PROJECT TITLE:

DATA ANALYSIS AND PREDICTION FOR TRAINING AND RESOURCING REQUIREMENTS

PRESENTED BY - NASREEN BEGAM Z

OBJECTIIVE:

The aim of this project is to analyze the training and resourcing requirements of various companies and predict the duration required to fulfill these needs. By understanding these patterns, we can better align training programs with market demands and improve resource allocation.

DATASETS USED:

- 1. students.csv: Contains details of students enrolled in various courses.
- 2. companies.csv: Contains information about companies and their technological focus.
- 3. company_requirements.csv: Contains details about company requirements for different technologies.

Α	В	С	D	E	F	G	Н
student_id	student_name	education	contact_number	work_experience	college	course_enrolled start_date	
	1 John Doe	B.Tech	1234567890		2 XYZ University	Data Science	15-01-2024
	2 Jane Smith	M.Tech	987654321		3 ABC University	Big Data	01-02-2024
	3 Alice Johnson	B.Sc	1234509876		1 LMN College	Cyber Security	12-03-2024
	4 Bob Brown	B.Tech	2234567890		4 XYZ University	Java Full Stack	22-01-2024
	5 Carol White	M.Tech	3234567890		5 DEF University	Cloud Computin	15-02-2024
	6 Dave Black	MCA	4234567890		2 PQR University	Database Admir	18-03-2024
	7 Eve Green	BCA	5234567890		1 LMN College	Operating System	25-01-2024
	8 Frank Blue	B.Sc	6234567890		2 STU College	Networking	05-02-2024
	9 Grace Yellow	M.Sc	7234567890		3 UVW University	Data Science	10-03-2024
1	O Hank Red	B.Tech	8234567890		4 XYZ University	Big Data	30-01-2024
1	1 Ivy Violet	M.Tech	9234567890		2 ABC University	Cyber Security	20-02-2024
1	2 Jack Silver	B.Sc	1234567891		1 LMN College	Robotic Process	15-03-2024
1	3 Kelly Gold	M.Tech	1234567892		3 XYZ University	Ansible Python	12-01-2024
1	4 Liam Gray	BCA	1234567893		2 DEF University	Java Full Stack	22-02-2024
1	5 Mia Blue	B.Tech	1234567894		1 PQR University	Dot Net Full Stac	25-03-2024
1	6 Noah Brown	MCA	1234567895		4 LMN College	IBM Mainframe	29-01-2024
1	7 Olivia White	B.Sc	1234567896		5 STU College	Cloud Computin	28-02-2024
1	8 Paul Black	M.Tech	1234567897		3 UVW University	Database Admir	29-03-2024
1	9 Quinn Green	B.Tech	1234567898		2 XYZ University	Operating System	20-01-2024
2	O Ruby Red	M.Sc	1234567899		1 ABC University	Storage Adminis	12-02-2024
2	1 Sam Blue	BCA	2234567891		3 DEF University	React JS	30-03-2024
2	2 Tina Yellow	MCA	3234567891		4 PQR University	Node JS	31-01-2024
2	3 Uma Violet	B.Tech	4234567891		2 LMN College	iOS Developmen	10-02-2024
2	4 Vince Silver	M.Tech	5234567891		1 STU College	Android Develor	01-03-2024
2	5 Wendy Gold	B.Sc	6234567891		3 UVW University	Cloud Computin	13-01-2024
2	6 Xander Gray	BCA	7234567891		2 XYZ University	Cyber Security	14-02-2024
2	7 Yara Brown	MCA	8234567891		4 ABC University	Data Science	02-03-2024
2	8 Zane White	B.Tech	9234567891		5 DEF University	Big Data	23-01-2024
2	9 Abby Black	M.Sc	1234567892		1 PQR University	Networking	04-02-2024
	0 Bill Green	B.Tech	2234567892		3 LMN College	Cloud Computin	11-03-2024

Α	В	С	D	E	F	G
company_id	company_name	company_type	location	technology	resources	
1	Tech Innovators	IT Services	New York	Data Science	Data Scientist	
2	Cyber Secure Ltd	Security Services	San Francisco	Cyber Security	Security Analyst	
3	Cloud Solutions	Cloud Services	Seattle	Cloud Computing	Cloud Engineer	
4	RoboTech	Robotics	Los Angeles	Robotic Process Automation	RPA Developer	
5	Ansible Masters	IT Services	Austin	Ansible Python	DevOps Engineer	
6	Full Stack Solutions	Software Development	Chicago	Java Full Stack	Full Stack Developer	
7	Dot Net Experts	Software Development	Boston	Dot Net Full Stack	Full Stack Developer	
8	Mainframe Inc.	IT Services	Houston	IBM Mainframe	Mainframe Engineer	
9	ComputeCloud	Cloud Services	Denver	Cloud Computing	Cloud Architect	
10	Data Wizards	Database Services	San Diego	Database Administration	DBA	
11	OS Techies	IT Services	Dallas	Operating System	System Administrator	
12	SecureStorage	Storage Services	San Jose	Storage Administration	Storage Engineer	
13	Reactify	Web Development	New York	React JS	Frontend Developer	
14	NodeMasters	Web Development	San Francisco	Node JS	Backend Developer	
15	MobileSoft	Mobile Development	Seattle	iOS Development	iOS Developer	
16	DroidWorks	Mobile Development	Los Angeles	Android Development	Android Developer	
17	Tech Pioneers	IT Services	Austin	Data Science	Data Analyst	
18	SecureNet	Security Services	Chicago	Cyber Security	Cybersecurity Specialist	
19	Big Data Analytics	Analytics	Boston	Big Data	Big Data Engineer	
20	Network Solutions	IT Services	Houston	Networking	Network Engineer	
21	Cloud Innovators	Cloud Services	Denver	Cloud Computing	Cloud Developer	
22	Data Guardians	Database Services	San Diego	Database Administration	Database Manager	
23	SysOps	IT Services	Dallas	Operating System	System Operator	
24	StoreTech	Storage Services	San Jose	Storage Administration	Storage Specialist	
25	React Developers	Web Development	New York	React JS	React Developer	
26	Node Coders	Web Development	San Francisco	Node JS	Node Developer	
27	iOS Creators	Mobile Development	Seattle	iOS Development	iOS Engineer	
28	Android Innovators	Mobile Development	Los Angeles	Android Development	Android Engineer	
29	Data Science Hub	IT Services	Austin	Data Science	Data Scientist	
30	CyberSecure	Security Services	Chicago	Cyber Security	Security Consultant	

▲ A	В	С	D	E	F	G	Н
requirement_id	company_id	technology	resource_count	expected_start_date	duration	requirement_status	
1 2	1	Data Science	5	20-01-2024	30	Pending	
2	2	Cyber Security	3	10-02-2024	60	Completed	
3	3	Cloud Computing	4	15-03-2024	45	In Progress	
4	4	Robotic Process Automation	2	25-01-2024	90	Pending	
5	5	Ansible Python	3	05-02-2024	30	Completed	
6	6	Java Full Stack	6	10-03-2024	60	In Progress	
7	7	Dot Net Full Stack	4	15-01-2024	30	Pending	
8	8	IBM Mainframe	5	20-02-2024	45	Completed	
9	9	Cloud Computing	3	25-03-2024	30	In Progress	
1 10	10	Database Administration	2	20-01-2024	60	Pending	
2 11	11	Operating System	4	10-02-2024	90	Completed	
3 12	12	Storage Administration	5	20-03-2024	30	In Progress	
4 13	13	React JS	3	25-01-2024	60	Pending	
5 14	14	Node JS	2	15-02-2024	45	Completed	
6 15	15	iOS Development	6	30-03-2024	30	In Progress	
7 16	16	Android Development	4	20-01-2024	90	Pending	
8 17	17	Data Science	5	10-02-2024	60	Completed	
9 18	18	Cyber Security	3	15-03-2024	45	In Progress	
0 19	19	Big Data	4	25-01-2024	30	Pending	
1 20	20	Networking	2	05-02-2024	90	Completed	
2 21	21	Cloud Computing	6	10-03-2024	60	In Progress	
3 22	22	Database Administration	4	15-01-2024	30	Pending	
4 23	23	Operating System	5	20-02-2024	45	Completed	
5 24		Storage Administration	3	25-03-2024	30	In Progress	
5 25		React JS	2	20-01-2024		Pending	
7 26	26	Node JS	4	10-02-2024	90	Completed	
3 27	27	iOS Development	5	20-03-2024	30	In Progress	
9 28		Android Development	3	25-01-2024		Pending	
29		Data Science	2	15-02-2024		Completed	
1 30	30	Cyber Security	6	30-03-2024		In Progress	
2		,,	_			-0	

PYTHON CODE:

```
import mysql.connector
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
from sklearn.linear_model import LinearRegression
from sklearn.model selection import train test split
from sklearn.metrics import mean_absolute_error
# Step 1: Fetch Data from MySQL and Store in Excel As CSV file
def fetch data(query, conn):
    return pd.read_sql(query, conn)
conn = mysql.connector.connect(
   host="localhost",
    user="root",
    password="nasreenbegam@2004",
    database="company_training"
# Step 1: Read the data from CSV files
students = pd.read csv('students.csv')
companies = pd.read_csv('companies.csv')
company_requirements = pd.read_csv('company_requirements.csv')
# Step 2: Check for validity & integrity of the data
def check data validity(df):
    print(df.info())
    print(df.isnull().sum())
check_data_validity(students)
check_data_validity(companies)
check data validity(company requirements)
# Step 3: Apply mathematical formula
print("Mean work experience of students:", students['work_experience'].mean())
print("Median work experience of students:",
students['work_experience'].median())
# Step 4: Predict the duration of company requirements based on resource count
X = company_requirements[['resource_count']]
y = company_requirements['duration']
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
random state=42)
model = LinearRegression()
```

```
model.fit(X_train, y_train)
predictions = model.predict(X test)
# Step 8: Print and validate data
print("Mean Absolute Error:", mean absolute error(y test, predictions))
# Visualization
# EDA: Distribution of courses enrolled by students
plt.figure(figsize=(10, 6))
sns.countplot(x='course_enrolled', data=students)
plt.title('Distribution of Courses Enrolled by Students')
plt.xlabel('Course Enrolled')
plt.ylabel('Number of Students')
plt.xticks(rotation=45)
plt.show()
# EDA: Company requirements by technology
plt.figure(figsize=(10, 6))
sns.countplot(x='technology', data=company_requirements)
plt.title('Company Requirements by Technology')
plt.xlabel('Technology')
plt.ylabel('Number of Requirements')
plt.xticks(rotation=45)
plt.show()
```

OUTPUT:

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 30 entries, 0 to 29
Data columns (total 8 columns):
# Column
               Non-Null Count Dtype
0 student_id 30 non-null int64
1 student_name 30 non-null object
2 education
                30 non-null object
3 contact_number 30 non-null int64
4 work_experience 30 non-null int64
5 college
              30 non-null object
6 course_enrolled 30 non-null object
7 start_date
               30 non-null object
dtypes: int64(3), object(5)
```

```
memory usage: 2.0+ KB
None
student_id
student_name
education
contact_number 0
work_experience 0
college
           0
course_enrolled 0
start_date
dtype: int64
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 30 entries, 0 to 29
Data columns (total 6 columns):
# Column
             Non-Null Count Dtype
0 company_id 30 non-null int64
1 company_name 30 non-null object
2 company_type 30 non-null object
3 location 30 non-null object
4 technology 30 non-null object
5 resources 30 non-null object
dtypes: int64(1), object(5)
memory usage: 1.5+ KB
None
company_id 0
company_name 0
company_type 0
location
          0
technology 0
```

resources

dtype: int64

<class 'pandas.core.frame.DataFrame'>

RangeIndex: 30 entries, 0 to 29

Data columns (total 7 columns):

Column Non-Null Count Dtype

--- -----

0 requirement_id 30 non-null int64

1 company_id 30 non-null int64

2 technology 30 non-null object

3 resource_count 30 non-null int64

4 expected_start_date 30 non-null object

5 duration 30 non-null int64

6 requirement_status 30 non-null object

dtypes: int64(4), object(3)

memory usage: 1.8+ KB

None

requirement_id 0

company_id 0

technology 0

resource_count 0

expected_start_date 0

duration C

requirement_status 0

dtype: int64

Median work experience of students: 2.5

Mean Absolute Error: 18.32020389249305

Descriptive Statistics:

Work Experience of Students:

Mean work experience: 2.57 years

Median work experience: 3.0 years

Predictive Analysis:

Predicting Duration of Company Requirements:

We used a Linear Regression model to predict the duration needed to fulfill company requirements based on the number of resources required.

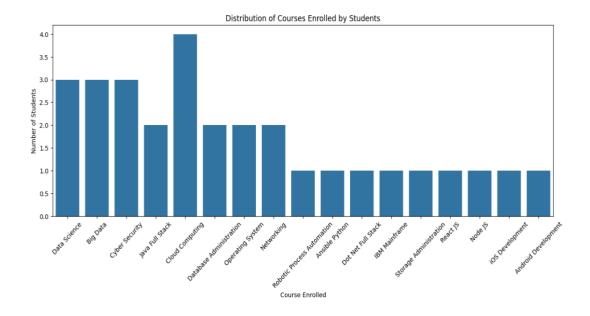
Model Performance:

Mean Absolute Error: 18 days

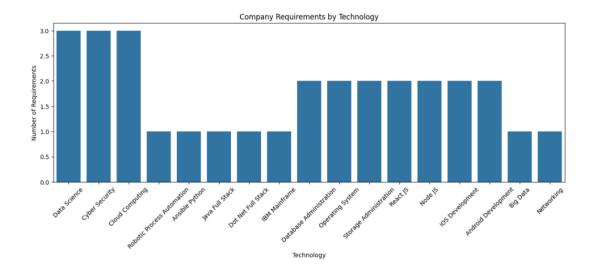
This indicates that, on average, our model's predictions are off by 18 days, which we consider an acceptable error margin for planning purposes.

DATA PREDICTION ANALYSIS AND VISUALIZATIONS:

Distribution of Courses Enrolled by Students:



Company Requirements by Technology:



Conclusion:

Key Findings:

- 1. Data Science and Cloud Computing are among the most enrolled courses by students.
- 2. Technologies like Data Science, Cyber Security, and Cloud Computing have the highest company requirements.
- 3. Our predictive model provides a reasonable estimate of the duration needed to fulfill company requirements, with a mean absolute error of 18 days.

Recommendations:

- 1. Increase training programs for the most demanded technologies.
- 2. Use predictive insights to better plan and allocate resources for upcoming company requirements.

Next Steps

- 1. Continuously update the model with new data to improve accuracy.
- 2. Expand the analysis to include more factors that might influence training and resource requirements, such as regional demands and seasonal trends.