

# Global AI Job Landscape

## Salary Trends Skill Demands 2025

August 4, 2025

```
[321]: import pandas as pd
import matplotlib.pyplot as plt
from collections import Counter
import seaborn as sns
```

```
[322]: df = pd.read_csv('../data/global_ai_job_dataset.csv')
print(df.head())
print(df.info())
```

	job_id	job_title	salary_usd	salary_currency	\
0	AI00001	AI Research Scientist	90376	USD	
1	AI00002	AI Software Engineer	61895	USD	
2	AI00003	AI Specialist	152626	USD	
3	AI00004	NLP Engineer	80215	USD	
4	AI00005	AI Consultant	54624	EUR	

	experience_level	employment_type	company_location	company_size	\
0	SE	CT	China	M	
1	EN	CT	Canada	M	
2	MI	FL	Switzerland	L	
3	SE	FL	India	M	
4	EN	PT	France	S	

	employee_residence	remote_ratio	\
0	China	50	
1	Ireland	100	
2	South Korea	0	
3	India	50	
4	Singapore	100	

	required_skills	education_required	\
0	Tableau, PyTorch, Kubernetes, Linux, NLP	Bachelor	
1	Deep Learning, AWS, Mathematics, Python, Docker	Master	
2	Kubernetes, Deep Learning, Java, Hadoop, NLP	Associate	
3	Scala, SQL, Linux, Python	PhD	
4	MLOps, Java, Tableau, Python	Master	

	years_experience	industry	posting_date	application_deadline	\
0	9	Automotive	2024-10-18	2024-11-07	
1	1	Media	2024-11-20	2025-01-11	
2	2	Education	2025-03-18	2025-04-07	
3	7	Consulting	2024-12-23	2025-02-24	
4	0	Media	2025-04-15	2025-06-23	

	job_description_length	benefits_score	company_name
0	1076	5.9	Smart Analytics
1	1268	5.2	TechCorp Inc
2	1974	9.4	Autonomous Tech
3	1345	8.6	Future Systems
4	1989	6.6	Advanced Robotics

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

RangeIndex: 15000 entries, 0 to 14999

Data columns (total 19 columns):

#	Column	Non-Null Count	Dtype
---	-----	-----	-----
0	job_id	15000 non-null	object
1	job_title	15000 non-null	object
2	salary_usd	15000 non-null	int64
3	salary_currency	15000 non-null	object
4	experience_level	15000 non-null	object
5	employment_type	15000 non-null	object
6	company_location	15000 non-null	object
7	company_size	15000 non-null	object
8	employee_residence	15000 non-null	object
9	remote_ratio	15000 non-null	int64
10	required_skills	15000 non-null	object
11	education_required	15000 non-null	object
12	years_experience	15000 non-null	int64
13	industry	15000 non-null	object
14	posting_date	15000 non-null	object
15	application_deadline	15000 non-null	object
16	job_description_length	15000 non-null	int64
17	benefits_score	15000 non-null	float64
18	company_name	15000 non-null	object

dtypes: float64(1), int64(4), object(14)

memory usage: 2.2+ MB

None

```
[323]: #Remaining rows Israel
df = df[
    (df["company_location"] != "Israel") &
    (df["employee_residence"] != "Israel")
]
```

```
[324]: # Missing values are checked for each column
print("Missing values:\n", df.isna().sum())
```

```
Missing values:
  job_id          0
  job_title       0
  salary_usd      0
  salary_currency 0
  experience_level 0
  employment_type 0
  company_location 0
  company_size     0
  employee_residence 0
  remote_ratio     0
  required_skills  0
  education_required 0
  years_experience 0
  industry         0
  posting_date     0
  application_deadline 0
  job_description_length 0
  benefits_score   0
  company_name     0
dtype: int64
```

```
[325]: def categorize_job(title):
        title = title.lower()
        if "engineer" in title:
            return "Engineering"
        elif "scientist" in title:
            return "Science"
        elif "consultant" in title:
            return "Consulting"
        elif "specialist" in title:
            return "Specialist"
        elif "nlp" in title:
            return "NLP"
        elif "analyst" in title:
            return "Analytics"
        elif "manager" in title:
            return "Management"
        else:
```

```

        return "Other"

# this function will create a new column
df["job_category"] = df["job_title"].apply(categorize_job)

```

```

[326]: #average salary for each experience level
avg_salary_by_exp = df.groupby("experience_level")["salary_usd"].mean()
print(avg_salary_by_exp)

```

```

experience_level
EN      63940.981029
EX     190360.687624
MI      88991.449732
SE     123886.967410
Name: salary_usd, dtype: float64

```

```

[327]: #number of jobs in each job category
job_count_by_title = df["job_title"].value_counts()
print(job_count_by_title.head(10)) # Top 10 job titles

```

```

job_title
Machine Learning Researcher    758
AI Software Engineer           734
Autonomous Systems Engineer    732
Head of AI                     728
AI Architect                   722
Machine Learning Engineer      721
NLP Engineer                   713
Robotics Engineer              708
AI Research Scientist          707
Data Engineer                  702
Name: count, dtype: int64

```

```

[328]: #top 5 countries with the highest average salary
top_salary_countries = df.groupby("company_location")["salary_usd"].mean().
    ↪nlargest(5)
print(top_salary_countries)

```

```

company_location
Switzerland      170393.910448
Denmark          166035.312500
Norway           158995.406470
United States    146894.834270
United Kingdom   128756.876045
Name: salary_usd, dtype: float64

```

```
[329]: #average salary for each employment type
avg_salary_by_employment = df.groupby("employment_type")["salary_usd"].mean()
print(avg_salary_by_employment)
```

```
employment_type
CT    117404.623822
FL    116667.894151
FT    117670.862525
PT    115639.058908
Name: salary_usd, dtype: float64
```

```
[330]: #jobs that are fully remote
remote_jobs_count = len(df[df["remote_ratio"] == 100])
print(f"Fully Remote Jobs: {remote_jobs_count}")
```

```
Fully Remote Jobs: 4617
```

```
[331]: #the highest salary in each job category
max_salary_by_title = df.groupby("job_title")["salary_usd"].max().
    ↪sort_values(ascending=False)
print(max_salary_by_title.head(10))
```

```
job_title
Machine Learning Researcher    399095
AI Architect                    398084
AI Research Scientist          394917
AI Specialist                   390292
Data Scientist                 388754
AI Product Manager             381575
AI Software Engineer           379418
Computer Vision Engineer       366957
Principal Data Scientist       364635
Autonomous Systems Engineer    364585
Name: salary_usd, dtype: int64
```

```
[332]: #average benefits score grouped by company location
avg_benefits_by_location = df.groupby("company_location")["benefits_score"].
    ↪mean()
print(avg_benefits_by_location)
```

```
company_location
Australia    7.518106
Austria      7.518675
Canada       7.563408
China        7.520884
Denmark      7.594271
Finland      7.540966
France       7.510554
Germany      7.485019
```

```
India          7.575034
Ireland        7.390515
Japan          7.466989
Netherlands    7.515110
Norway         7.474684
Singapore      7.490426
South Korea    7.422772
Sweden         7.501877
Switzerland    7.451560
United Kingdom 7.440808
United States  7.566573
Name: benefits_score, dtype: float64
```

```
[333]: df.groupby("experience_level")["salary_usd"].mean().plot(kind="bar")
plt.title("Average Salary by Experience Level")
plt.xlabel("Experience Level")
plt.ylabel("Salary (USD)")
plt.show()
```



```
[334]: #total jobs in each industry
industry_count = df["industry"].value_counts()
print(industry_count)
```

```
industry
Retail          989
Media           978
Automotive      960
Technology      954
Consulting      947
Real Estate     943
Government      938
Transportation  937
Telecommunications 928
Healthcare      926
Energy          923
Finance         923
Education       902
Gaming          900
Manufacturing   895
Name: count, dtype: int64
```

```
[335]: #average job description length per job category
avg_desc_length_by_category = df.
    ↳groupby("job_category")["job_description_length"].mean()
print(avg_desc_length_by_category)
```

```
job_category
Analytics      1462.441261
Consulting     1538.048338
Engineering    1497.678132
Management     1503.218705
Other          1521.726902
Science        1505.388325
Specialist     1496.320175
Name: job_description_length, dtype: float64
```

```
[336]: #how many jobs require a PhD
phd_jobs_count = len(df[df["education_required"] == "PhD"])
print(f"Jobs Requiring PhD: {phd_jobs_count}")
```

```
Jobs Requiring PhD: 3420
```

```
[337]: #average salary by remote ratio
avg_salary_by_remote = df.groupby("remote_ratio")["salary_usd"].mean()
print(avg_salary_by_remote)
```

```
remote_ratio
0      115678.025609
50     117440.283355
100    117465.242582
Name: salary_usd, dtype: float64
```

```
[338]: df.groupby("years_experience")["salary_usd"].mean().plot(kind="line")
plt.title("Average Salary by Years of Experience")
plt.xlabel("Years of Experience")
plt.ylabel("Salary (USD)")
plt.show()
```





```
[339]: #Top 10 companies offering the most jobs
top_companies = df["company_name"].value_counts().nlargest(10)
print(top_companies)
```

```
company_name
TechCorp Inc          933
Digital Transformation LLC  903
AI Innovations        900
Quantum Computing Inc  896
Cognitive Computing    896
Future Systems        894
Cloud AI Solutions     889
Predictive Systems     881
Smart Analytics        872
Neural Networks Co     872
Name: count, dtype: int64
```

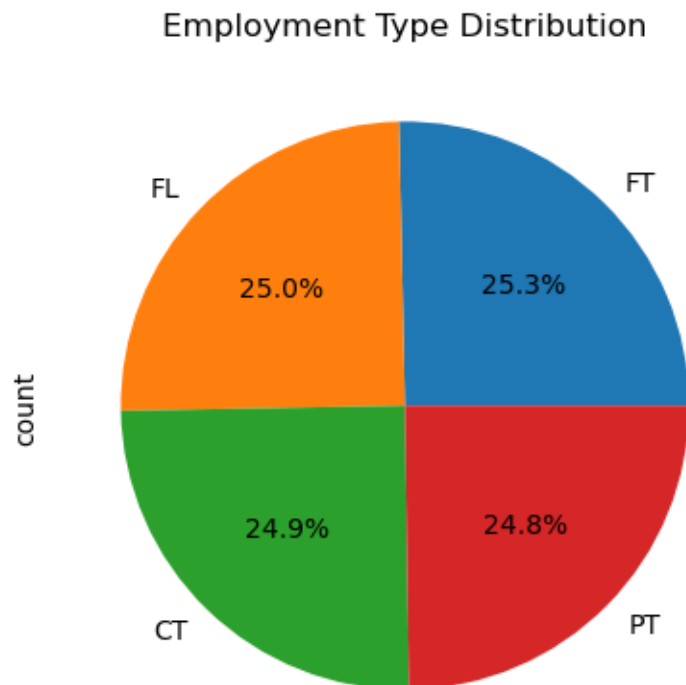
```
[340]: #Average salary by industry
avg_salary_by_industry = df.groupby("industry")["salary_usd"].mean()
print(avg_salary_by_industry)
```

```
industry
Automotive          115564.963542
Consulting          118849.026399
Education           117341.180710
Energy              116456.836403
Finance             116727.957746
Gaming              115472.036667
Government          116644.184435
Healthcare          115322.268898
Manufacturing       117592.871508
Media               117655.751534
Real Estate         117806.036055
Retail              116608.309403
Technology          117053.314465
Telecommunications  117072.004310
Transportation      116545.572038
Name: salary_usd, dtype: float64
```

```
[341]: #total jobs by employee residence  
residence_count = df["employee_residence"].value_counts()  
print(residence_count)
```

```
employee_residence  
Sweden            776  
Germany           765  
Denmark           764  
France            764  
Austria           761  
India             758  
South Korea       753  
Canada            750  
China             746  
Netherlands       744  
United Kingdom    741  
Switzerland       739  
Singapore         732  
Ireland           728  
Australia         715  
Norway            714  
United States     702  
Finland           700  
Japan             691  
Name: count, dtype: int64
```

```
[342]: df["employment_type"].value_counts().plot(kind="pie", autopct='%1.1f%%')
plt.title("Employment Type Distribution")
plt.show()
```



```
[343]: #Minimum salary in each experience level
min_salary_by_exp = df.groupby("experience_level")["salary_usd"].min()
print(min_salary_by_exp)
```

```
experience_level
EN    32519
EX    86560
MI    45900
SE    64824
Name: salary_usd, dtype: int64
```

```
[344]: #Average years of experience by company location
avg_exp_by_location = df.groupby("company_location")["years_experience"].mean()
print(avg_exp_by_location)
```

```
company_location
Australia      6.036212
Austria        6.335099
Canada         6.321004
China          6.064257
Denmark        6.815104
Finland        6.580690
France         6.163588
Germany        6.302122
India          6.112011
Ireland        6.055556
Japan          6.236188
Netherlands    6.217033
Norway         6.403657
Singapore      6.283245
South Korea    6.090523
Sweden         6.276139
Switzerland    5.934871
United Kingdom 6.122563
United States  6.366573
Name: years_experience, dtype: float64
```

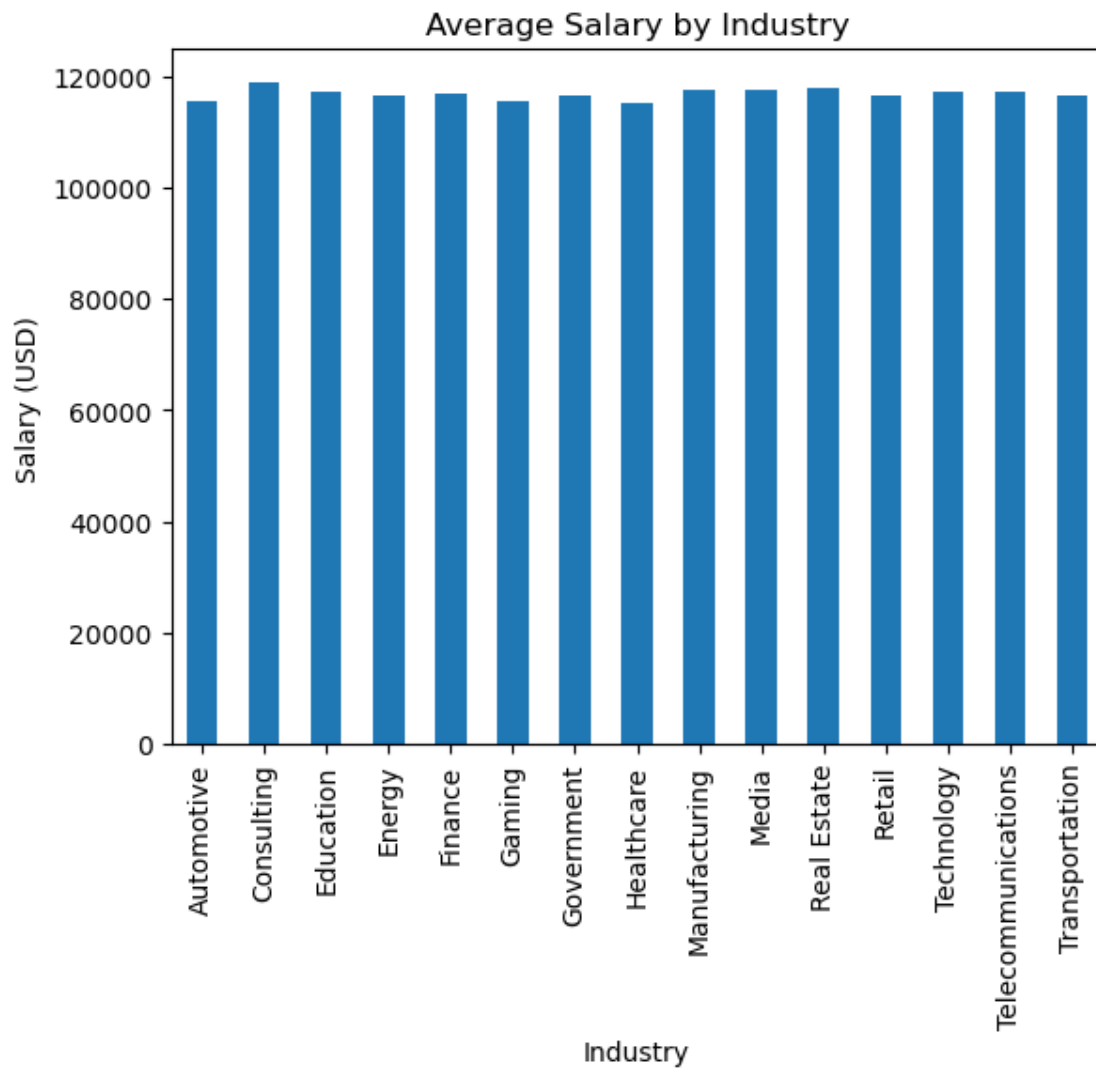
```
[345]: #jobs requiring Python as a skill
python_jobs = len(df[df["required_skills"].str.contains("Python", na=False)])
print(f"Jobs Requiring Python: {python_jobs}")
```

```
Jobs Requiring Python: 4151
```

```
[346]: #Average job description length by company size
avg_desc_by_size = df.groupby("company_size")["job_description_length"].mean()
print(avg_desc_by_size)
```

```
company_size
L      1506.393517
M      1506.251117
S      1497.365557
Name: job_description_length, dtype: float64
```

```
[347]: df.groupby("industry")["salary_usd"].mean().plot(kind="bar")
plt.title("Average Salary by Industry")
plt.xlabel("Industry")
plt.ylabel("Salary (USD)")
plt.show()
```



```
[348]: #Maximum benefits score for each job category
max_benefits_by_category = df.groupby("job_category")["benefits_score"].max()
print(max_benefits_by_category)
```

```
job_category
Analytics    10.0
Consulting   10.0
Engineering   10.0
```

```
Management      10.0
Other            10.0
Science          10.0
Specialist       10.0
Name: benefits_score, dtype: float64
```

```
[349]: #average salary for fully remote jobs
avg_remote_salary = df[df["remote_ratio"] == 100]["salary_usd"].mean()
print(f"Avg Salary for Fully Remote: {avg_remote_salary:.2f}")
```

```
Avg Salary for Fully Remote: 117465.24
```

```
[350]: #posting_date to datetime and count jobs posted in 2025
df["posting_date"] = pd.to_datetime(df["posting_date"])
jobs_2025 = len(df[df["posting_date"].dt.year == 2025])
print(f"Jobs Posted in 2025: {jobs_2025}")
```

```
Jobs Posted in 2025: 3412
```

```
[351]: #Average years of experience for jobs paying more than 100K
avg_exp_high_salary = df[df["salary_usd"] > 100000]["years_experience"].mean()
print(f"Avg Experience for Salary > 100K: {avg_exp_high_salary:.2f}")
```

```
Avg Experience for Salary > 100K: 9.82
```

```
[352]: #Top 10 countries (employee residence) by average salary
top_residence_salary = df.groupby("employee_residence")["salary_usd"].mean().
    ↪nlargest(10)
print(top_residence_salary)
```

```
employee_residence
Switzerland      153582.131258
Denmark          151289.712042
Norway           145006.631653
United States    138516.018519
Singapore        124986.124317
Netherlands      124122.331989
United Kingdom   122288.591093
Sweden           121396.645619
Germany          119481.945098
Australia        119109.619580
Name: salary_usd, dtype: float64
```

```
[353]: #the top 10 highest paying jobs
top_highest_paying = df.sort_values(by="salary_usd", ascending=False).head(10)
print(top_highest_paying[["job_title", "salary_usd", "company_name"]])
```

	job_title	salary_usd	company_name
9891	Machine Learning Researcher	399095	TechCorp Inc
2309	AI Architect	398084	Future Systems
12499	AI Research Scientist	394917	Machine Intelligence Group
12803	AI Specialist	390292	DataVision Ltd
317	Data Scientist	388754	DeepTech Ventures
5483	AI Architect	383142	Future Systems
10468	AI Product Manager	381575	Neural Networks Co
9637	AI Software Engineer	379418	Algorithmic Solutions
8377	AI Research Scientist	372206	Quantum Computing Inc
14431	AI Research Scientist	371087	Future Systems

```
[354]: #average benefits score by education level
avg_benefits_by_education = df.groupby("education_required")["benefits_score"].
    ↪mean()
print(avg_benefits_by_education)
```

```
education_required
Associate    7.518840
Bachelor    7.542760
Master      7.477699
PhD         7.471287
Name: benefits_score, dtype: float64
```

```
[355]: #how many jobs require both Python and SQL
python_sql_jobs = df[df["required_skills"].str.contains("Python", na=False) &
    df["required_skills"].str.contains("SQL", na=False)]
print(f"Jobs requiring both Python and SQL: {len(python_sql_jobs)}")
```

Jobs requiring both Python and SQL: 543

```
[356]: #average salary by education level
avg_salary_by_education = df.groupby("education_required")["salary_usd"].mean()
print(avg_salary_by_education)
```

```
education_required
Associate    116135.988172
Bachelor    117264.852667
Master      118479.750921
PhD         115482.582456
Name: salary_usd, dtype: float64
```

```
[357]: #job titles with the longest descriptions
longest_descriptions = df.sort_values(by="job_description_length",
↪ascending=False).head(10)
print(longest_descriptions[["job_title", "job_description_length"]])
```

	job_title	job_description_length
2782	Autonomous Systems Engineer	2499
12736	Principal Data Scientist	2499
3331	AI Product Manager	2499
2054	Machine Learning Researcher	2499
3386	AI Consultant	2499
1123	Head of AI	2499
7089	Head of AI	2499
1138	Head of AI	2498
9939	AI Research Scientist	2498
9550	Data Engineer	2498

```
[358]: #the number of jobs posted per month
df["posting_date"] = pd.to_datetime(df["posting_date"])
jobs_per_month = df["posting_date"].dt.to_period("M").value_counts().sort_index()
print(jobs_per_month)
```

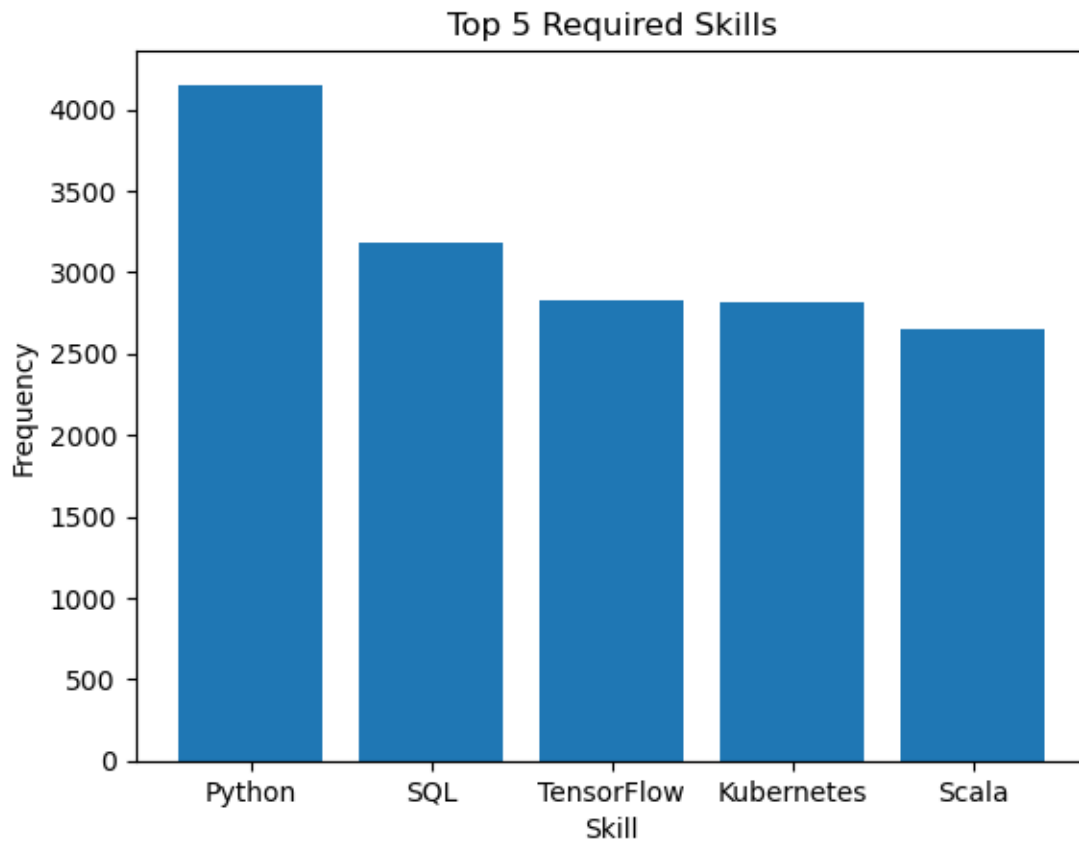
posting_date	
2024-01	889
2024-02	867
2024-03	873
2024-04	924
2024-05	895
2024-06	880
2024-07	906
2024-08	885
2024-09	827
2024-10	894
2024-11	881
2024-12	910
2025-01	897
2025-02	778
2025-03	865
2025-04	872

Freq: M, Name: count, dtype: int64

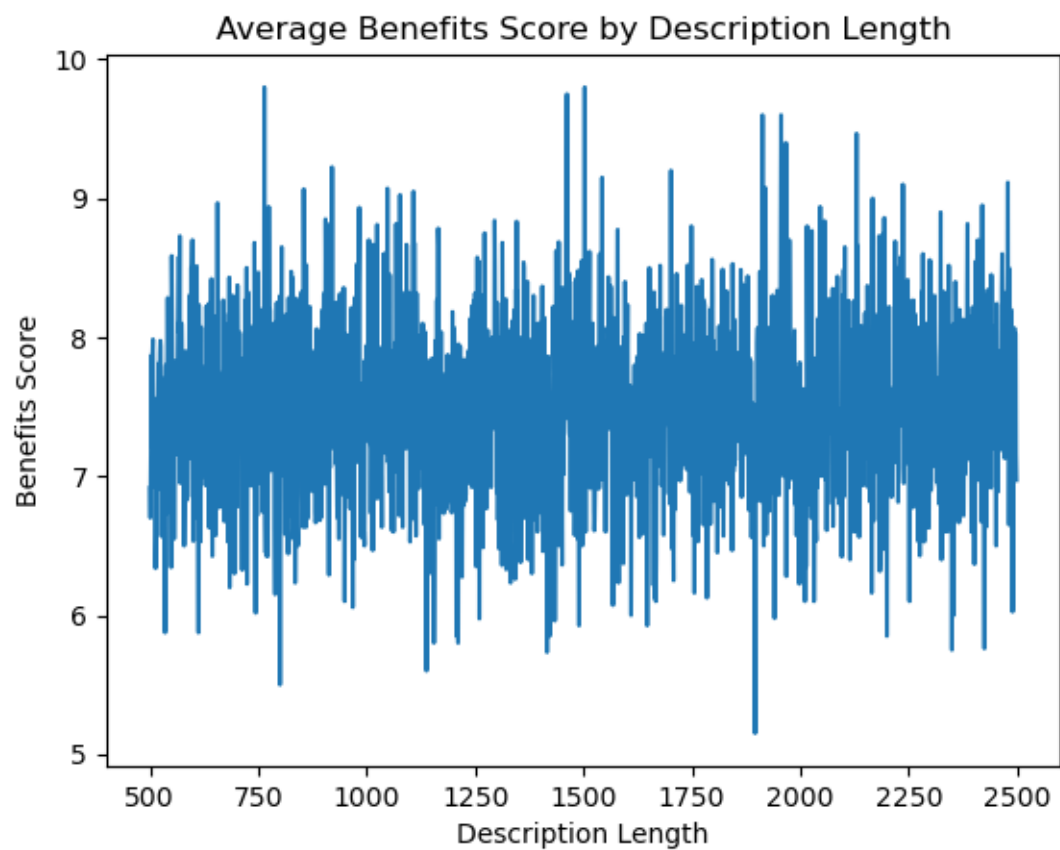


```
[359]: all_skills = df["required_skills"].str.split(", ").sum()
top_skills = Counter(all_skills).most_common(5)
skills, counts = zip(*top_skills)

plt.bar(skills, counts)
plt.title("Top 5 Required Skills")
plt.xlabel("Skill")
plt.ylabel("Frequency")
plt.show()
```



```
[360]: df.groupby("job_description_length")["benefits_score"].mean().plot(kind="line")
plt.title("Average Benefits Score by Description Length")
plt.xlabel("Description Length")
plt.ylabel("Benefits Score")
plt.show()
```



```
[361]: df["salary_usd"].plot(kind="hist", bins=50, edgecolor="black")
plt.title("Salary Distribution")
plt.xlabel("Salary (USD)")
plt.ylabel("Frequency")
plt.show()
```

