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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only showing top 5 rows
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```
# Histogram of SALARY distribution (cast + filter)
salary_df = (
    df.select(col("SALARY").cast("float"))
      .filter(col("SALARY").isNotNull() & (col("SALARY") > 0))
)

fig = px.histogram(
    salary_df.toPandas(),
    x="SALARY",
    nbins=50,
    title="Salary Distribution"
)
fig.update_layout(bargap=0.1)
fig
```

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Unable to display output for mime type(s): text/html
```

```
Unable to display output for mime type(s): application/vnd.plotly.v1+json, text/html
```

1 Data Preparation

```
# Step 1: Casting salary and experience columns
from pyspark.sql.functions import col

df = df.withColumn("SALARY", col("SALARY").cast("float")) \
       .withColumn("SALARY_FROM", col("SALARY_FROM").cast("float")) \
       .withColumn("SALARY_TO", col("SALARY_TO").cast("float")) \
```

```
        .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"
]
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.

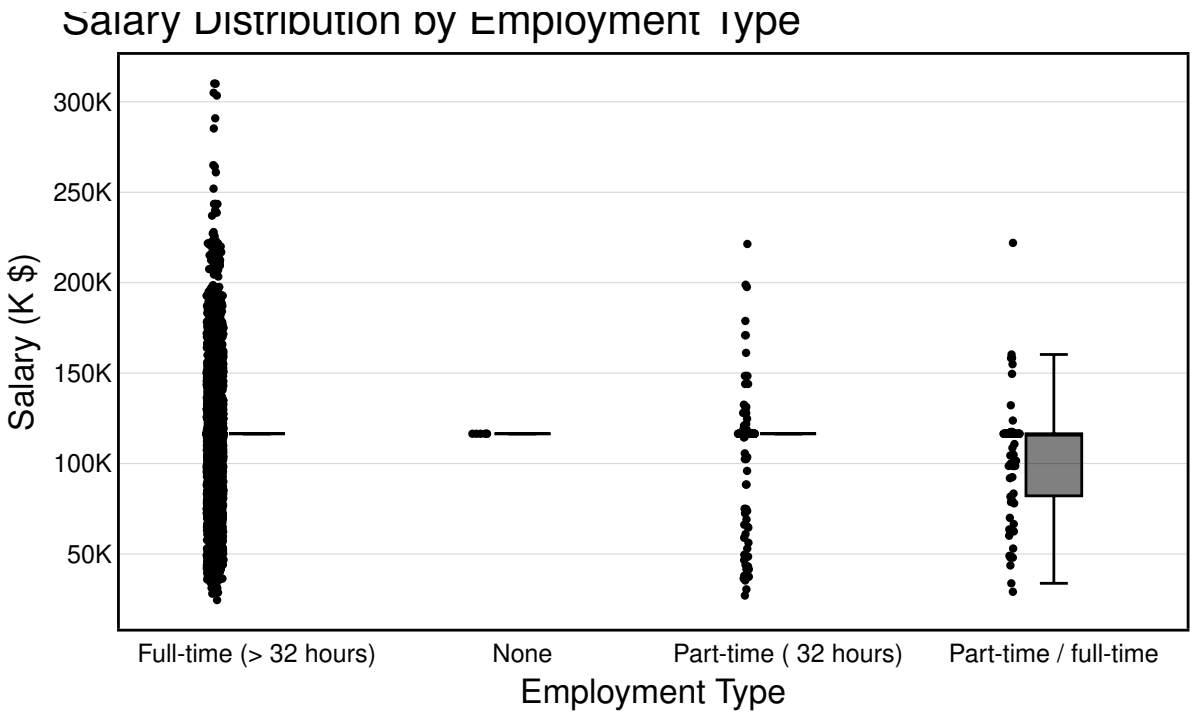
Filter the dataset - Remove records where **salary is missing or zero**.

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



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:
    - **X-axis** = `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:
    - **X-axis** = `ONET_NAME`
    - **Y-axis** = `Median 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 1: Spark SQL - Median salary and job count per TITLE_NAME
salary_analysis = spark.sql("""
    SELECT
        LOT_OCCUPATION_NAME AS Occupation_name,
        PERCENTILE(SALARY, 0.5) AS Median_Salary,
        COUNT(*) AS Job_Postings
    FROM job_postings
    GROUP BY LOT_OCCUPATION_NAME
    ORDER BY Job_Postings DESC
    LIMIT 10
""")
```

```
# 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"
    )
)

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

```
Unable to display output for mime type(s): application/vnd.plotly.v1+json, text/html
```

```
# Defining education level groupings
lower_deg = ["Bachelor's", "Associate", "GED", "No Education Listed", "High school"]
higher_deg = ["Master's degree", "PhD or professional degree"]

# Adding EDU_GROUP column
df = df.withColumn(
    "EDU_GROUP",
    when(col("EDUCATION_LEVELS_NAME").rlike("|".join([f"(?i){deg}" for deg in lower_deg])), '
    .when(col("EDUCATION_LEVELS_NAME").rlike("|".join([f"(?i){deg}" for deg in higher_deg]))
    .otherwise("Other")
)

# Casting necessary columns to float
df = df.withColumn("MAX_YEARS_EXPERIENCE", col("MAX_YEARS_EXPERIENCE").cast("float"))
df = df.withColumn("Average_Salary", col("Average_Salary").cast("float"))

# Filtering for non-null and positive values
df = df.filter(
    col("MAX_YEARS_EXPERIENCE").isNotNull() &
    col("Average_Salary").isNotNull() &
    (col("MAX_YEARS_EXPERIENCE") > 0) &
    (col("Average_Salary") > 0)
)

# Filtering for just the two education groups
df_filtered = df.filter(col("EDU_GROUP").isin("Bachelor's or lower", "Master's or PhD"))
```

```
# Converting to Pandas for plotting
df_pd = df_filtered.toPandas()
df_pd.head()
```

```
[Stage 137:>                                                        (0 + 1) / 1]
```

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1	0cb072af26757b6c4ea9464472a50a443af681ac	8/2/2024	2024-08-02 17:08:58.838
2	5a843df632e1ff756fa19d80a0871262d51becc0	6/21/2024	2024-06-21 07:00:00.000
3	229620073766234e814e8add21db7dfaef69b3bd	10/9/2024	2024-10-09 18:07:44.758
4	138ce2c9453b47a9b33403c364d4fd80996caa4f	8/10/2024	2024-08-10 19:36:49.244