index (BP)', 'Japan steam spot CIF price (BP)', 'Asian marker price (BP)', 'Japan coking coal import CIF price (BP)', 'Northwest Europe marker price (BP)', 'Japan steam coal import CIF price (BP)', 'China Qinhuangdao spot price (BP)')
ent_counts = (21, 21, 21, 21, 21, 21, 21)

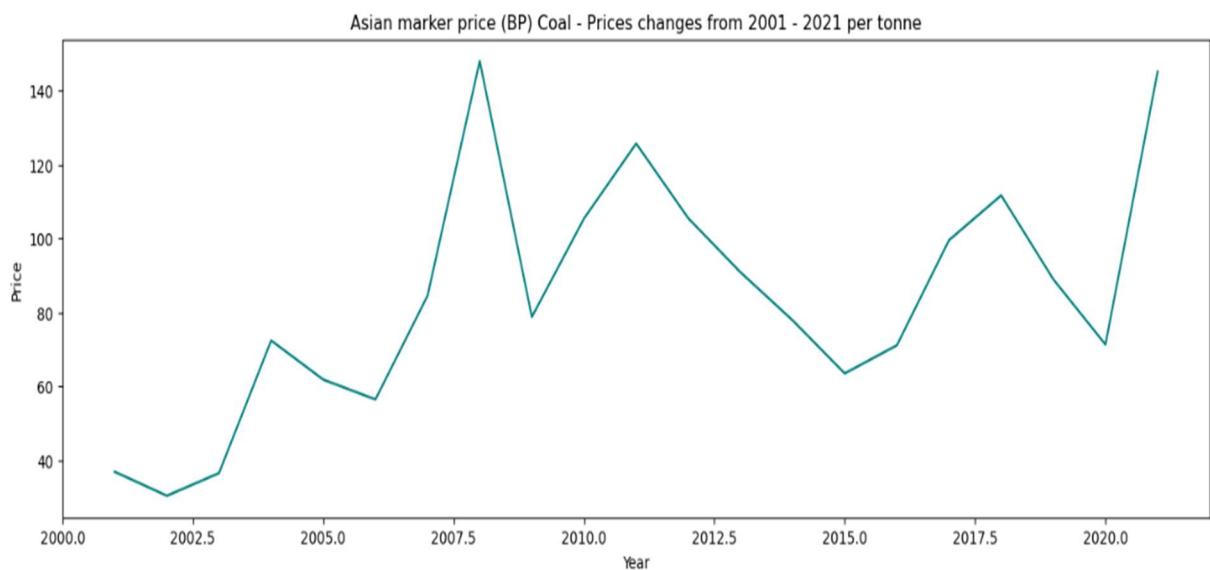
colors = ('blue','pink', 'red', 'green', 'brown', 'orange', 'yellow')
plt.figure(figsize = (16, 5))
plt.pie(
    ent_counts,
    labels=entity,
    autopct='%.1f%%',
    colors=colors,
    shadow=True
)
plt.show()
```



```

PY an = dx.loc[0:20]
n1_0 = an["Coal - Prices"].tolist()
x_0 = (2001, 2002, 2003, 2004, 2005, 2006, 2007, 2008, 2009, 2010, 2011, 2012, 2013, 2014, 2015, 2016, 2017, 2018, 2019, 2020, 2021)
y_0 = (36.89459991455078,
       30.406665802001957,
       36.52919769287109,
       72.41666412353516,
       61.84000015258789,
       56.47361373901367,
       84.56500244140625,
       148.05833435058594,
       78.8066635131836,
       105.43083190917967,
       125.73666381835938,
       105.50416564941406,
       90.89666748046876,
       77.88749694824219,
       63.51668548583984,
       71.12264251708984,
       99.57555389404295,
       111.69308471679688,
       88.98251342773438,
       71.3316650390625,
       145.16250610351562)
plt.figure(figsize = (16, 5))
plt.title("Asian marker price (BP) Coal - Prices changes from 2001 - 2021 per tonne")

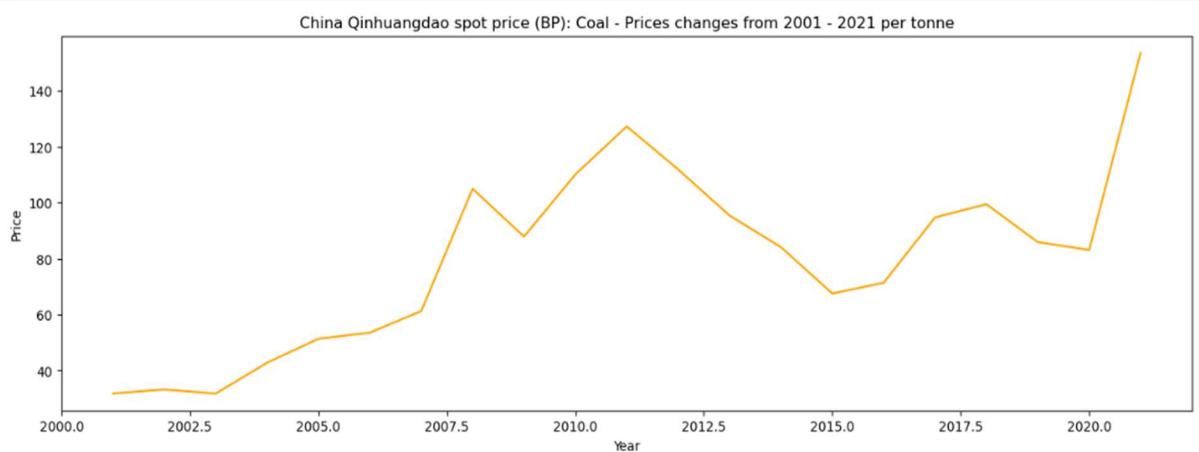
```



```

PY 33.19486618041992,
31.744373321533203,
42.76045989990234,
51.34153747558594,
53.52624130249024,
61.22782516479492,
104.97216033935548,
87.86170196533203,
110.07920837402344,
127.26840209960938,
111.88970947265624,
95.41863250732422,
84.11642456054688,
67.5318832397461,
71.34886169433594,
94.72347259521484,
99.4467544555664,
85.89109802246094,
83.09520721435547,
153.5463409423828)
plt.figure(figsize = (16, 5))
plt.title("China Qinhuangdao spot price (BP): Coal - Prices changes from 2001 - 2021 per tonne")
plt.xlabel("Year")
plt.ylabel("Price")
plt.plot(x_1, y_1, color ="orange")
plt.show()

```

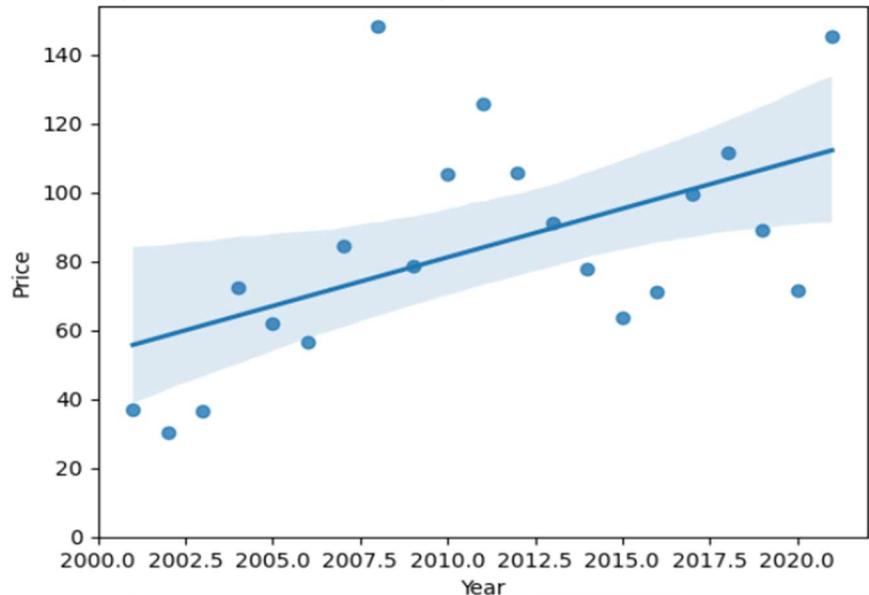


```

PY sns.regplot(x="y_θ", y="p_θ", data=df_θ)
plt.title("Asian marker price (BP): 'Coal - Prices per tonne' and 'Year' (2001-2021) Correlation")
plt.xlabel("Year")
plt.ylabel("Price")
plt.ylim(0,)

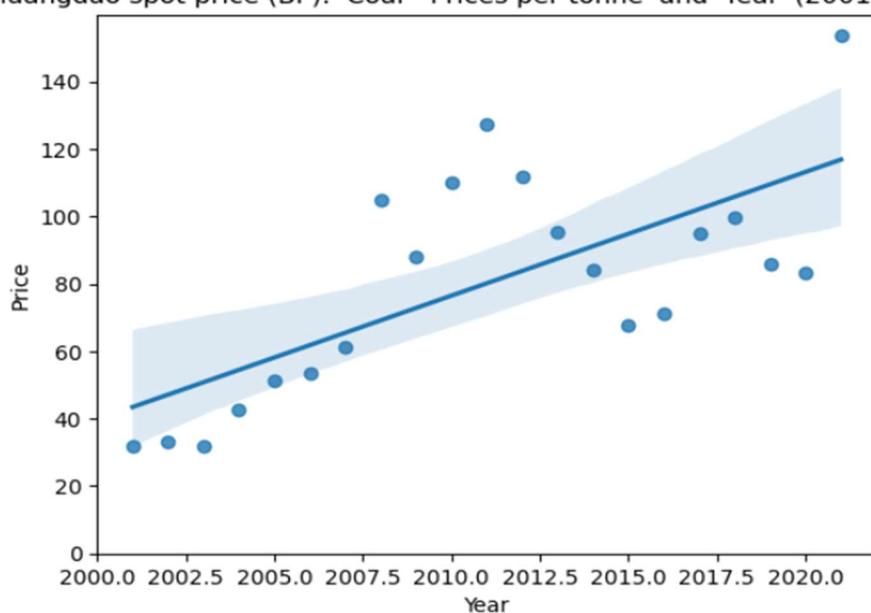
```

Asian marker price (BP): 'Coal - Prices per tonne' and 'Year' (2001-2021) Correlation



```
py sns.regplot(x="y_1", y="p_1", data=df_1)
plt.title("China Qinhuangdao spot price (BP): 'Coal - Prices per tonne' and 'Year' (2001-2021) Correlation")
plt.xlabel("Year")
plt.ylabel("Price")
plt.ylim(0,)
```

China Qinhuangdao spot price (BP): 'Coal - Prices per tonne' and 'Year' (2001-2021) Correlation

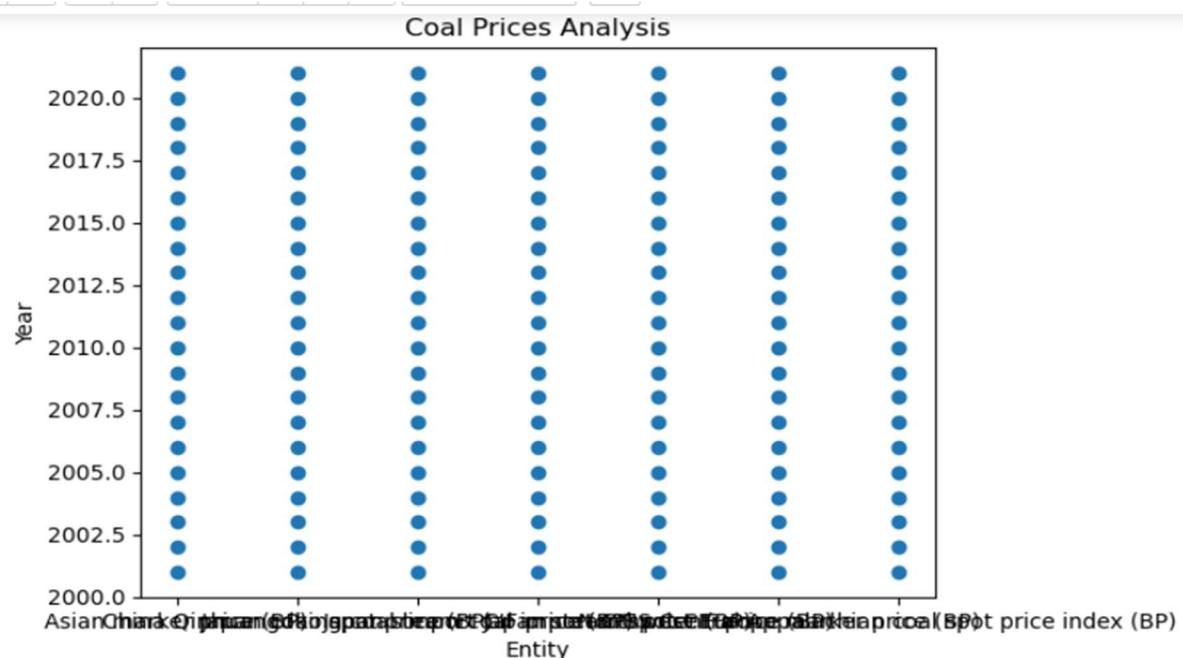


```

PY plt.scatter(df['Entity'], df['Year'])

# Adding Title to the Plot
plt.title("Coal Prices Analysis")
# Setting the X and Y labels
plt.xlabel('Entity')
plt.ylabel('Year')
plt.show()

```



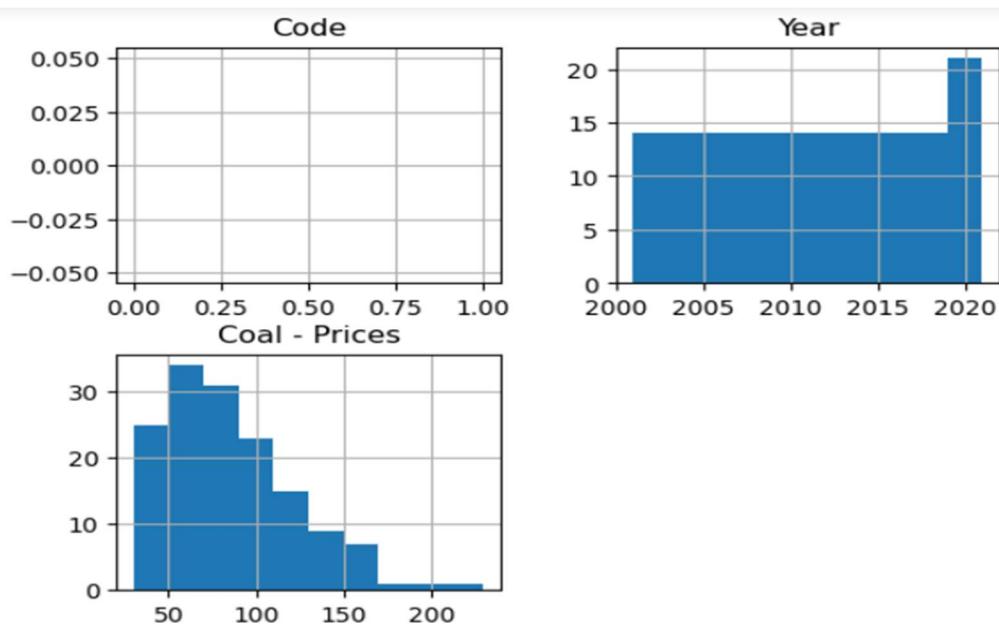
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```

PY df.hist(grid=True)
plt.title("Coal Prices Distribution ($)")
plt.show()

```



# INDIAN COAL CONSUMPTION

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```
I2 PY import pandas as pd
df = xl("A1:G1958", headers=True)
```

srcYear	Types of coal	Coal Consuming Sectors	Coal Consumed	YearCode	Year
2011	RAW COAL	OTHER	0.131	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	POWER (UTILITY)	12.545	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	POWER (CAPTIVE)	3.363	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	PULP & PAPER	0.095	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	FERTILISERS	1.111	2011	Financial Year (Apr - Mar), 2011
2011	WASHED COAL	POWER (UTILITY)	1.599	2011	Financial Year (Apr - Mar), 2011
2011	MIDDINGLS	POWER (CAPTIVE)	0.462	2011	Financial Year (Apr - Mar), 2011
2011	WASHED COAL	CEMENT	0.045	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	STEEL (BOILERS)	0.121	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	POWER (UTILITY)	34.997	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	SPONGE IRON	0.177	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	POWER (CAPTIVE)	11.644	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	TEXTILES & RAYONS	0.023	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	CHEMICAL	0.021	2011	Financial Year (Apr - Mar), 2011
2011	WASHED COAL	OTHER	0.102	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	SPONGE IRON	0.821	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	BRICKS	0.159	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	POWER (UTILITY)	21.162	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	POWER (CAPTIVE)	1.235	2011	Financial Year (Apr - Mar), 2011
2011	LIGNITE	POWER (UTILITY)	5.824	2011	Financial Year (Apr - Mar), 2011
2011	MIDDINGLS	POWER (CAPTIVE)	2.061	2011	Financial Year (Apr - Mar), 2011

1957x7 DataFrame

srcStateName	srcYear	Types of coal	Coal Consuming Sectors
DELHI	2011	RAW COAL	OTHER
PUNJAB	2011	RAW COAL	POWER (UTILITY)
ANDHRA PRA...	2011	RAW COAL	POWER (CAPTIVE)
ASSAM	2011	RAW COAL	PULP & PAPER
PUNJAB	2011	RAW COAL	FERTILISERS
...	...	...	...
TAMILNADU	2020	LIGNITE	BRICKS
OTHERS	2020	RAW COAL	OTHER
ODISHA	2020	RAW COAL	COLLIERY OWN - CONSUM...
WEST BENGAL	2020	RAW COAL	OTHER
ODISHA	2020	RAW COAL	PULP & PAPER

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```
I4 PY df.head()
```

srcYear	Types of coal	Coal Consuming Sectors	Coal Consumed	YearCode	Year
2011	RAW COAL	OTHER	0.131	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	POWER (UTILITY)	12.545	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	POWER (CAPTIVE)	3.363	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	PULP & PAPER	0.095	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	FERTILISERS	1.111	2011	Financial Year (Apr - Mar), 2011
2011	WASHED COAL	POWER (UTILITY)	1.599	2011	Financial Year (Apr - Mar), 2011
2011	MIDDINGLS	POWER (CAPTIVE)	0.462	2011	Financial Year (Apr - Mar), 2011
2011	WASHED COAL	CEMENT	0.045	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	STEEL (BOILERS)	0.121	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	POWER (UTILITY)	34.997	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	SPONGE IRON	0.177	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	POWER (CAPTIVE)	11.644	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	TEXTILES & RAYONS	0.023	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	CHEMICAL	0.021	2011	Financial Year (Apr - Mar), 2011
2011	WASHED COAL	OTHER	0.102	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	SPONGE IRON	0.821	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	BRICKS	0.159	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	POWER (UTILITY)	21.162	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	POWER (CAPTIVE)	1.235	2011	Financial Year (Apr - Mar), 2011
2011	LIGNITE	POWER (UTILITY)	5.824	2011	Financial Year (Apr - Mar), 2011
2011	MIDDINGLS	POWER (CAPTIVE)	2.061	2011	Financial Year (Apr - Mar), 2011

5x2 DataFrame

	Coal Consumed (ton)	Price per ton (USD)
0	10	20
1	15	30
2	8	25
3	12	32
4	9	28

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I5 PY df.tail()

	B	C	D	E	F	G	H	I	J	K	L	M	N	O
1	srcYear	Types of coal	Coal Consuming Sectors	Coal Consumed	YearCode	Year		[r] DataFrame		[r] DataFrame		[r] Image		
2	2011	RAW COAL	OTHER	0.131	2011	Financial Year (Apr - Mar), 2011								
3	2011	RAW COAL	POWER (UTILITY)	12.545	2011	Financial Year (Apr - Mar), 2011								
4	2011	RAW COAL	POWER (CAPTIVE)	3.363	2011	Financial Year (Apr - Mar), 2011								
5	2011	RAW COAL	PULP & PAPER	0.095	2011	Financial Year (Apr - Mar), 2011								
6	2011	RAW COAL	FERTILISERS	1.111	2011	Financial Year (Apr - Mar), 2011								
7	2011	WASHED COAL	POWER (UTILITY)	1.599	2011	Financial Year (Apr - Mar), 2011								
8	2011	MIDDINGLS	POWER (CAPTIVE)	0.462	2011	Financial Year (Apr - Mar), 2011								
9	2011	WASHED COAL	CEMENT	0.045	2011	Financial Year (Apr - Mar), 2011								
10	2011	RAW COAL	STEEL (BOILERS)	0.121	2011	Financial Year (Apr - Mar), 2011								
11	2011	RAW COAL	POWER (UTILITY)	34.997	2011	Financial Year (Apr - Mar), 2011								
12	2011	RAW COAL	SPONGE IRON	0.177	2011	Financial Year (Apr - Mar), 2011								
13	2011	RAW COAL	POWER (CAPTIVE)	11.644	2011	Financial Year (Apr - Mar), 2011								
14	2011	RAW COAL	TEXTILES & RAYONS	0.023	2011	Financial Year (Apr - Mar), 2011								
15	2011	RAW COAL	CHEMICAL	0.021	2011	Financial Year (Apr - Mar), 2011								
16	2011	WASHED COAL	OTHER	0.102	2011	Financial Year (Apr - Mar), 2011								
17	2011	RAW COAL	SPONGE IRON	0.821	2011	Financial Year (Apr - Mar), 2011								
18	2011	RAW COAL	BRICKS	0.159	2011	Financial Year (Apr - Mar), 2011								
19	2011	RAW COAL	POWER (UTILITY)	21.162	2011	Financial Year (Apr - Mar), 2011								
20	2011	RAW COAL	POWER (CAPTIVE)	1.235	2011	Financial Year (Apr - Mar), 2011								
21	2011	LIGNITE	POWER (UTILITY)	5.824	2011	Financial Year (Apr - Mar), 2011								
22	2011	INDUSNOC	POWER (CAPTIVE)	0.001	2011	Financial Year (Apr - Mar), 2011								

5x2 DataFrame

	Coal Consumed (ton)	Price per ton (USD)
1	5	25
2	6	22
3	7	38
4	8	29
5	9	24

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I10 PY df.describe()

	srcYear	Types of coal	Coal Consuming Sectors	Coal Consumed	YearCode	Year	H	I	J	K	L	M	N	O
1	2011	RAW COAL	OTHER	0.131	2011	Financial Year (Apr - Mar), 2011								
2	2011	RAW COAL	POWER (UTILITY)	12.545	2011	Financial Year (Apr - Mar), 2011								
3	2011	RAW COAL	POWER (CAPTIVE)	3.363	2011	Financial Year (Apr - Mar), 2011								
4	2011	RAW COAL	PULP & PAPER	0.095	2011	Financial Year (Apr - Mar), 2011								
5	2011	RAW COAL	FERTILISERS	1.111	2011	Financial Year (Apr - Mar), 2011								
6	2011	WASHED COAL	POWER (UTILITY)	1.599	2011	Financial Year (Apr - Mar), 2011								
7	2011	MIDDINGLS	POWER (CAPTIVE)	0.462	2011	Financial Year (Apr - Mar), 2011								
8	2011	WASHED COAL	CEMENT	0.045	2011	Financial Year (Apr - Mar), 2011								
9	2011	RAW COAL	STEEL (BOILERS)	0.121	2011	Financial Year (Apr - Mar), 2011								
10	2011	RAW COAL	POWER (CAPTIVE)	11.644	2011	Financial Year (Apr - Mar), 2011								
11	2011	RAW COAL	TEXTILES & RAYONS	0.023	2011	Financial Year (Apr - Mar), 2011								
12	2011	RAW COAL	CHEMICAL	0.021	2011	Financial Year (Apr - Mar), 2011								
13	2011	WASHED COAL	OTHER	0.102	2011	Financial Year (Apr - Mar), 2011								
14	2011	RAW COAL	SPONGE IRON	0.821	2011	Financial Year (Apr - Mar), 2011								
15	2011	RAW COAL	BRICKS	0.159	2011	Financial Year (Apr - Mar), 2011								

8x2 DataFrame

	srcYear	Coal Consumed
count	10	10
mean	2011	5.4469
std	0	11.05962138
min	2011	0.045
25%	2011	0.1235
50%	2011	0.7865
75%	2011	2.922
max	2011	34.997

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I11 PY df.isnull().sum()

	B	C	D	E	F	G	H	I	J	K	L	M	N	O
1	srcYear	Types of coal	Coal Consuming Sectors	Coal Consumed	YearCode	Year								
2	2011	RAW COAL	OTHER	0.131	2011	Financial Year (Apr - Mar), 2011								
3	2011	RAW COAL	POWER (UTILITY)	12.545	2011	Financial Year (Apr - Mar), 2011								
4	2011	RAW COAL	POWER (CAPTIVE)	3.363	2011	Financial Year (Apr - Mar), 2011								
5	2011	RAW COAL	PULP & PAPER	0.095	2011	Financial Year (Apr - Mar), 2011								
6	2011	RAW COAL	FERTILISERS	1.111	2011	Financial Year (Apr - Mar), 2011								
7	2011	WASHED COAL	POWER (UTILITY)	1.599	2011	Financial Year (Apr - Mar), 2011								
8	2011	MIDDINGLS	POWER (CAPTIVE)	0.462	2011	Financial Year (Apr - Mar), 2011								
9	2011	WASHED COAL	CEMENT	0.045	2011	Financial Year (Apr - Mar), 2011								
10	2011	RAW COAL	STEEL (BOILERS)	0.121	2011	Financial Year (Apr - Mar), 2011								
11	2011	RAW COAL	POWER (UTILITY)	34.997	2011	Financial Year (Apr - Mar), 2011								

3x1 Series

StateName	0
Coal Consuming Sectors	0
Coal Consumed	0

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I14 : PY df.dropna().head()

srcYear	Types of coal	Coal Consuming Sectors	Coal Consumed	YearCode	Year
2011	RAW COAL	OTHER	0.131	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	POWER (UTILITY)	12.545	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	POWER (CAPTIVE)	3.363	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	PULP & PAPER	0.095	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	FERTILISERS	1.111	2011	Financial Year (Apr - Mar), 2011
2011	WASHED COAL	POWER (UTILITY)	1.599	2011	Financial Year (Apr - Mar), 2011
2011	MIDDLELINGS	POWER (CAPTIVE)	0.462	2011	Financial Year (Apr - Mar), 2011
2011	WASHED COAL	CEMENT	0.045	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	STEEL (BOILERS)	0.121	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	POWER (UTILITY)	34.997	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	SPONGE IRON	0.177	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	POWER (CAPTIVE)	11.644	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	TEXTILES & RAYONS	0.023	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	CHEMICAL	0.021	2011	Financial Year (Apr - Mar), 2011
2011	WASHED COAL	OTHER	0.102	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	SPONGE IRON	0.821	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	BRICKS	0.159	2011	Financial Year (Apr - Mar), 2011

5x3 DataFrame

	StateName	Coal Consuming Sectors	Coal Consumed
0	DELHI	OTHER	0.131
1	PUNJAB	POWER (UTILITY)	12.545
2	ANDHRA PRADESH	POWER (CAPTIVE)	3.363
3	ASSAM	PULP & PAPER	0.095
4	PUNJAB	FERTILISERS	1.111

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I16 : PY df['Coal Consuming Sectors'].value\_counts()  
df1.head()

srcYear	Types of coal	Coal Consuming Sectors	Coal Consumed	YearCode	Year
2011	RAW COAL	OTHER	0.131	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	POWER (UTILITY)	12.545	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	POWER (CAPTIVE)	3.363	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	PULP & PAPER	0.095	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	FERTILISERS	1.111	2011	Financial Year (Apr - Mar), 2011
2011	WASHED COAL	POWER (UTILITY)	1.599	2011	Financial Year (Apr - Mar), 2011
2011	MIDDLELINGS	POWER (CAPTIVE)	0.462	2011	Financial Year (Apr - Mar), 2011
2011	WASHED COAL	CEMENT	0.045	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	STEEL (BOILERS)	0.121	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	POWER (UTILITY)	34.997	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	SPONGE IRON	0.177	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	POWER (CAPTIVE)	11.644	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	TEXTILES & RAYONS	0.023	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	CHEMICAL	0.021	2011	Financial Year (Apr - Mar), 2011
2011	WASHED COAL	OTHER	0.102	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	SPONGE IRON	0.821	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	BRICKS	0.159	2011	Financial Year (Apr - Mar), 2011

5x1 Series

Coal Consuming Sectors	count
POWER (UTILITY)	3
POWER (CAPTIVE)	2
OTHER	1
PULP & PAPER	1
FERTILISERS	1

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I18 : PY value\_numbers = df['Coal Consumed'].value\_counts()  
value\_numbers.head(10)

srcYear	Types of coal	Coal Consuming Sectors	Coal Consumed	YearCode	Year
2011	RAW COAL	OTHER	0.131	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	POWER (UTILITY)	12.545	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	POWER (CAPTIVE)	3.363	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	PULP & PAPER	0.095	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	FERTILISERS	1.111	2011	Financial Year (Apr - Mar), 2011
2011	WASHED COAL	POWER (UTILITY)	1.599	2011	Financial Year (Apr - Mar), 2011
2011	MIDDLELINGS	POWER (CAPTIVE)	0.462	2011	Financial Year (Apr - Mar), 2011
2011	WASHED COAL	CEMENT	0.045	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	STEEL (BOILERS)	0.121	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	POWER (UTILITY)	34.997	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	SPONGE IRON	0.177	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	POWER (CAPTIVE)	11.644	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	TEXTILES & RAYONS	0.023	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	CHEMICAL	0.021	2011	Financial Year (Apr - Mar), 2011
2011	WASHED COAL	OTHER	0.102	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	SPONGE IRON	0.821	2011	Financial Year (Apr - Mar), 2011
2011	RAW COAL	BRICKS	0.159	2011	Financial Year (Apr - Mar), 2011

10x1 Series

Coal Consumed	count
0.131	1
12.545	1
3.363	1
0.095	1
1.111	1
1.599	1
0.462	1
0.045	1
0.121	1
34.997	1

COAL CONSUMPTION (F).xlsx • Saved to this PC

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L2 PY df.head(10)

Year	G	H	I	J	K	L	M	N	O	P	Q	R	S	T	U	V	W	X	Y
1 Year																			
2 Financial Year (Apr - Mar), 2011						[r] DataFrame													
3 Financial Year (Apr - Mar), 2011							[r] DataFrame												
4 Financial Year (Apr - Mar), 2011							[r] DataFrame												
5 Financial Year (Apr - Mar), 2011							[r] DataFrame												
6 Financial Year (Apr - Mar), 2011							[r] DataFrame												
7 Financial Year (Apr - Mar), 2011							[r] tuple												
8 Financial Year (Apr - Mar), 2011							[r] 60												
9 Financial Year (Apr - Mar), 2011							[r] method												
10 Financial Year (Apr - Mar), 2011							[r] DataFrame												
11 Financial Year (Apr - Mar), 2011							[r] Series												
12 Financial Year (Apr - Mar), 2011								[r] Image											
13 Financial Year (Apr - Mar), 2011									[r] FALSE										
14 Financial Year (Apr - Mar), 2011									[r] DataFrame										
15 Financial Year (Apr - Mar), 2011										[r] Image									
16 Financial Year (Apr - Mar), 2011											[r] Series								
17 Financial Year (Apr - Mar), 2011												[r] Series							
18 Financial Year (Apr - Mar), 2011													[r] Series						
19 Financial Year (Apr - Mar), 2011														[r] Series					

10x7 DataFrame

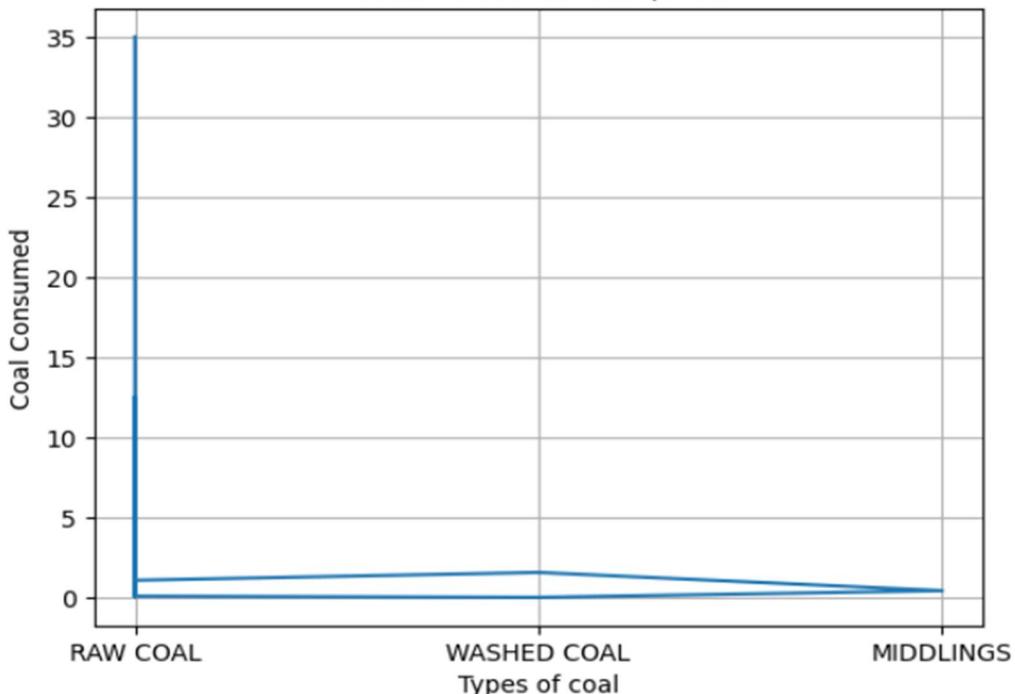
srcStateName	srcY...	Types of coal	Coal Consuming Sect...	Coal Consu...	YearC...	Year
0 DELHI	2011	RAW COAL	OTHER	0.131	2011	Financial Year (Apr - Mar), 2011
1 PUNJAB	2011	RAW COAL	POWER (UTILITY)	12.545	2011	Financial Year (Apr - Mar), 2011
2 ANDHRA PRA...	2011	RAW COAL	POWER (CAPTIVE)	3.363	2011	Financial Year (Apr - Mar), 2011
3 ASSAM	2011	RAW COAL	PULP & PAPER	0.095	2011	Financial Year (Apr - Mar), 2011
4 PUNJAB	2011	RAW COAL	FERTILISERS	1.111	2011	Financial Year (Apr - Mar), 2011
5 DELHI	2011	WASHED C...	POWER (UTILITY)	1.599	2011	Financial Year (Apr - Mar), 2011
6 WEST BENGAL	2011	MIDLINGS	POWER (CAPTIVE)	0.462	2011	Financial Year (Apr - Mar), 2011
7 KARNATAKA	2011	WASHED C...	CEMENT	0.045	2011	Financial Year (Apr - Mar), 2011
8 ANDHRA PRA...	2011	RAW COAL	STEEL (BOILERS)	0.121	2011	Financial Year (Apr - Mar), 2011
9 MAHARASHT...	2011	RAW COAL	POWER (UTILITY)	34.997	2011	Financial Year (Apr - Mar), 2011

```

PY x = np.array(['RAW COAL','RAW COAL','RAW COAL','RAW COAL','RAW COAL','WASHED COAL','MIDLINGS','WASHED COAL','RAW COAL','RAW COAL'])
y = np.array([0.131,12.545,3.363,0.095,1.111,1.599,0.462,0.045,0.121,34.997])
plt.title("Indian Coal Consumption")
plt.xlabel("Types of coal")
plt.ylabel("Coal Consumed")
plt.plot(x, y)
plt.grid()
plt.show()

```

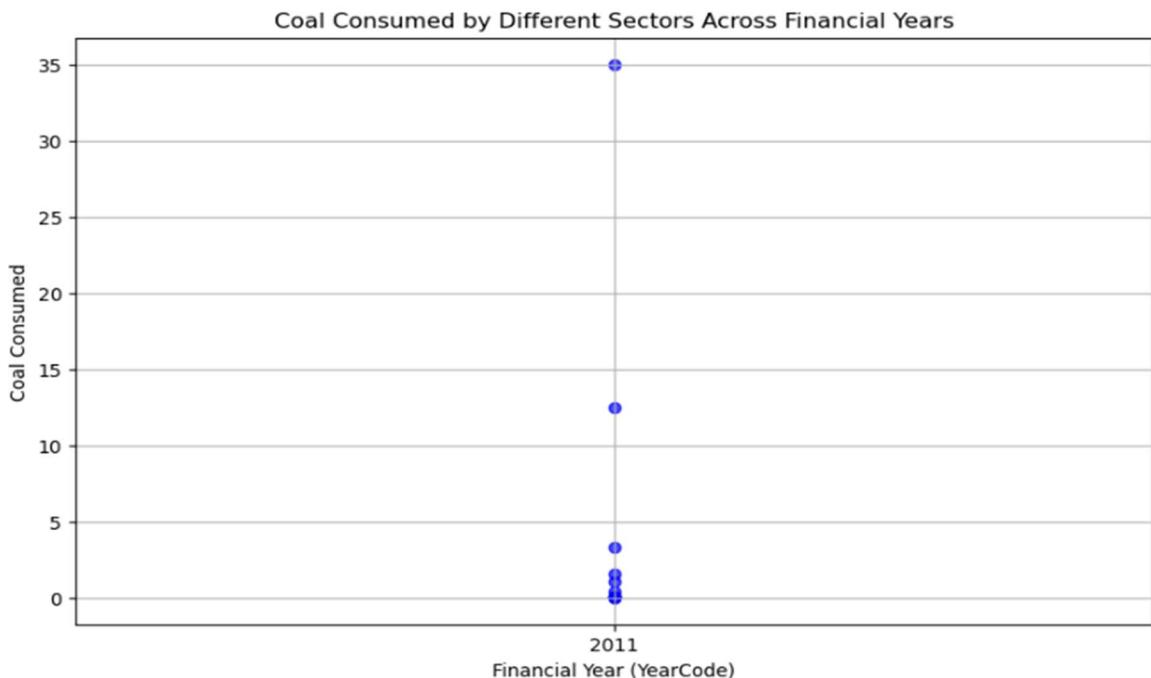
Indian Coal Consumption



```

PY df = pd.DataFrame({
    'src StateName': ['DELHI', 'PUNJAB', 'ANDHRA PRADESH', 'ASSAM', 'PUNJAB', 'DELHI', 'WEST BENGAL', 'KARNATAKA', 'ANDHRA PRADESH',
    'MAHARASHTRA'],
    'srcYear': [2011, 2011, 2011, 2011, 2011, 2011, 2011, 2011, 2011],
    'Types of coal': ['RAW COAL', 'RAW COAL', 'RAW COAL', 'RAW COAL', 'RAW COAL', 'WASHED COAL', 'MIDLINGS', 'WASHED COAL', 'RAW COAL',
    'RAW COAL'],
    'Coal Consuming Sectors': ['OTHER', 'POWER (UTILITY)', 'POWER (CAPTIVE)', 'PULP & PAPER', 'FERTILISERS', 'POWER (UTILITY)', 'POWER (CAPTIVE)',
    'CEMENT', 'STEEL (BOILERS)', 'POWER (UTILITY)'],
    'Coal Consumed': [0.131, 12.545, 3.363, 0.095, 1.111, 1.599, 0.462, 0.045, 0.121, 34.997],
    'YearCode': ['2011', '2011', '2011', '2011', '2011', '2011', '2011', '2011', '2011', '2011']
})
plt.figure(figsize=(10, 6))
plt.scatter(df['YearCode'], df['Coal Consumed'], c='blue', alpha=0.7)
plt.xlabel('Financial Year (YearCode)')
plt.ylabel('Coal Consumed')
plt.title('Coal Consumed by Different Sectors Across Financial Years')
plt.grid(True)
plt.show()

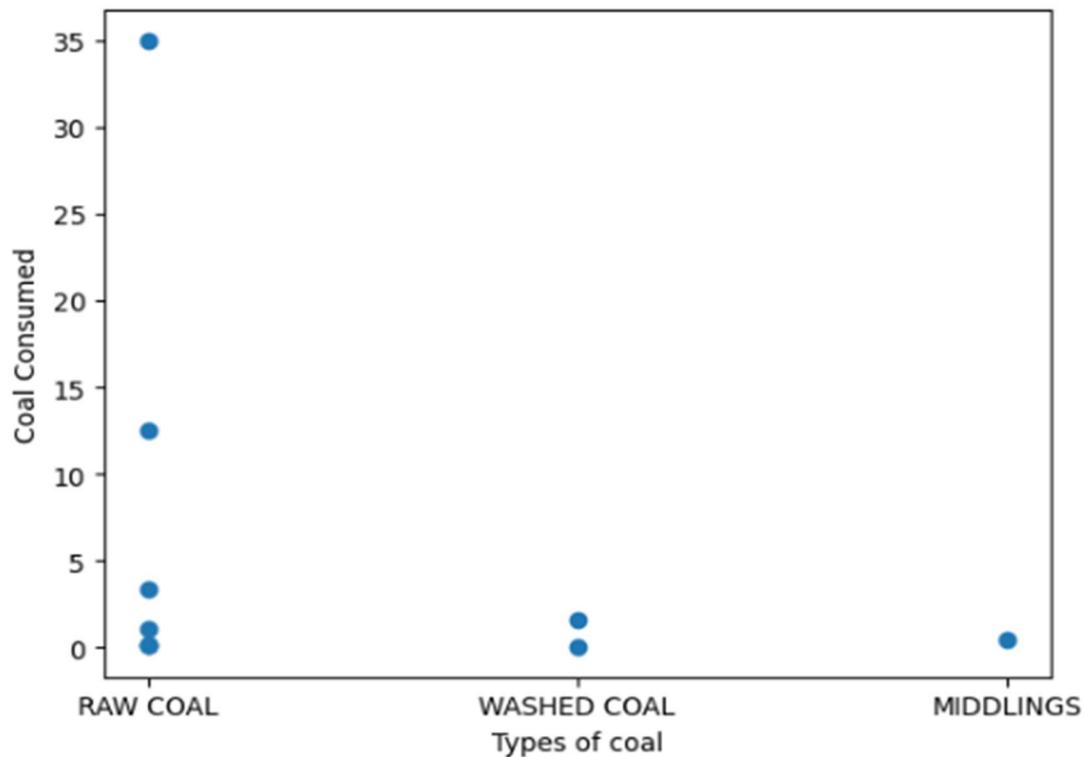
```



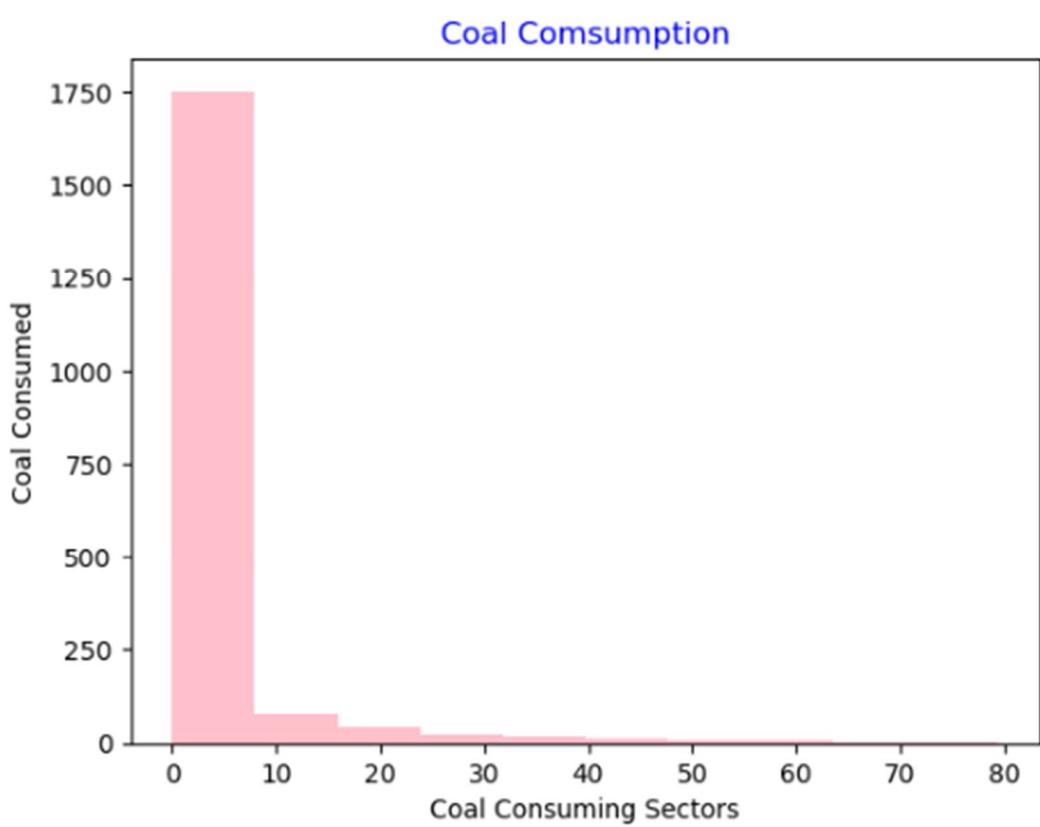
```

PY x = np.array(['RAW COAL', 'RAW COAL', 'RAW COAL', 'RAW COAL', 'RAW COAL', 'WASHED COAL', 'MIDLINGS', 'WASHED COAL', 'RAW COAL', 'RAW COAL'])
y = np.array([0.131, 12.545, 3.363, 0.095, 1.111, 1.599, 0.462, 0.045, 0.121, 34.997])
plt.xlabel('Types of coal')
plt.ylabel('Coal Consumed')
plt.scatter(x, y)
plt.show()

```



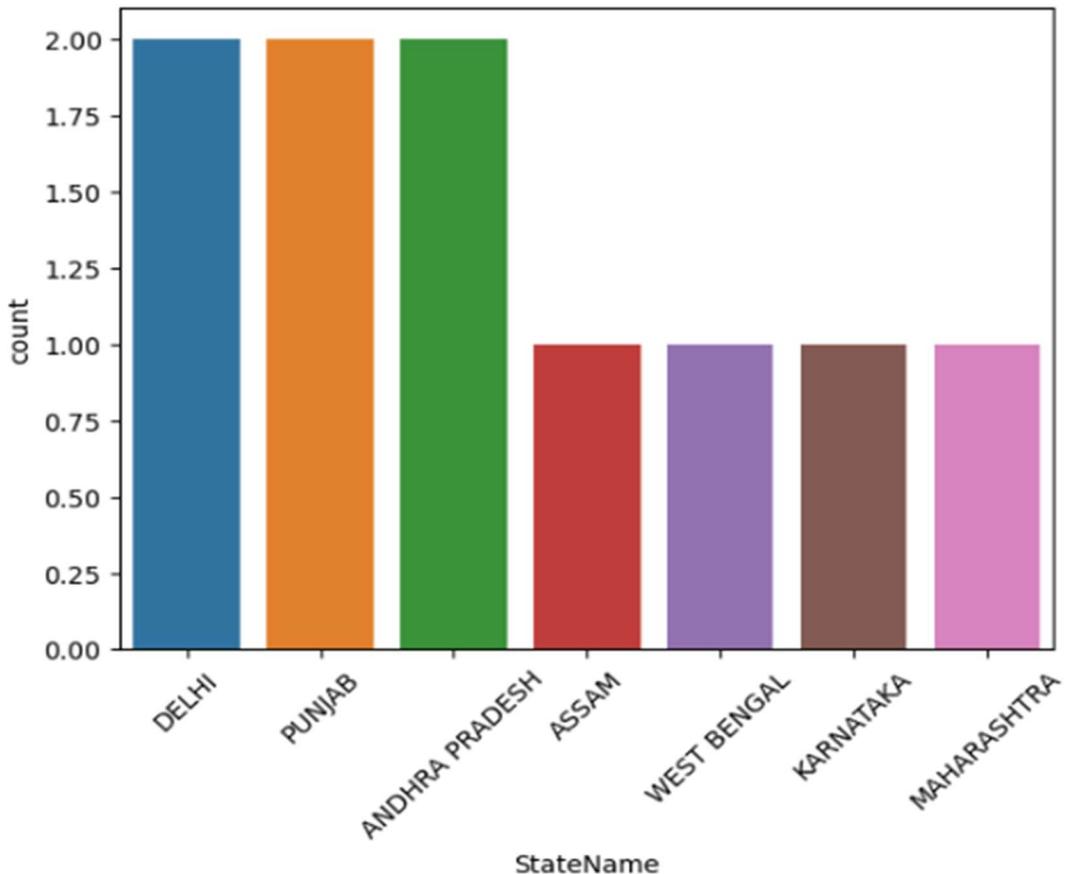
```
plt.hist(df['Coal Consumed'],color = "pink")
plt.title("Coal Consumption",color = 'blue')
plt.xlabel('Coal Consuming Sectors')
plt.ylabel('Coal Consumed')
plt.show()
```



```

PY data = {'StateName': ['DELHI', 'PUNJAB', 'ANDHRA PRADESH', 'ASSAM', 'PUNJAB', 'DELHI', 'WEST BENGAL', 'KARNATAKA', 'ANDHRA PRADESH',
'MAHARASHTRA'],
          'Coal Consuming Sectors': ['OTHER', 'POWER (UTILITY)', 'POWER (CAPTIVE)', 'PULP & PAPER', 'FERTILISERS', 'POWER (UTILITY)',
'POWER (CAPTIVE)', 'CEMENT', 'STEEL (BOILERS)', 'POWER (UTILITY)']}
df = pd.DataFrame(data)
sns.countplot(x = "StateName", data=df)
plt.xticks(rotation=45)
plt.show()

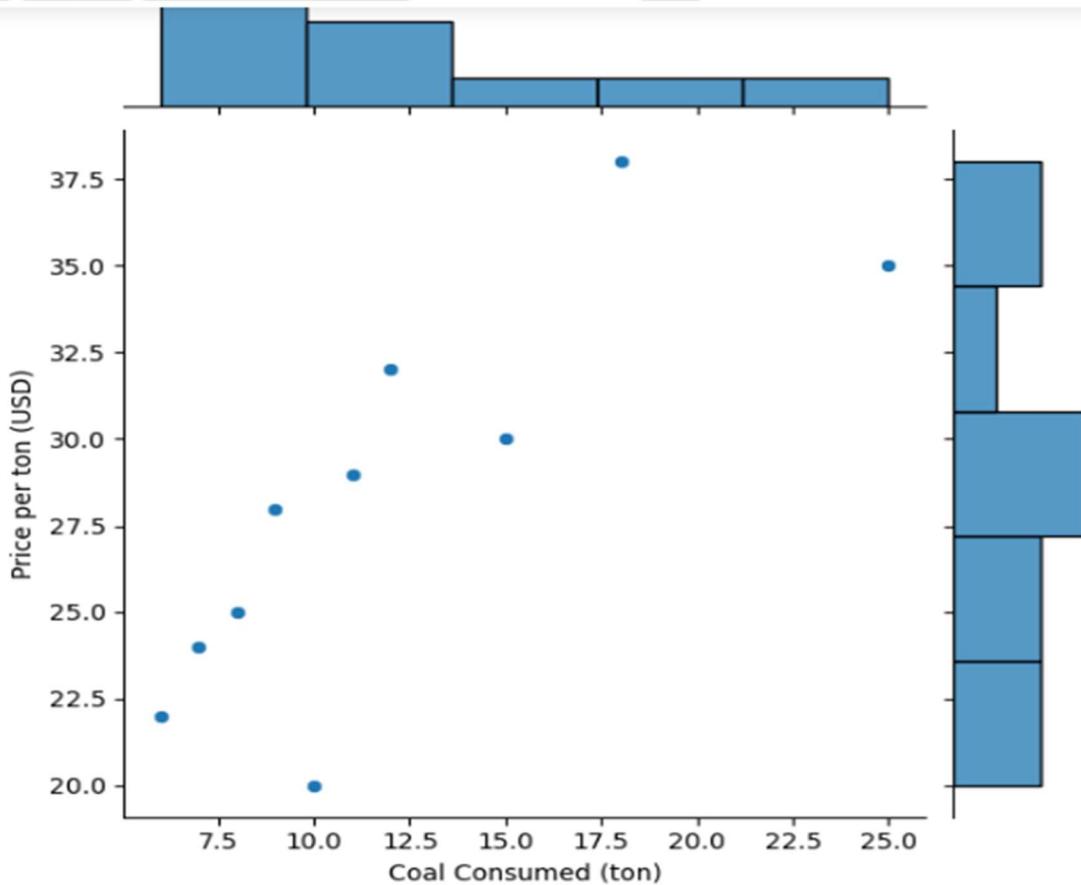
```



```

PY data = {'Coal Consumed (ton)': [10, 15, 8, 12, 9, 25, 6, 18, 11, 7],
          'Price per ton (USD)': [20, 30, 25, 32, 28, 35, 22, 38, 29, 24]}
df = pd.DataFrame(data)
sns.jointplot(x = "Coal Consumed (ton)", y = "Price per ton (USD)", data=df)
plt.show()

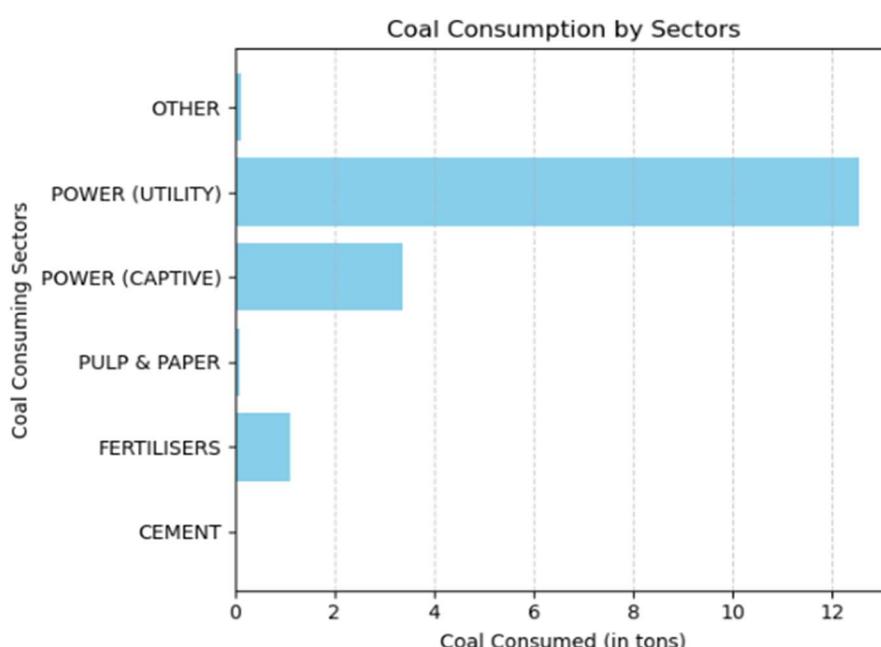
```



```

PY coal_sectors = ['OTHER', 'POWER (UTILITY)', 'POWER (CAPTIVE)', 'PULP & PAPER', 'FERTILISERS', 'CEMENT']
coal_consumption = [0.131, 12.545, 3.363, 0.095, 1.111, 0.045]
plt.barh(coal_sectors, coal_consumption, color='skyblue')
plt.xlabel('Coal Consumed (in tons)')
plt.ylabel('Coal Consuming Sectors')
plt.title('Coal Consumption by Sectors')
plt.gca().invert_yaxis()
plt.grid(axis='x', linestyle='--', alpha=0.6)
plt.tight_layout()
plt.show()

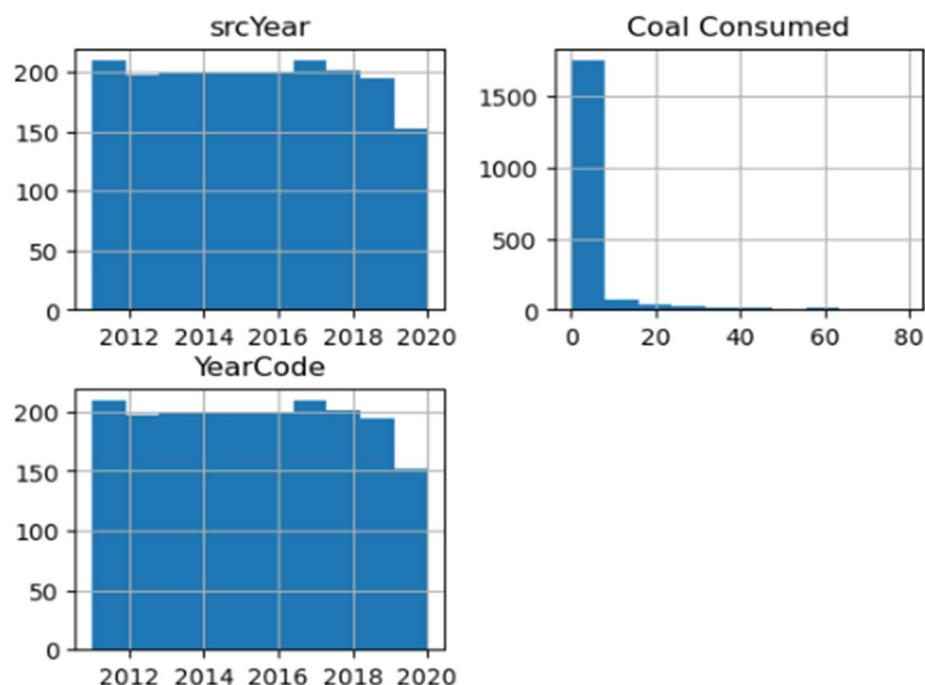
```



```

PY df.hist(grid=True)
plt.title("Indian Coal Consumption")
plt.show()

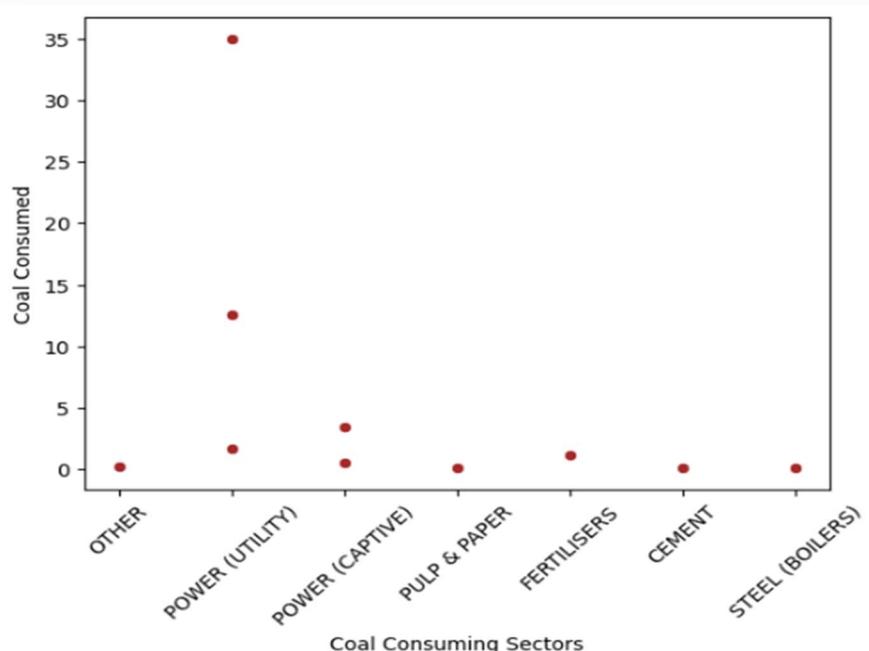
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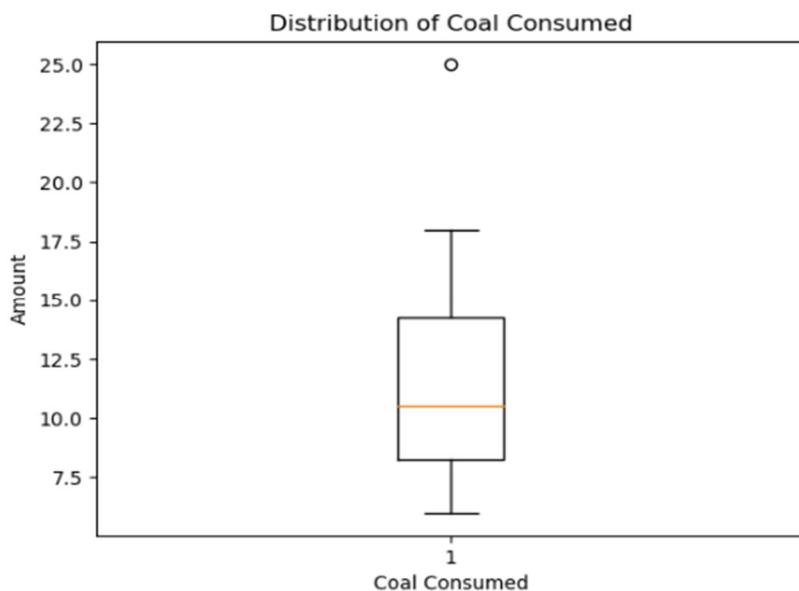
```

PY data = {'StateName': ['DELHI', 'PUNJAB', 'ANDHRA PRADESH', 'ASSAM', 'PUNJAB', 'DELHI', 'WEST BENGAL', 'KARNATAKA', 'ANDHRA PRADESH',
'MAHARASHTRA'],
'Coal Consuming Sectors': ['OTHER', 'POWER (UTILITY)', 'POWER (CAPTIVE)', 'PULP & PAPER', 'FERTILISERS', 'POWER (UTILITY)',
'POWER (CAPTIVE)', 'CEMENT', 'STEEL (BOILERS)', 'POWER (UTILITY)'],
'Coal Consumed': [0.131, 12.545, 3.363, 0.095, 1.111, 1.599, 0.462, 0.045, 0.121, 34.997]}
df = pd.DataFrame(data)
sns.scatterplot(x = "Coal Consuming Sectors", y = "Coal Consumed", data=df,color = 'brown')
plt.xticks(rotation=45)
plt.show()

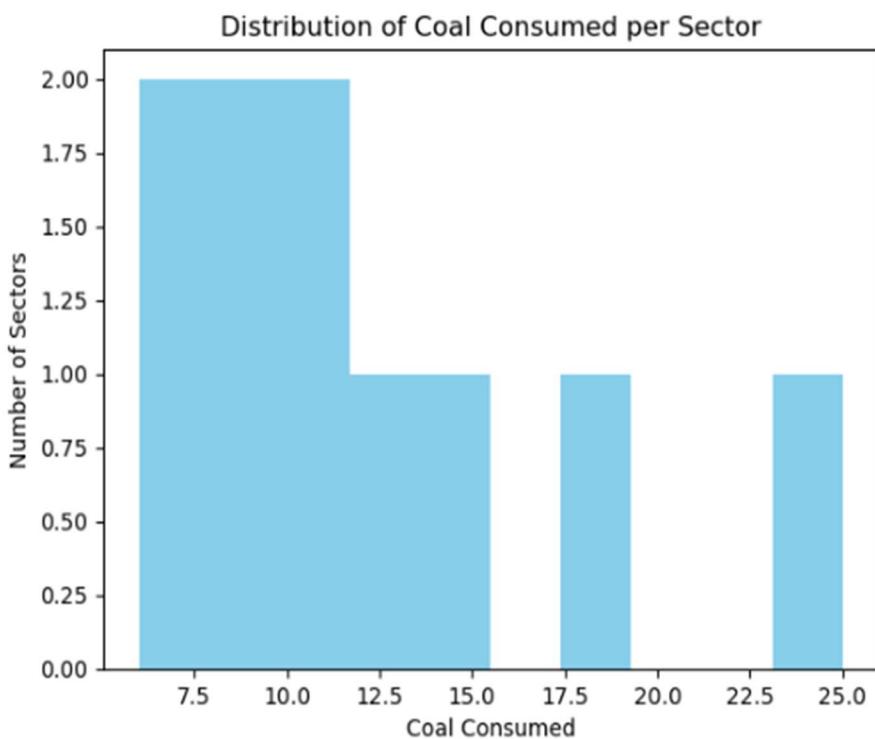
```



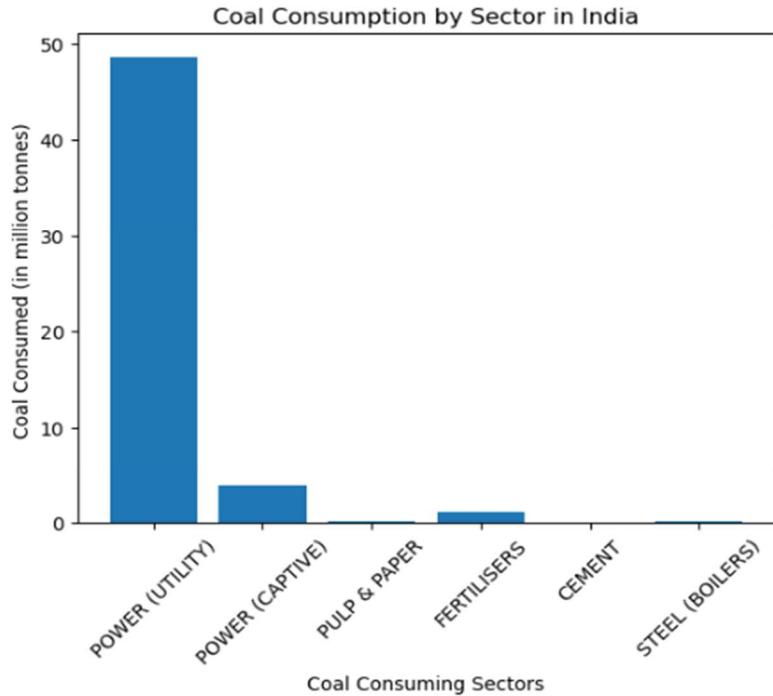
```
PY coal_consumed = [10, 15, 8, 12, 9, 25, 6, 18, 11, 7]
plt.boxplot(coal_consumed)
plt.xlabel('Coal Consumed')
plt.ylabel('Amount')
plt.title('Distribution of Coal Consumed')
plt.show()
```



```
PY coal_consumed_per_sector = [10, 15, 8, 12, 9, 25, 6, 18, 11, 7]
plt.hist(coal_consumed_per_sector,color = 'skyblue')
plt.xlabel('Coal Consumed')
plt.ylabel('Number of Sectors')
plt.title('Distribution of Coal Consumed per Sector')
plt.show()
```



```
py sectors = ['POWER (UTILITY)', 'POWER (CAPTIVE)', 'PULP & PAPER', 'FERTILISERS', 'CEMENT', 'STEEL (BOILERS)']
coal_consumption = [48.542, 3.825, 0.095, 1.111, 0.045, 0.121]
plt.bar(sectors, coal_consumption)
plt.xlabel('Coal Consuming Sectors')
plt.ylabel('Coal Consumed (in million tonnes)')
plt.title('Coal Consumption by Sector in India')
plt.xticks(rotation=45)
plt.show()
```



## RESULT

The Result of this Coal Prices Analysis project is a portal that depends on the data available and the focus of the analysis . Effective data visualization helps stakeholders make informed decisions based on clear and comprehensible insights from the data . Data visualization in coal prices analysis for Bharat Coking Coal Limited (BCCL) typically provides insights into various aspects of the coal market . Some potential results have might be obtain that is Trends Over Time , Price Distribution , Geographical Insights , Market Factors , Cost Analysis , Demand and Supply Analysis . Analysing coal prices is essential for BCCL to navigate the complex and dynamic coal market. By understanding the factors influencing coal prices and utilizing robust analytical methods, BCCL can make informed decisions that enhance its competitiveness and financial performance. This analysis will support strategic planning, operational efficiency, and market positioning, ensuring the company's long-term success in the coal industry.

As for Coal Consumption project that help stakeholders understand consumption patterns, identify trends, and make informed decisions about production, distribution, and policy measures . It can provide valuable insights into how coal is consumed across different sectors, regions, and over time . Some potential results and insights that can be obtained that is Consumption Trends Over Time , Sector-wise Consumption , Geographical Insights , Industrial Consumption Analysis , Environmental Impact , Cost Analysis , Policy and Regulation Impact , Demand-Supply Balance . Coal consumption analysis is a vital component of BCCL's efforts to enhance operational efficiency, manage costs, and reduce environmental impact. By leveraging advanced data analytics and predictive modelling, BCCL can achieve significant improvements in its coal consumption practices, contributing to sustainable and responsible mining operations.

## DECLARATION

We here by declare that the work , which is being presented in the project , titled “ **COAL PRICES ANALYSIS AND COAL CONSUMPTION** ” submitted by me [ PRAKSHI PRIYA ] to BCCL , Dhanbad and “ **SIKSHA ‘O’ ANUSANDHAN University** , Bhubaneswar fulfilment of the requirement for the award of Degree of “ **BACHELOR’S OF TECHNOLOGY** ” in Department of Computer Science and Engineering is a record of Internship Project and our own investigation carried under the guidance of **MR. AVINASH KUMAR NANDAN** , System Department BCCL , Dhanbad . The information has been collected from genuine and authentic sources . The work has been submitted in fulfilment of the requirement of **BACHELOR’S OF TECHNOLOGY** .

NAME : PRAKSHI PRIYA

BRANCH : CSE [ DATA SCIENCE ]

REG NO : 2241016464

