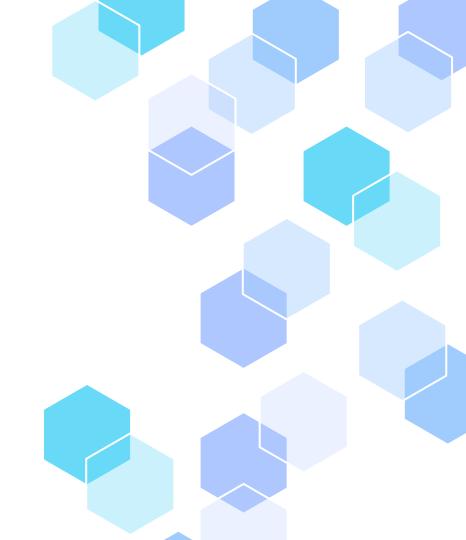
# Data Science Job Salary Predictions

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#### **Motivation**

- Emergence of Al boom
- Relevance and appeal of data science have grown
- Demand for skilled data science professionals is evident across various industries and regions
- Disparities in compensation based on company size, job title, etc.



#### **General Questions**

- How do different factors influence job salaries?
- Are there any notable trends or patterns in salary structures?



### **Specific Question**

If we were to become an entry-level data engineer in 2024, how much would our salary change working in Canada compared to US? Does remote work play a factor?



#### **Dataset Breakdown**

- id
- work\_year
- experience\_level
- employment\_type
- job\_title
- salary
- salary\_currency
- salary\_in\_usd
- employment\_residence
- remote\_ratio
- company\_location
- company\_size



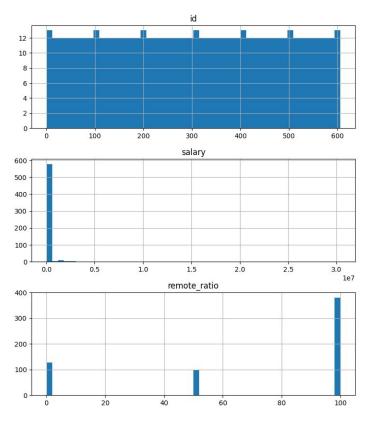
## ds\_salary.info()

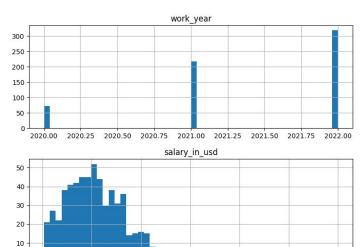
```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 607 entries, 0 to 606
Data columns (total 12 columns):
```

#	Column	Non-Null Count	Dtype	
0	id	607 non-null	int64	
1	work_year	607 non-null	int64	
2	experience_level	607 non-null	object	
3	employment_type	607 non-null	object	
4	job_title	607 non-null	object	
5	salary	607 non-null	int64	
6	salary_currency	607 non-null	object	
7	salary_in_usd	607 non-null	int64	
8	employee_residence	607 non-null	object	
9	remote_ratio	607 non-null	int64	
10	company_location	607 non-null	object	
11	company_size	607 non-null	object	
dtypes: int64(5), object(7)				

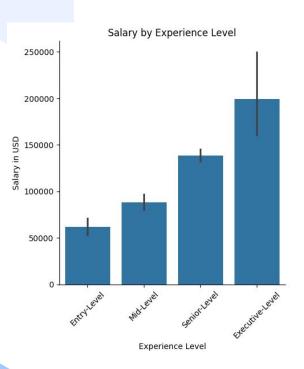
dtypes: int64(5), object(7) memory usage: 57.0+ KB

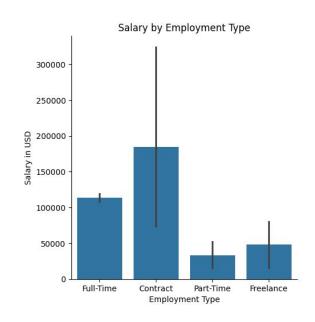
## ds\_salary.hist()

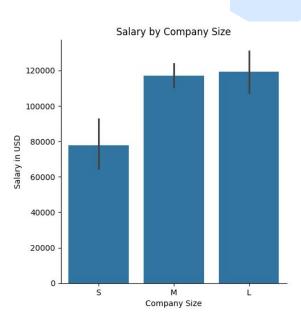




## **Plotting Against Salary**







#### **Problems**

- Seven categorical attributes
- Small dataset with more than half of our data based in the US
- Outliers in most of our categorical data

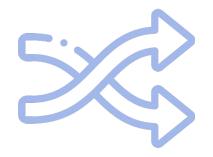
#### Example

job_title			
Data Scientist	143		
Data Engineer	132		
Data Analyst	97		
Machine Learning Engineer	41		
Research Scientist	16		
Data Science Manager	12		
Data Architect	11		
Big Data Engineer	8		
Machine Learning Scientist	8		
Principal Data Scientist	7		
AI Scientist	7		
Data Science Consultant	7		
Director of Data Science	7		
Data Analytics Manager	7		
ML Engineer	6		
Computer Vision Engineer			
BI Data Analyst	6		
Lead Data Engineer			
Data Engineering Manager			
Business Data Analyst	5		
Head of Data	5		
Applied Data Scientist	5		
Applied Machine Learning Scientist	4		
Head of Data Science	4		
***			
Finance Data Analyst	1		
Marketing Data Analyst	1		
Machine Learning Manager			
Data Analytics Lead			
Name: count, dtype: int64			

#### **Feature Engineering**

- Removed unnecessary columns (id and salary)
- Converted categorical data -> Numeric data
  - Entry-level -> 1, Mid-level -> 2 and so on..
- Bucketed the outliers
  - For the job title, if the count is less than 10 grouped into 0ther
- Created a new column same\_country
  - If employee\_residence = company\_location then 1 otherwise 0
- Created three new columns salary\_mean, salary\_min and salary\_max for each of the salaries grouped by clusters
  - Clusters: work\_year, experience\_level, employment\_type, job\_title, remote\_ratio, company\_size and same\_location

#### **Data Preprocessing**



- Numerical: MinMaxScaler()
- Categorical: OneHotEncoder()



## **Modeling for Regression**

- Ridge()
- SGDRegressor()
- RandomForestRegressor()
- DecisionTreeRegressor()

#### **Results**

#### Mean Squared Error (MSE)

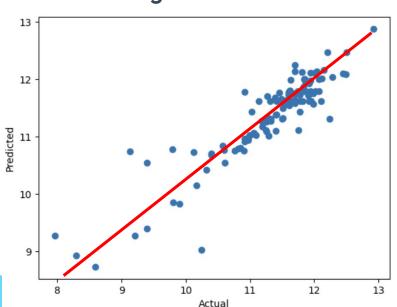
Ridge	0.19725377418055814
SGD	0.4231541951213471
Random Forest	0.12491690719113324
Decision Tree	0.14350962674677212

#### **R2 Score**

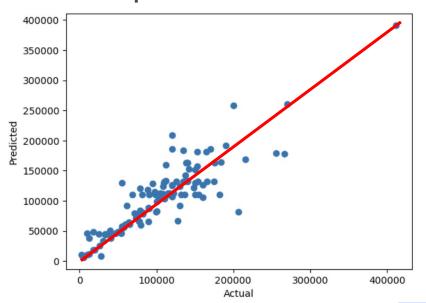
Ridge	0.7195537582301159
SGD	0.39837904646471056
Random Forest	0.8223989512960932
Decision Tree	0.7959646873874725

#### **Actual vs Predicted Salaries**

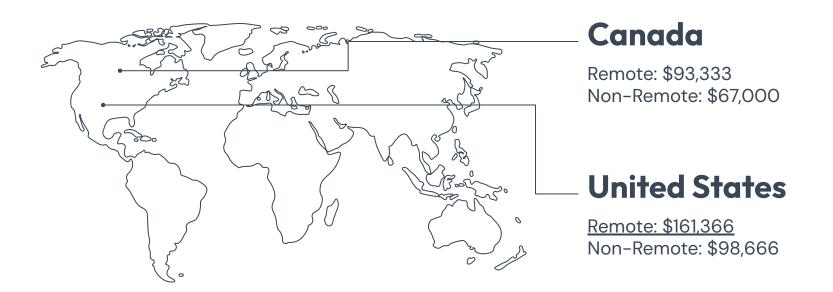
#### **Log Transformed**



#### **Exponential Transformed**



#### Salary Prediction Results (USD)





## Thank you