# **Assignment 03**

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September 21, 2025

```
from pyspark.sql import SparkSession
from pyspark.sql import functions as F
from pyspark.sql.functions import col, split, explode, regexp_replace, transform, when
from pyspark.sql.functions import col, monotonically_increasing_id
from pyspark.sql.types import StructType # to/from JSON
import json
import re
import numpy as np
import pandas as pd
import plotly.express as px
import plotly.io as pio
import plotly.graph_objects as go
np.random.seed(30) # set a fixed seed for reproducibility
pio.renderers.default = "vscode+notebook"
# Initialize Spark Session
spark = SparkSession.builder.appName("JobPostingsAnalysis").getOrCreate()
# Load schema from JSON file
with open("data/schema_lightcast.json") as f:
    schema = StructType.fromJson(json.load(f))
# Load Data
df = (spark.read
      .option("header", "true")
      .option("inferSchema", "false")
      .schema(schema)
                                   # saved schema
      .option("multiLine", "true")
      .option("escape", "\"")
```

```
.csv("data/lightcast_job_postings.csv")
.limit(5000))

df.createOrReplaceTempView("job_postings")
# Show Schema and Sample Data
#df.printSchema()
df.show(5)
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# 1 Data Preparation

```
.withColumn("MIN_YEARS_EXPERIENCE", col("MIN_YEARS_EXPERIENCE").cast("float"))\
       .withColumn("MAX_YEARS_EXPERIENCE", col("MAX_YEARS_EXPERIENCE").cast("float"))
# Step 2: Computing medians for salary columns
def compute median(sdf, col name):
    q = sdf.approxQuantile(col_name, [0.5], 0.01)
    return q[0] if q else None
median_from = compute_median(df, "SALARY_FROM")
median_to = compute_median(df, "SALARY_TO")
median_salary = compute_median(df, "SALARY")
print("Medians:", median_from, median_to)
# Step 3: Imputing missing salaries, but not experience
df = df.fillna({
    "SALARY_FROM": median_from,
    "SALARY_TO": median_to,
    "SALARY": median_salary
})
# Step 5: Computing average salary
df = df.withColumn("Average_Salary",
                   (col("SALARY_FROM") + col("SALARY_TO")) / 2)
# Step 6: Selecting required columns
export_cols = [
    "EDUCATION_LEVELS_NAME",
    "REMOTE_TYPE_NAME",
    "MAX_YEARS_EXPERIENCE",
    "Average_Salary",
    "SALARY",
    "LOT V6 SPECIALIZED OCCUPATION NAME"
1
df_selected = df.select(*export_cols)
# Step 7: Saving to CSV
pdf = df selected.toPandas()
pdf.to_csv("data/lightcast_cleaned.csv", index=False)
print("Data cleaning complete. Rows retained:", len(pdf))
```

Medians: 89565.0 131400.0

Data cleaning complete. Rows retained: 5000

#### 1.1 Salary Distribution by Industry and Employment Type

• Compare salary variations across industries.

Filter the dataset - Remove records where salary is missing or zero.

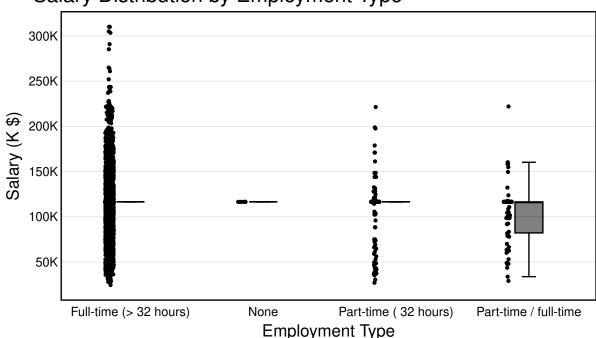
**Aggregate Data** - Group by **NAICS industry codes** (e.g., NAICS2\_NAME). - Group by **employment type** (EMPLOYMENT\_TYPE\_NAME) and compute salary distribution. - Calculate **salary percentiles** (25th, 50th, 75th) for each group.

Visualize results - Create a box plot where: - X-axis = NAICS2\_NAME - Y-axis = SALARY\_FROM, or SALARY\_TO, or SALARY - Group/color = EMPLOYMENT\_TYPE\_NAME - Customize colors, fonts, and styles.

**Explanation:** Write two sentences about what the graph reveals (e.g., median differences across industries and dispersion by employment type).

### 2 Set up plotly template

### Salary Distribution by Employment Type



### 3 Salary Distribution by Industry and Employment Type

- Compare salary variations across industries.
- Filter the dataset
  - Remove records where **Salary is missing or zero**.
- Aggregate Data
  - Group by **NAICS** industry codes.
  - Group by **employment type** and compute salary-distribution.
- Visualize results
  - Create a **box plot** where:
    - \*  $\mathbf{X}$ - $\mathbf{axis} = \mathtt{NAICS2}$ \_NAME.
    - \* Y-axis = SALARY\_FROM.
    - \* Group by EMPLOYMENT\_TYPE\_NAME.
  - Customize colors, fonts, and styles.
- Explanation: Write two sentences about what the graph reveals.

### 4 Salary Analysis by ONET Occupation Type (Bubble Chart)

- Analyze how salaries differ across ONET occupation types.
- Aggregate Data
  - Compute median salary for each occupation in the **ONET taxonomy**.
- Visualize results
  - Create a **bubble chart** where:
    - $* \mathbf{X}\text{-}\mathbf{axis} = \mathtt{ONET}\mathtt{\_NAME}$
    - $* \mathbf{Y}\text{-}\mathbf{axis} = \mathtt{Median} \ \mathtt{Salary}$
    - \* **Size** = Number of job postings
  - Apply custom colors and font styles.
- Explanation: Write two sentences about what the graph reveals.

## 5 Salary by Education Level

- Create two groups:
  - Bachelor's or lower (Bachelor's, GED, Associate, No Education Listed)

- Master's or PhD (Master's degree, Ph.D. or professional degree)
- Plot scatter plots for each group using MAX\_YEARS\_EXPERIENCE (with jitter), Average\_Salary, LOT\_V6\_SPECIALIZED\_OCCUPATION\_NAME.
- Then, plot histograms overlaid with KDE curves for each group.
  - This would generate two scatter plots and two histograms.
- After each graph, add a short explanation of key insights.

### 6 Salary by Remote Work Type

- Split into three groups based on REMOTE\_TYPE\_NAME:
  - Remote
  - Hybrid
  - Onsite (includes [None] and blank)
- Plot scatter plots for each group using MAX\_YEARS\_EXPERIENCE (with jitter), Average\_Salary, LOT\_V6\_SPECIALIZED\_OCCUPATION\_NAME.
- Also, create salary histograms for all three groups.
- · After each graph, briefly describe any patterns or comparisons.

#### **Submission Instructions**

- Submit the Word Document (part of git repo) containing:
  - The **HTTPS URL** of your GitHub repository.
  - Answer to the questions.
  - Visualizations created using matplotlib, Seaborn or plotly (preferred).
  - Answers to the questions below.

```
# Step 2: Convert to Pandas DataFrame
salary_pd = salary_analysis.toPandas()
salary_pd.head()
# Step 3: Bubble chart using Plotly
import plotly.express as px
fig = px.scatter(
   salary_pd,
    x="Occupation_name",
    y="Median_Salary",
    size="Job_Postings",
    title="Salary Analysis by LOT Occupation Type (Bubble Chart)",
    labels={
        "Occupation_name": "LOT Occupation",
        "Median_Salary": "Median Salary",
        "Job_Postings": "Number of Job Postings"
    },
    hover_name="Occupation_name",
    size_max=60,
    width=1000,
    height=600,
    color="Job_Postings",
    color_continuous_scale="Plasma"
# Step 4: Layout customization
fig.update_layout(
    font_family="Arial",
    font size=14,
    title_font_size=25,
    xaxis_title="LOT Occupation",
    yaxis_title="Median Salary",
    plot_bgcolor="white",
    xaxis=dict(
        tickangle=-45,
        showline=True,
```

```
linecolor="black"
),
yaxis=dict(
    showline=True,
    linecolor="black"
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)

# Step 5: Show and export
fig.show()
fig.write_image("output/Q7.svg", width=1000, height=600, scale=1)
```

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