# BAIS:3200 Final Project

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#### Introduction

Every data scientist wants to earn a higher salary, but how can they attain a higher salary? In our project, we applied data-driven analysis to identify key factors such as experience, work setting, location, and more that contribute to salaries in the popular field of data science. Our database application and analysis may be useful for data scientists looking to increase their salary, or serve as a tool for data scientists and aspiring data scientists to examine their earnings potential in the data science field.

#### Data

Our project uses data collected from a 2024 Kaggle dataset about the data science industry across the world (<u>Job and Salaries in Data Science</u>). The original dataset contained 9,355 rows across 12 items. Our dataset was reduced to 10 items relevant to our analysis, and we reduced the size of the dataset to only include data from 2022-2023, leaving us with 9,087 rows across the 10 items. To uniquely identify records in our dataset, four primary keys were created, one for each relation. Table 1 displays a description of the data.

Table 1 Data Dictionary

Field	Туре	Description
employment_id company_id job_id salary_id work_year job_title job_category salary_in_usd employee_residence experience_level employment_type work_setting company_location	Text Text Text Text Numeric Text Text Numeric Text Numeric Text Text Text Text Text Text Text Text	Unique ID for each employment Unique ID for each company Unique ID for each job Unique ID for each salary Indicates the year the data was recorded Specifies the specific job role Classifies the job role into broader categories Presents the annual gross salary converted to United States Dollars Specifies the employee's country of residence Classifies the professional experience level Specifies the type of employment (e.g., full-time, part-time, contract) Describes the work setting (e.g., hybrid, remote, in-person) Indicates the country where the company is located
company_size	Text	Represents the size of the employer company

The dataset features three strong entities. First is EMPLOYMENT, identified by employment\_id, second is COMPANY, identified by company\_id, and third is JOB, identified by job\_id. All attributes in each entity are required, and there are no multivalued attributes. Additionally, we created an associative entity, SALARY, identified by salary\_id, to connect the three strong entities. Each strong entity has a one-to-many relationship with the SALARY entity, and the primary keys of the strong entities are stored as foreign keys in the SALARY entity. Figure 1 displays the ERD for our data.

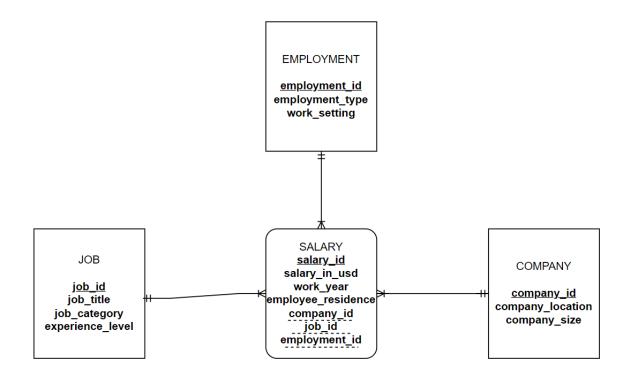


Fig. 1 Entity Relationship Diagram

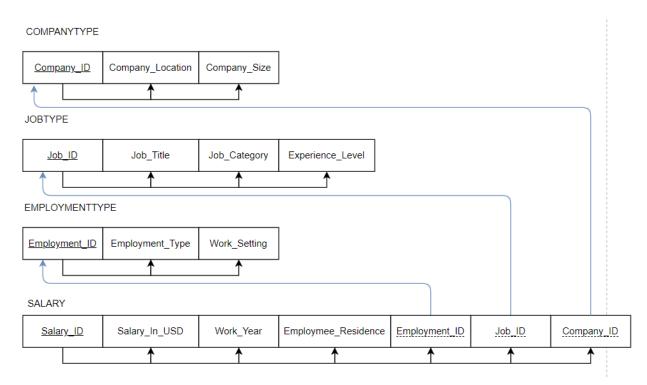


Fig. 2 Graphical Relational Schema

## **Database Implementation**

```
COMPANY Table:
CREATE TABLE COMPANY (
Company ID VARCHAR(12) NOT NULL,
Company_Location VARCHAR(25) NOT NULL,
Company Size CHAR(1) NOT NULL,
CONSTRAINT COMPANY_PK PRIMARY KEY (Company ID)
);
JOB Table:
CREATE TABLE JOB (
job_id varchar(25) not null,
job title varchar(50) not null,
job category varchar(50) not null,
experience level varchar(25) not null,
CONSTRAINT job_pk PRIMARY KEY (job_id)
);
EMPLOYMENT Table:
CREATE TABLE EMPLOYMENT (
EMPLOYMENT ID VARCHAR(20) NOT NULL,
EMPLOYMENT_TYPE VARCHAR(20) NOT NULL,
WORK SETTING VARCHAR(20) NOT NULL,
CONSTRAINT EMPLOYMENT_PK PRIMARY KEY (EMPLOYMENT_ID)
);
SALARY Table:
CREATE TABLE SALARY (
SALARY ID VARCHAR2(10) NOT NULL,
COMPANY ID VARCHAR2(25) NOT NULL,
JOB_ID VARCHAR2(10) NOT NULL,
EMPLOYMENT ID VARCHAR2(20) NOT NULL,
SALARY IN USD NUMBER(7,0) NOT NULL,
WORK YEAR NUMBER(4,0) NOT NULL,
EMPLOYEE RESIDENCE VARCHAR2(25) NOT NULL,
CONSTRAINT SALARY PK PRIMARY KEY (SALARY ID),
CONSTRAINT COMPANY FK FOREIGN KEY (COMPANY ID) REFERENCES COMPANY
(COMPANY ID),
CONSTRAINT EMPLOYMENT FK FOREIGN KEY (EMPLOYMENT ID) REFERENCES EMPLOYMENT
(EMPLOYMENT_ID),
CONSTRAINT JOB FK FOREIGN KEY (JOB ID) REFERENCES JOB (JOB ID)
);
```

# **Analysis**

Question 1: What are the differences in pay for different job categories?

```
SELECT Job_Category, TO_CHAR(AVG(Salary_In_USD), '$999,999.00') AS
AverageSalary
FROM JOB
    JOIN SALARY ON JOB.Job_ID = SALARY.Job_ID
GROUP BY Job_Category
ORDER BY AverageSalary DESC;
```

JOB_CATEGORY	AVERAGESALARY
Machine Learning and Al	\$178,925.85
Data Science and Research	\$163,758.58
Data Architecture and Modeling	\$156,002.36
Cloud and Database	\$155,000.00
Data Engineering	\$146,197.66
Leadership and Management	\$145,476.02
BI and Visualization	\$135,092.10
Data Analysis	\$108,505.72
Data Management and Strategy	\$103,139.93
Data Quality and Operations	\$100,879.47

Fig. 3 Average Salary by Category

Our data suggests that the job category has a large impact on average salary. Salaries range from an average of \$100,879 in Data Quality and Operations to \$178,925 in Machine Learning & AI. A difference of over \$70,000 from the minimum average salary to the maximum average salary demonstrates that a large difference in average salary based on job category exists.

Question 2: Are there major differences in the salaries for different job titles in the same category?

```
SELECT JOB.Job_Category, TO_CHAR(MAX(s.AverageSalary) - MIN(s.AverageSalary),
'$999,999.00') AS SalaryDifference
FROM JOB
    JOIN SALARY ON JOB.Job_ID = SALARY.Job_ID
```

JOB_CATEGORY	SALARYDIFFERENCE
Leadership and Management	\$321,040.60
Data Science and Research	\$312,496.00
Machine Learning and Al	\$204,000.00
Data Architecture and Modeling	\$162,166.11
Data Engineering	\$141,375.42
Data Analysis	\$99,327.00
BI and Visualization	\$86,623.92
Data Management and Strategy	\$82,750.00
Data Quality and Operations	\$80,645.00
Cloud and Database	\$.00

Fig. 4 Max Difference in Average Salary for Job Titles in the Same Category

Our data shows that different jobs within the same category can come with major salary differences. For example, in the Machine Learning and AI category, salaries can differ by as much as 204,000 dollars based on job title. The Cloud and Database category has a \$0 difference because the category only has one job title in our dataset, Cloud Database Engineer.

## Question 3: What work setting pays the highest salary?

WORK_SETTING	AVERAGESALARY	
In-person	\$155,524.12	
Remote	\$144,996.02	
Hybrid	\$88,912.29	

Fig. 5 Average Salary by Work Setting

The "In-person" work setting pays the highest average salary based on our data, with "Remote" coming in second, and "Hybrid" in last. "In-person" has the highest average, which makes logical sense, but it is interesting that "Hybrid" is the lowest, as a "Hybrid" working arrangement features both "In-person" and "Remote" components.

Question 4: Do Data Scientists get paid more when working outside of the United States? What proportion of Data Scientists work in the United States?

```
select CASE
   when employee_residence != 'United States' then 'NotUnitedStates'
   else 'UnitedStates'
   end as Region,
   to_char(avg(salary_in_usd), '$999,999.00') as AverageSalary
from salary
group by CASE
   when employee_residence != 'United States' then 'NotUnitedStates'
   else 'UnitedStates'
   end;
```

REGION	AVERAGESALARY	
NotUnitedStates	\$97,498.00	
UnitedStates	\$158,586.00	

Fig. 6 Average Salary by Work Location

According to our data, data scientists working in the United States make about \$61,088 more per year on average than their counterparts outside the United States.

```
SELECT CASE
    WHEN Employee_Residence != 'United States' THEN 'NotUnitedStates'
```

```
ELSE 'UnitedStates'
END AS Region, COUNT(Employment_ID) AS COUNT,
ROUND((COUNT(Employment_ID) / (SELECT COUNT(Employment_ID) FROM SALARY))
* 100, 2) || '%' AS Proportion
FROM SALARY
GROUP BY CASE
   WHEN Employee_Residence != 'United States' THEN 'NotUnitedStates'
ELSE 'UnitedStates'
END;
```

REGION	COUNT	PROPORTION
NotUnitedStates	1269	13.56%
UnitedStates	8086	86.44%

Fig. 7 Proportion of Workers by Work Location

The vast majority – over 85% – of the Data Scientists in our dataset work in the United States, which is higher than we expected as our dataset is composed of global data. A possible explanation for the high proportion of American data scientists is that the data may have been collected more heavily on U.S. based jobs, or the U.S. is simply leading the Data Science industry.

## Question 5: How much does experience level affect salary?

```
SELECT JOB.experience_level, TO_CHAR(AVG(SALARY.salary_in_usd),
'$999,999.00') as AverageSalary
FROM SALARY
FULL OUTER JOIN JOB on JOB.job_id = SALARY.job_id
GROUP BY JOB.experience_level
ORDER BY AverageSalary DESC;
```

EXPERIENCE_LEVEL	AVERAGESALARY
Executive	\$189,463.00
Senior	\$162,356.00
Mid-level	\$117,524.00
Entry-level	\$88,535.00

Fig. 8 Average Salary by Work Experience

The analysis of our dataset demonstrates that a data scientist's experience level leads to major differences in average salary. On average, a data scientist with more experience will be paid more than a data scientist with less experience. For example, if a data scientist progresses from entry-level to executive level, their annual salary can nearly double based on the trends in the dataset.

# **Web Application**

**Database Management Final Project Application** 

<u>Home Page</u>: This page gives an introduction to our application as well as a short description of the dataset we used for our analyses.

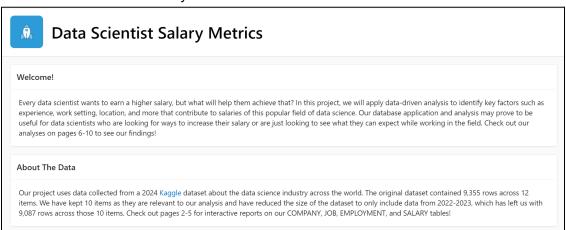


Fig. 9 Home Page (Page 1)

COMPANY Table: This page displays an interactive report for our COMPANY table.

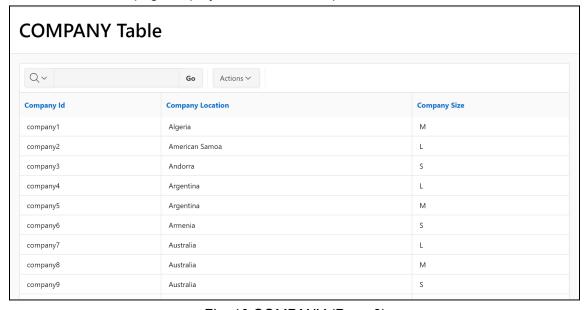


Fig. 10 COMPANY (Page 2)

JOB Table: This page displays an interactive report for our JOB table.

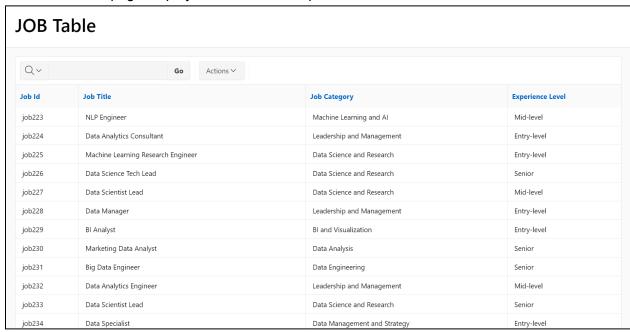


Fig. 11 JOB (Page 3)

EMPLOYMENT Table: This page displays an interactive report for our EMPLOYMENT table.

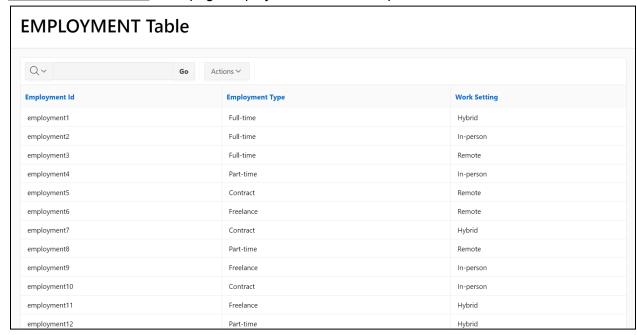


Fig. 12 EMPLOYMENT (Page 4)

SALARY Table: This page displays an interactive report for our SALARY table.

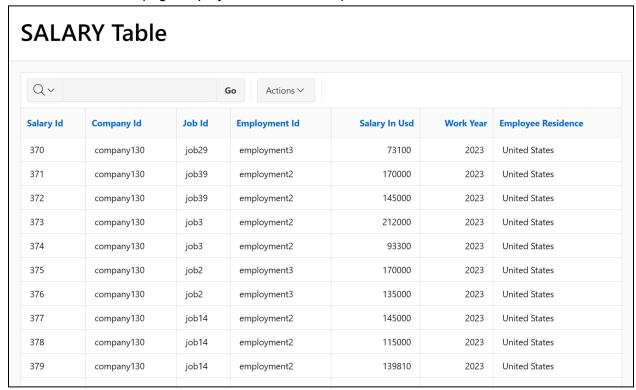


Fig. 13 SALARY (Page 5)

<u>Differences in Salary for Different Job Categories</u>: Page 6 displays a bar chart comparing average salaries across different data science job categories.

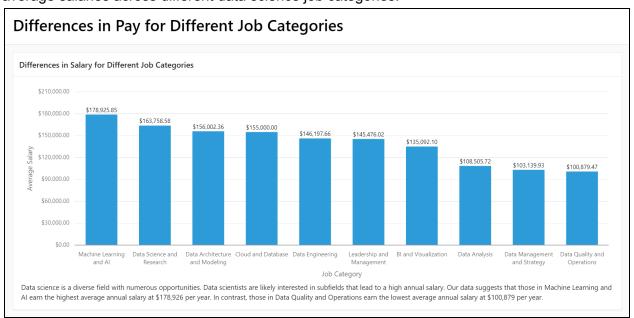


Fig. 14 Average Salary by Job Category (Page 6)

<u>Differences in Salary Within Same Job Category</u>: Page 7 displays a bar chart comparing the highest differences in salaries within the same job category.

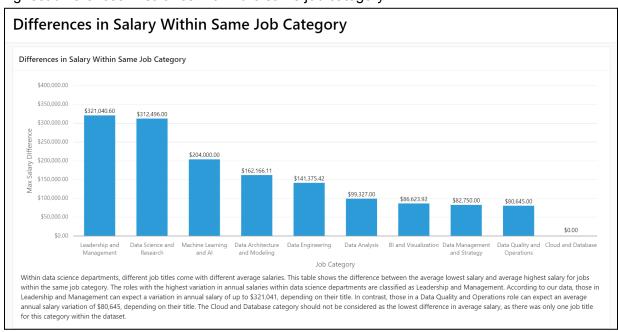


Fig. 15 Max Difference in Average Salary for Job Titles in the same Category (Page 7)

<u>Work Setting Salary</u>: Page 8 displays a bar chart comparing the average salary by work setting (i.e., in-person, remote, hybrid).



Fig. 16 Average Salary by Work Setting (Page 8)

<u>Experience Level Impact on Salary</u>: Page 9 displays a bar chart comparing the difference in average salary across different experience levels.

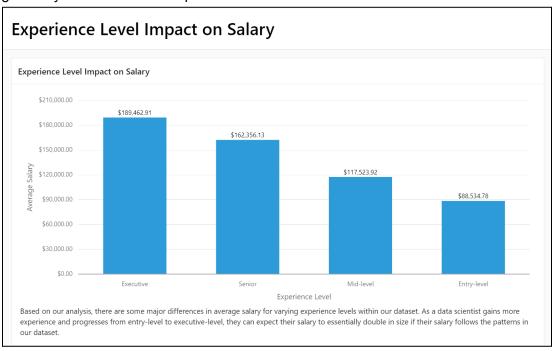


Fig. 17 Average Salary by Experience Level (Page 9)

<u>Data Scientist Salary Inside U.S. vs. Outside U.S.</u>: Page 10 displays a bar chart and a pie chart comparing average data science salaries within the U.S. to average data science salaries outside the U.S. and the proportion of data scientists inside vs. outside the U.S.

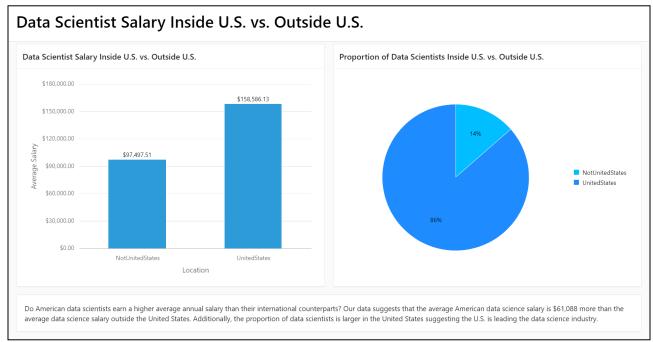


Fig. 18 Average Salary and Proportion Inside/Outside the U.S. (Page 10)