

# SINGAPORE LABOUR MARKET ANALYSIS

[IT8701] Introduction to Programming for Data Science

## OVERVIEW

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02 Objectives

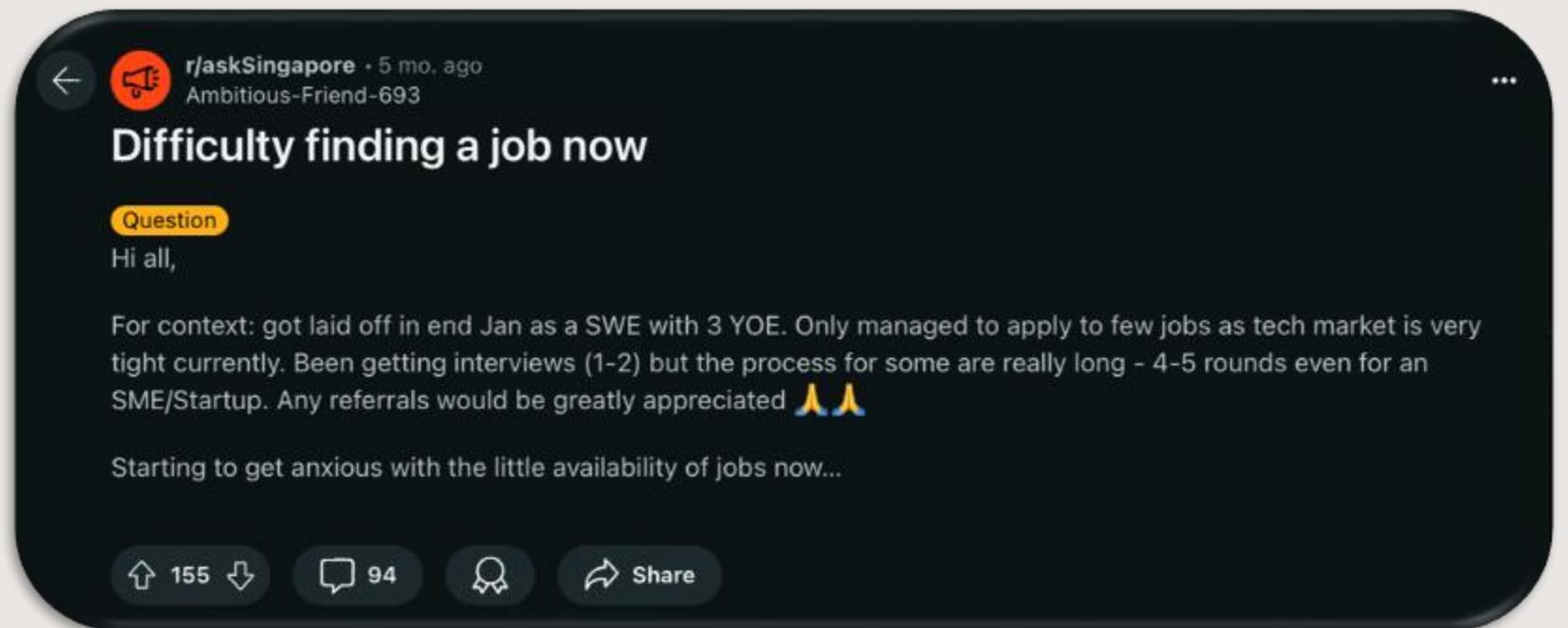
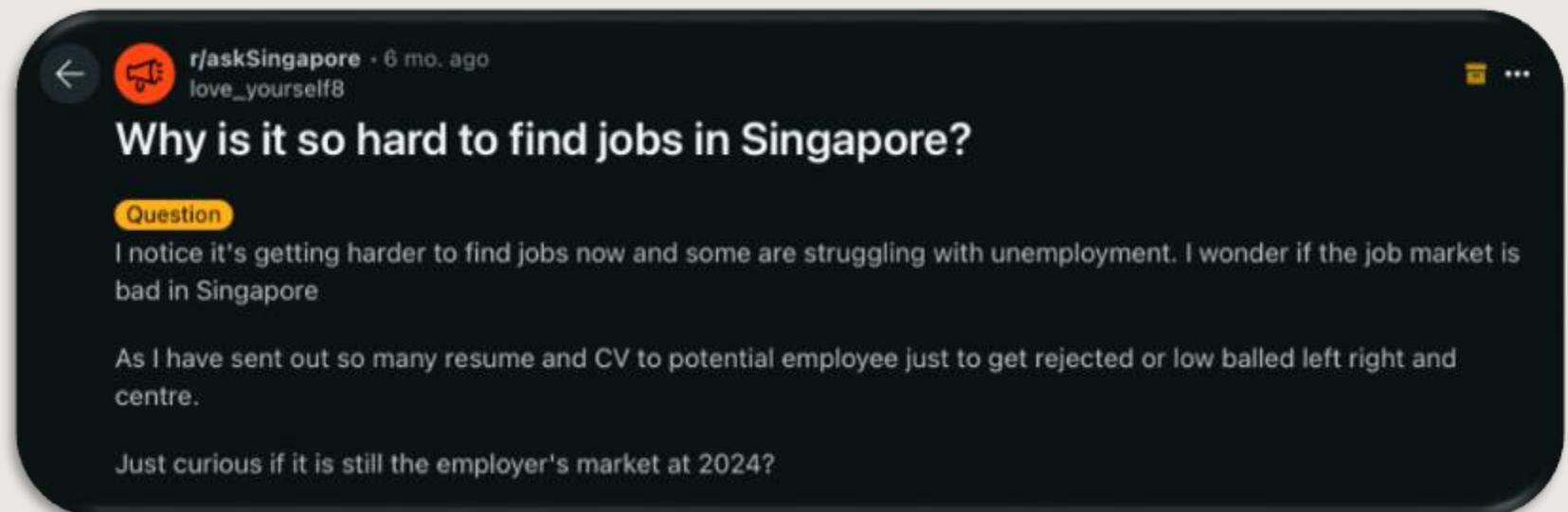
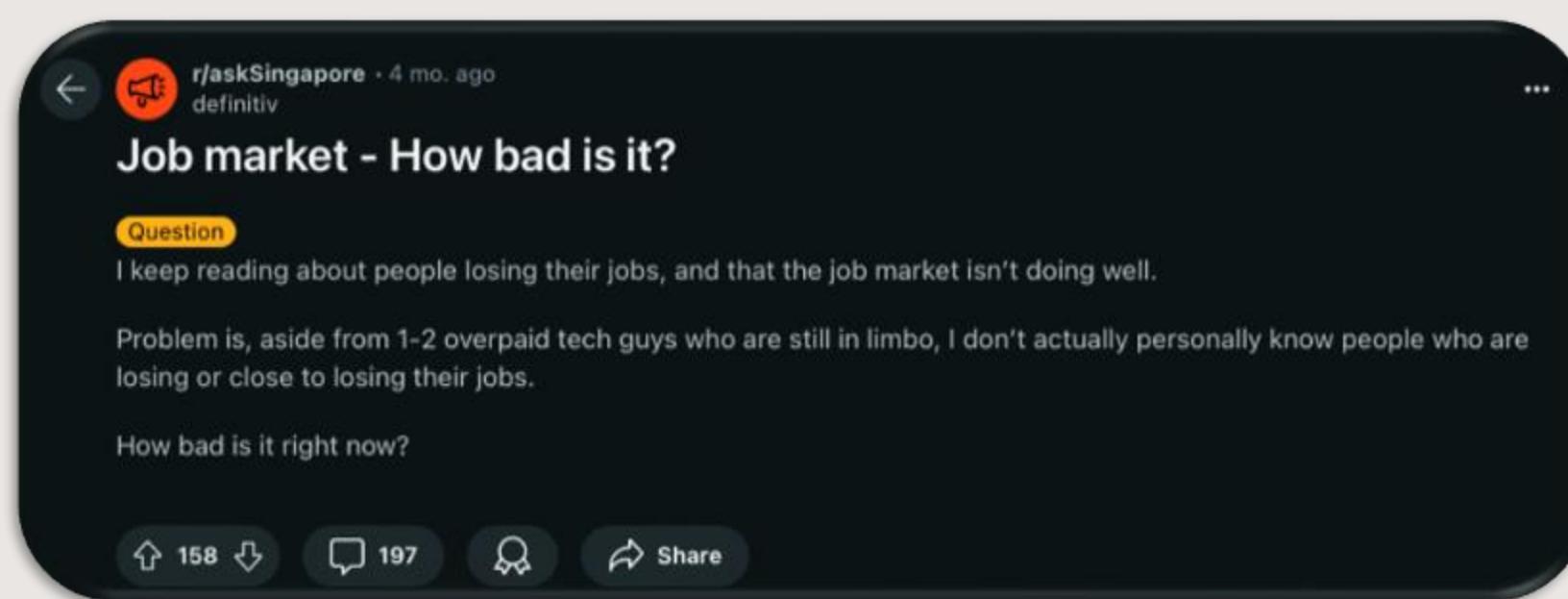
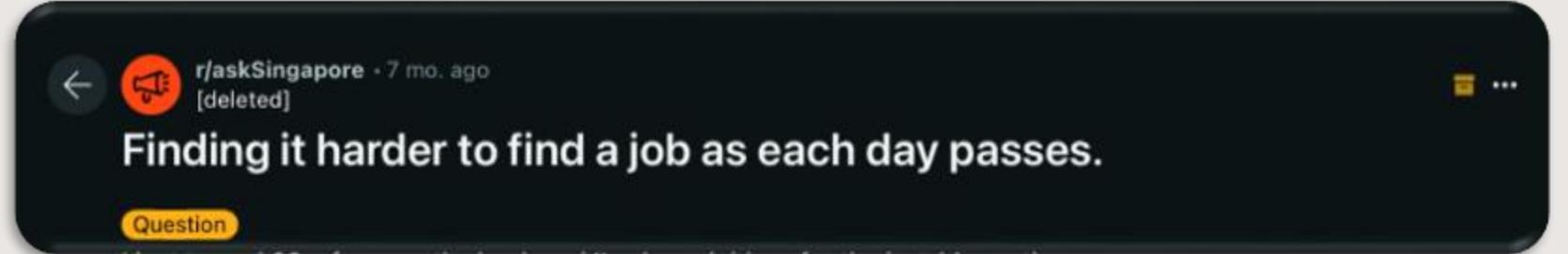
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## INTRODUCTION



## INTRODUCTION

“With an improved economic outlook for 2024, sustained increase in the number of job vacancies, increased hiring optimism among firms over the next quarter, as well as declining retrenchment numbers, we expect the labour market to continue expanding and unemployment rates to remain low”

[Ministry of Manpower \(MOM\)](#)  
[Q1 2024 Labour Market Report](#)

## INTRODUCTION

# How Is The Labour Market?

In recent months, there has been growing sentiment that Singapore's economy is facing significant challenges. Concerns about economic downturns, fluctuating markets, and domestic economic pressures have sparked discussions about the state of the job market. Despite these worries, the Ministry of Manpower's (MOM) outlook suggests that growth opportunities may still be on the horizon.

As such, this analysis seeks to assess the current state of Singapore's labour market amidst prevailing concerns. Specifically, we will focus on comparing Singapore's labour market globally and explore job opportunities across various industries and sectors, with the aim of helping job-seekers achieve both job security and satisfaction.



## OBJECTIVES

### Research Questions

1. **What** is the current state of Singapore's labour market in comparison to other countries?
2. **Which** sectors or industries in Singapore offer the greatest job security in the present economic climate?
3. **Where** can job-seekers look for opportunities within the various sectors or industries?
4. **How** can job-seekers find employment that aligns with their expectations and offers satisfaction?

### Assumption

Job-seekers assumed to possess versatile skillsets with relevant qualifications, allowing them to apply for similar job roles across various industries (e.g., working as a data analyst / sales manager in the IT sector versus the petroleum industry)



## METHODOLOGY (DATASETS & TOOLS)

### Datasets

1. [Overall Unemployment Rate, Annual](#)
2. [OECD Annual Unemployment Rates](#)
3. [World Bank Country and Lending Groups](#)
4. [Retrenchment By Industry](#)
5. [Average Monthly Recruitment/Resignation Rate by Industry and Occupational Group](#)
6. [Occupational Wages within each Major Occupational Group by Industry](#)
7. [Average Weekly Paid Hours Worked Per Employee by Industry and Type of Employment](#)
8. [Distribution of Full-Time Employees By Annual Leave Entitlement](#)
9. [Proportion of Establishments That Provided Non-Statutory Leave by Type](#)
10. [Proportion of Establishments Offering Scheduled FWAs-Industry](#)

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### Technologies/Libraries Used

#### Python

Pandas      Seaborn

Numpy      Plotly

Requests    Matplotlib

Sklearn     MySQLconnector

#### MySQL

MySQL 8.0.35 for database creation

inserting data into tables, and querying



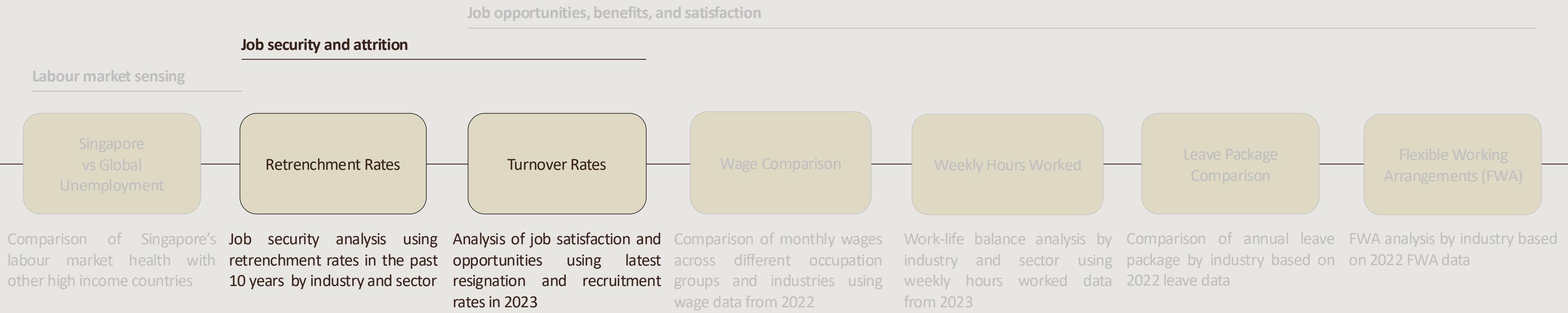
## METHODOLOGY (APPROACH)

# Analysis Framework



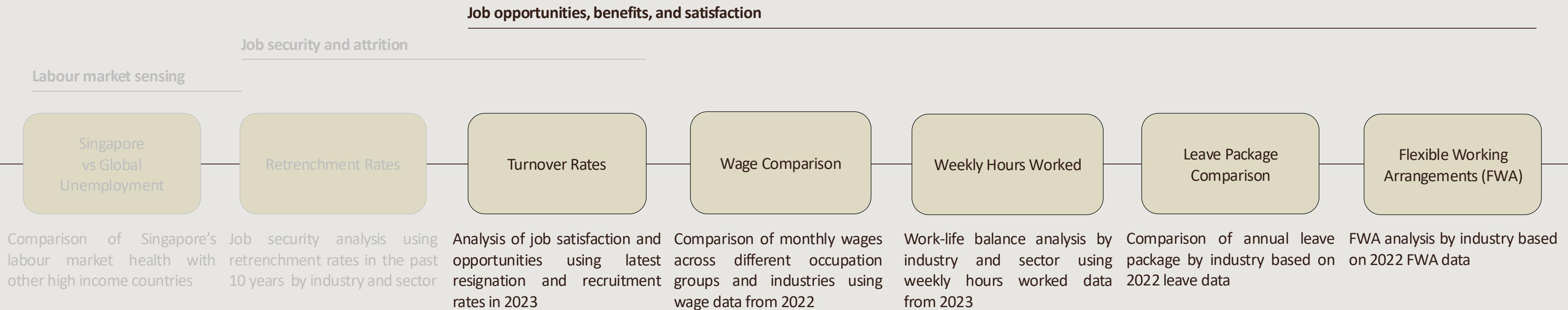
## METHODOLOGY (APPROACH)

# Analysis Framework



## METHODOLOGY (APPROACH)

# Analysis Framework



## METHODOLOGY (DATA PRE-PROCESSING)

### Extract-Transform-Load (ETL) Workflow

- Downloaded datasets using self-defined function to call APIs (wherever possible)
- Cleaned raw data (minimally) by replacing blanks and nulls with `None` data type to mitigate data integrity errors enforced by MySQL when executing `INSERT INTO` statement

Frequently encountered error during attempts to import data into MySQL!

```
DatabaseError: 1366 (HY000): Incorrect integer value: '-' for column 'RETRENCH_CONTRACT' at row 1
```

```
1 # Retrieve data
2 retrenchment = get_data('d_c794e2ad25c1d3bb7965d13dff0b411')
3
4 # Data cleaning for importing into MySQL Replace NULL values as None type
5 retrenchment.replace('-', None, inplace=True)
```

```
1 # Retrieve data
2 wfh = get_data('d_9190c1b4524b6361ce6afdf39fa94e')
3
4 # Data cleaning for importing into MySQL Different characters to replace to ensure
5 # Data cleaning for importing into MySQL data cleanliness and integrity
6 wfh.replace({'-': None, 'na': None, 's': None}, inplace=True)
```

```
1 def get_data(resource_id): Custom function to call API and save data into pandas data frame
2 """
3     Returns data.gov.sg dataset through use of OpenAPI query.
4
5     Args:
6         resource_id (string): Resource ID string pointing to the API endpoint
7
8     Returns:
9         pandas dataframe: Dataframe of data.gov.sg dataset
10    """
11
12    # Define key arguments for data retrieval and formatting
13    base_url = 'https://data.gov.sg/api/action/dataset_search' Define API URL
14    offset = 0
15    df = []
16
17    while True:
18        # Set url and parameters for API endpoint
19        parameters = {
20            'resource_id': resource_id,
21            'offset': offset # define offset
22        }
23
24        # Retrieve data
25        response = requests.get(base_url, params=parameters) Specify end-point for individual datasets
26
27        # Check if request is successful
28        if response.status_code == 200:
29            # Retrieve json text
30            response_dict = response.json() Parse JSON data into pandas data frame
31
32            # Check if at least 1 record
33            if len(response_dict['result']['records']) > 0:
34                # Normalize and append records to list, df
35                df_temp = pd.json_normalize(response_dict['result']['records'])
36                df.append(df_temp)
37
38            # Continue the loop and increase offset to retrieve the next 100 records
39            offset += 100
40
41            # Repeat retrieval of every 100 records until the request returns 0 records - break loop
42        else:
43            break Iterate until all available rows have been read and saved
44
45        # If request not successful, print error message with status code
46        else:
47            print(f'The request failed and returned status code: {response.status_code}')
48
49    # Concatenate list of dataframes
50    df = pd.concat(df, ignore_index=True) Concatenate and save data into data frame
51
52    # Return df
53
54    return df
```

## METHODOLOGY (DATA PRE-PROCESSING)

### Database Schema Creation

- Created MySQL database schema to facilitate loading of downloaded tables

```
1 # Create database schema if schema does not yet exist
2 query = ''
3 CREATE DATABASE IF NOT EXISTS SGMARKET
4 ...
5         Ensures that I can re-run the code without error
6 try:
7     cursor.execute(query)
8     con.commit()
9 except:
10    print('Unexpected error:', sys.exc_info()[0])
11   print(sys.exc_info()[1])
12 finally:
13     cursor.close()
14     con.close()
```

```
1 # Retrieve data
2 sg_unemployment = get_data('d_e3598914c86699a9a36e68190f78c59a')
3
4 # Create table      Define and create table in MySQL for loading dataset
5 query = ''
6 CREATE TABLE IF NOT EXISTS SG_UNEMPLOYMENT (
7     ID INT NOT NULL,
8     YEAR INT NOT NULL,
9     RESIDENTIAL_STATUS VARCHAR(20) NOT NULL,
10    UNEMPLOYMENT_RATE FLOAT(10) NOT NULL,
11    PRIMARY KEY (ID)
12 )
13 ...
14
15 try:
16     cursor.execute(query)      Execute SQL query through Python to create table
17     con.commit()
18 except:
19     print('Unexpected error:', sys.exc_info()[0])
20     print(sys.exc_info()[1])
21
22 # Insert data into table
23 for index, row in sg_unemployment.iterrows():
24     data = {
25         'id': row['_id'],           Iterate through each row to retrieve data for inserting
26         'year': row['year'],        into table
27         'residential_status': row['residential_status'],
28         'unemployment_rate': row['unemployment_rate']
29     }
30
31 # %(key)s: This syntax is used for string formatting where key is a placeholder
32 # that will be replaced by the value associated with key in a dictionary.
33 query = ''
34     INSERT INTO SG_UNEMPLOYMENT (      Insert query to run in MySQL
35         ID,
36         YEAR,
37         RESIDENTIAL_STATUS,
38         UNEMPLOYMENT_RATE
39     )
40     VALUES (
41         %(id)s,
42         %(year)s,
43         %(residential_status)s,
44         %(unemployment_rate)s
45     )
46     ...
47     cursor.execute(query, data)      Execute SQL query through Python to insert values
48     con.commit()
49
50 # View data
51 query = ''
52 SELECT * FROM SG_UNEMPLOYMENT
53 ...
54
55 pd.read_sql(query, con).head()      Check that data has been successfully inserted by
                                         querying data and doing a quick inspection
```

### Importing of Data Tables into MySQL Database

- After cleaning of datasets to ensure data integrity based on constraints defined in CREATE TABLE statement, data loaded using INSERT INTO statement
- Once loaded, test query is run to extract table from MySQL in Python using SELECT statement and result set is visually inspected for validation

## METHODOLOGY (DATA CLEANING & MANIPULATION)

### Data Cleaning

- To facilitate analysis, datasets were cleaned using Pandas after extracting the necessary from MySQL
- Common issues encountered:
  - NANs which affects calculation of mean
  - Raw data often in wide format (not friendly for visualisation)
  - Tables with common columns unable to be merged due to inconsistent casing/punctuation

```
1 # Extract 2022 other leave data
2 query = ''
3 SELECT * FROM OTHER_LEAVE
4 WHERE YEAR = 2022
5 ...
6 other_leave = pd.read_sql(query, con)
7
8 # Data cleaning
9 other_leave.fillna(0, inplace=True)
10 other_leave_melt = other_leave.melt(id_vars=['INDUSTRY1', 'INDUSTRY2', 'INDUSTRY3'],
11                                     value_vars=['COMPASSIONATE_LEAVE', 'MARRIAGE_LEAVE',
12                                     'UNPAID_LEAVE_MORETHAN_1MTH', 'UNPAID_LEAVE_LESSTHAN_1MTH',
13                                     'STUDY_LEAVE', 'CHILD_SICK_LEAVE', 'PARENT_CARE_LEAVE'],
14                                     var_name='LEAVE_TYPE')
15 other_leave_group = other_leave_melt.groupby(['INDUSTRY1', 'INDUSTRY2', 'LEAVE_TYPE'])['value'].mean().reset_index()
16 other_leave_pivot_cleaned = pd.pivot(data=other_leave_group, index=['INDUSTRY1', 'INDUSTRY2'], columns='LEAVE_TYPE', values='value').reset_index()
```

- Fill missing NANs with 0 to make sure means are calculated correctly
- Melt wide format data frames into long format for easier manipulation
- Re-pivot into long format for specific analysis / visualisation

```
1 # Update join keys in other_leave_pivot_cleaned as they are slightly different from annual leave table
2 other_leave_pivot_cleaned['INDUSTRY2'].replace('electronic and computer products', 'electronic, computer and optical products', inplace=True)
3 other_leave_pivot_cleaned['INDUSTRY2'].replace('fabricated metal products and machinery', 'fabricated metal products, machinery and equipment', inplace=True)
```

### Data Manipulation

- Joining relevant data tables for comprehensive analysis
  - e.g., Left join Annual Leave and Other Leave tables on INDUSTRY column to analyse overall leave package
- Aggregating data to calculate means at different levels for analysis

	INDUSTRY1	Join key	INDUSTRY2	14 days & below	15 to 21 days	over 21 days	Annual leave data	Other leave data
0	construction		construction	90.6	8.7	0.7	50.9	78.2
1	manufacturing		electronic, computer and optical products	51.7	44.8	3.6	35.1	100.0
2	manufacturing		fabricated metal products, machinery and equip...	70.4	26.6	3.0	47.8	93.7
3	manufacturing		food, beverages and tobacco	78.3	17.4	4.3	46.7	84.3
4	manufacturing		other manufacturing industries	56.5	39.6	4.0	46.4	93.2

```
1 # Merge annual_leave and other_leave for analysis
2 df = pd.merge(annual_leave_pivot_cleaned, other_leave_pivot_cleaned, on='INDUSTRY2', how='left', suffixes=('', '_OL'))
```

### Dimensionality Reduction

- Used Principal Component Analysis (PCA) to reduce number of dimensions (fields)
- Allows for plotting of data on a 2-D axis for analysis while preserving information from raw data

```
1 turnover_group = turnover.groupby(['YEAR', 'INDUSTRY1', 'INDUSTRY2']).agg(RECRUITMENT_MEAN = ('RECRUITMENT', 'mean'),
2 RESIGNATION_MEAN = ('RESIGNATION', 'mean'),
3 TURNOVER_MEAN = ('TURNOVER_RATE', 'mean')).reset_index()
```

```
1 # Normalise and apply dimensionality reduction to 2 dimensions only for plotting on scatterplot
2 pca = PCA(n_components=2)
3 df_scaled = StandardScaler().fit_transform(df_tidy)
4 df_pca = pca.fit_transform(df_scaled)
```

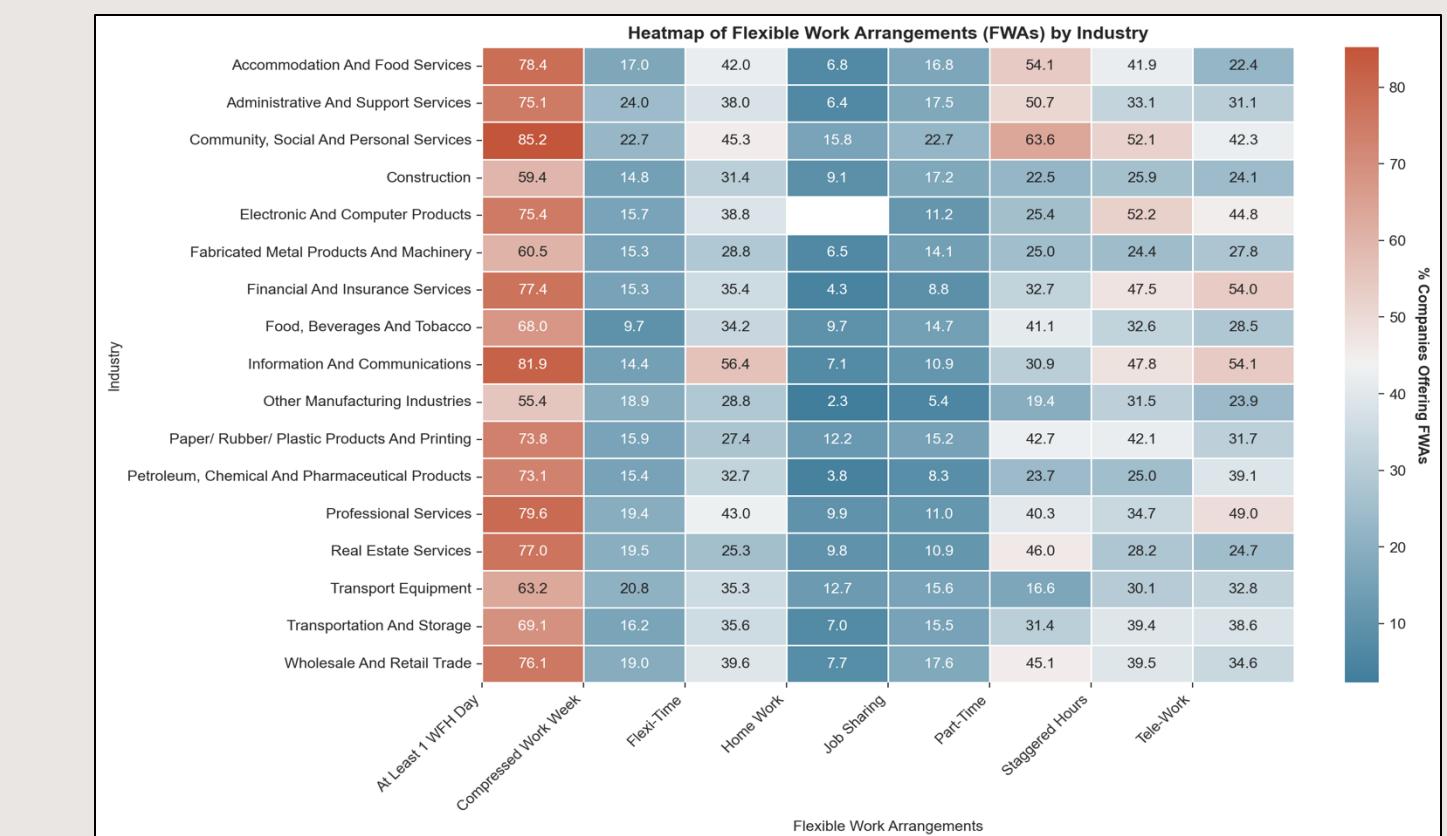
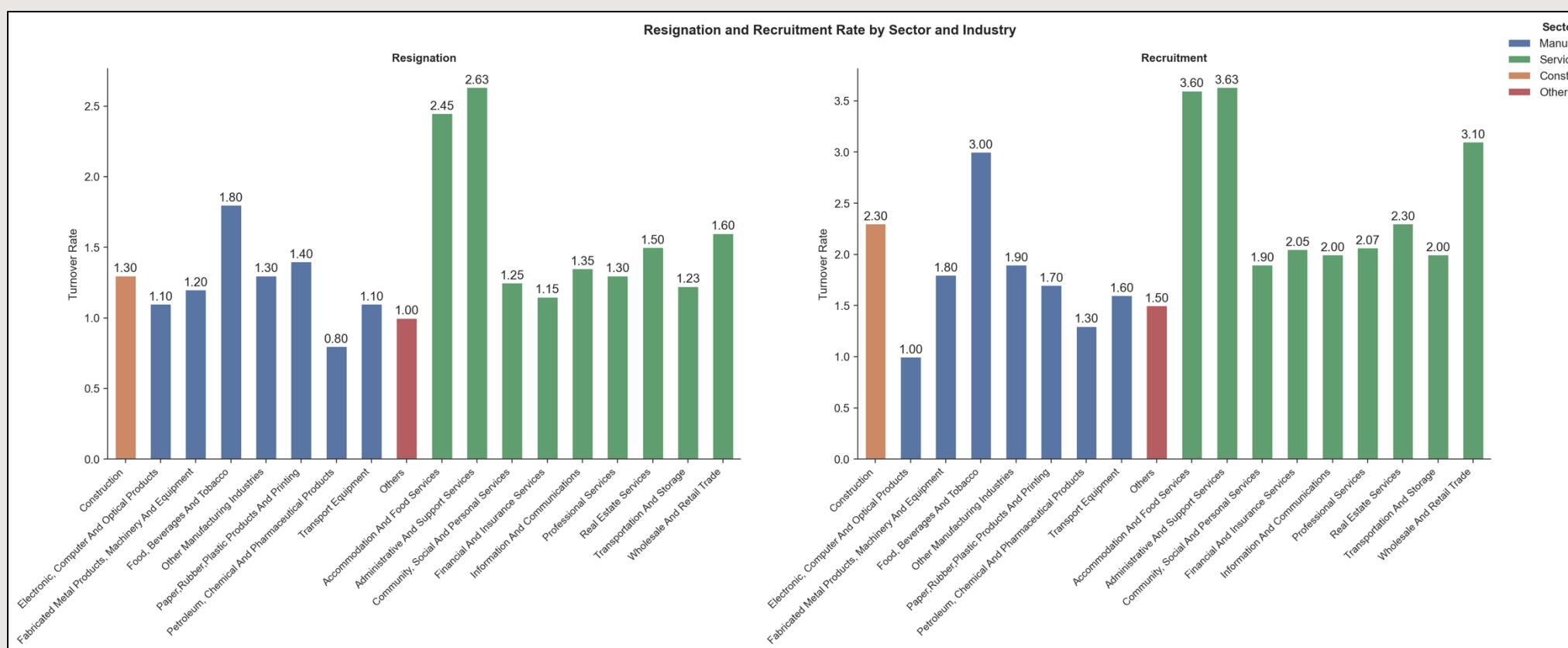
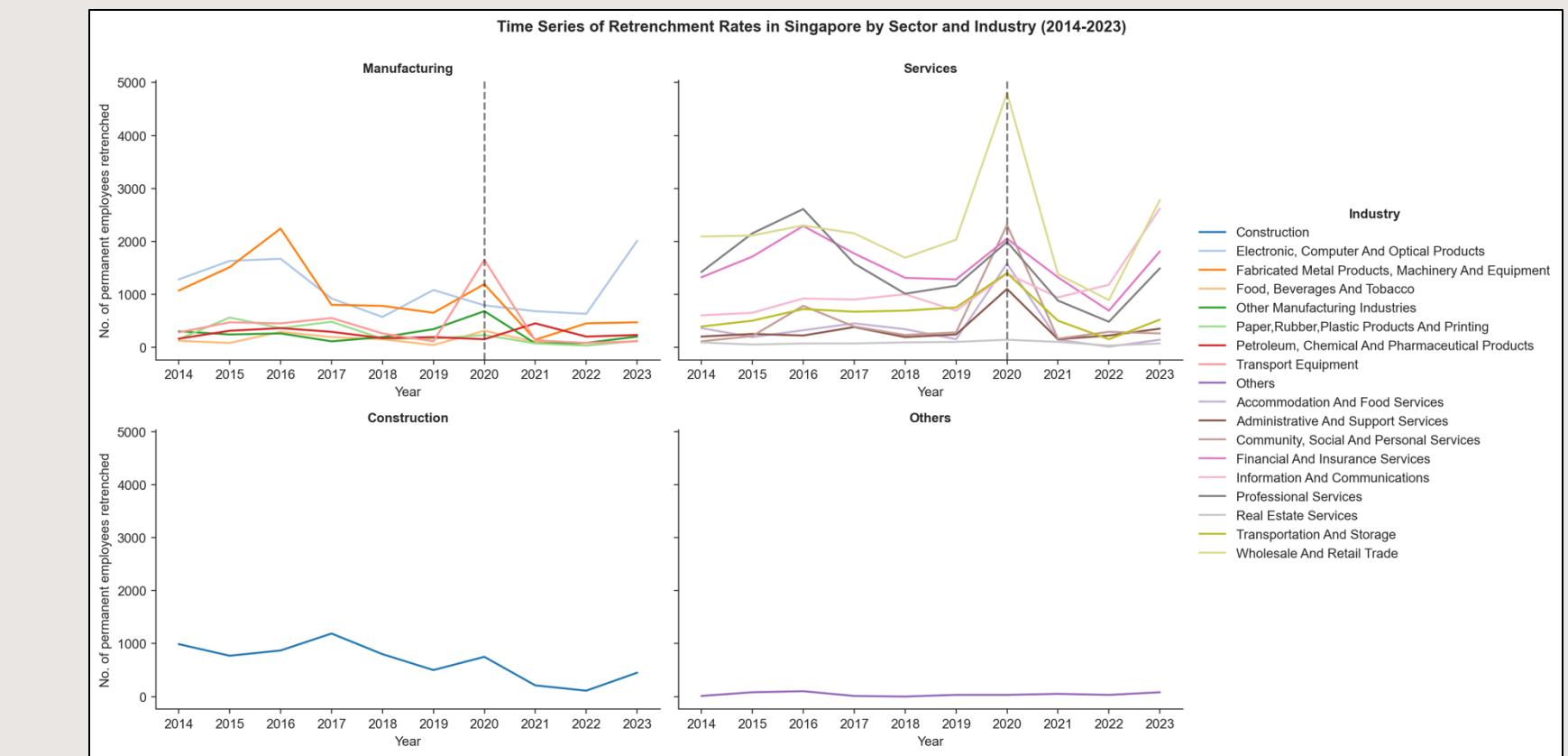
# METHODOLOGY (ANALYSIS)

## Data Visualisation

- Used Seaborn, Plotly to analyse different facets of the labour market

- Types of charts used in this analysis:

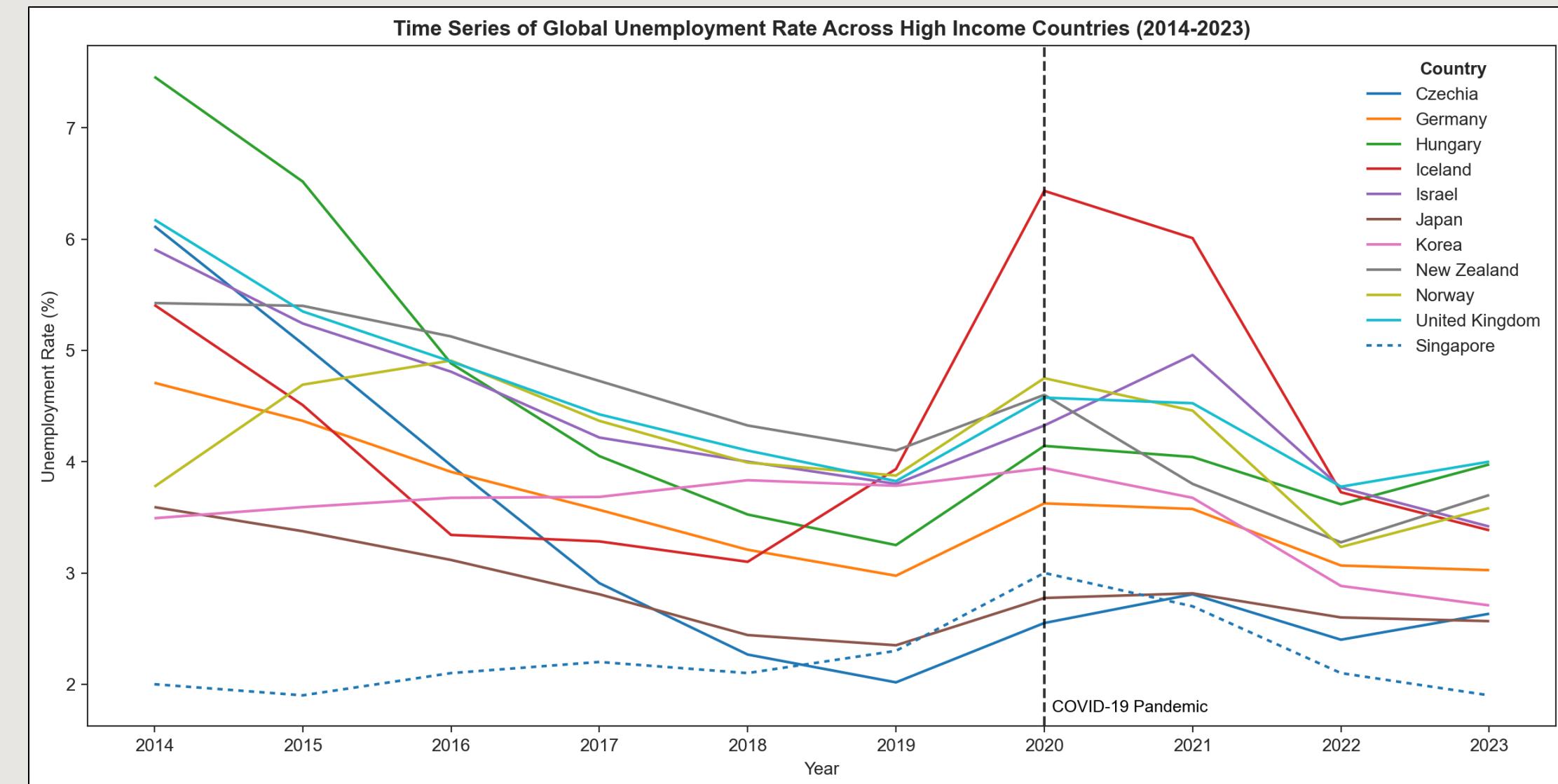
- Line chart** (for time series)
- Pie chart** (for comparison of proportions)
- Bar chart** (for comparison of means across different categories)
- Scatter plot** (for PCA analysis)
- Heat map** (for comparison of multiple dimensions across different categories)



**What is the current state of Singapore's  
labour market in comparison to other  
countries?**

## FINDINGS (TOPLINE LABOUR MARKET COMPARISON)

Relative to the top 10 high income countries with the lowest average unemployment rates, Singapore consistently has one of the lowest unemployment rates



**Which sectors or industries in Singapore offer  
the greatest job security in the present  
economic climate?**

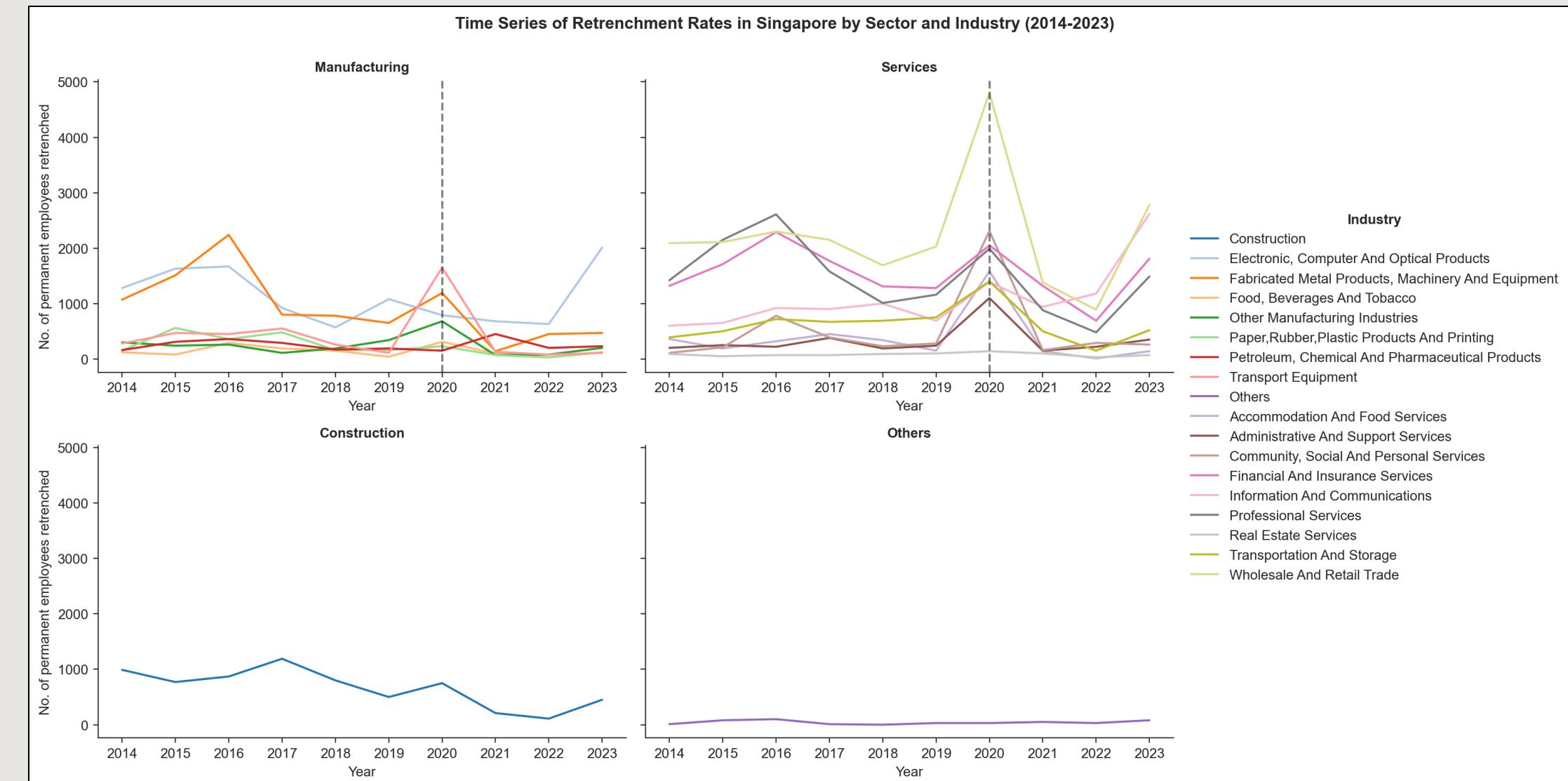
## FINDINGS (JOB SECURITY)

### Analysis

Time series to allow job-seekers to identify industries that are thriving or not doing well

### Insights

- Certain industries continue to see rising retrenchment levels, possibly indicating the decline (either short/long-term) of these industries
- Most of the decline appears to be in the Services sector (IT, Wholesale & Retail Trade, Finance & Insurance industries)
- Spike in 2020 due to COVID-19



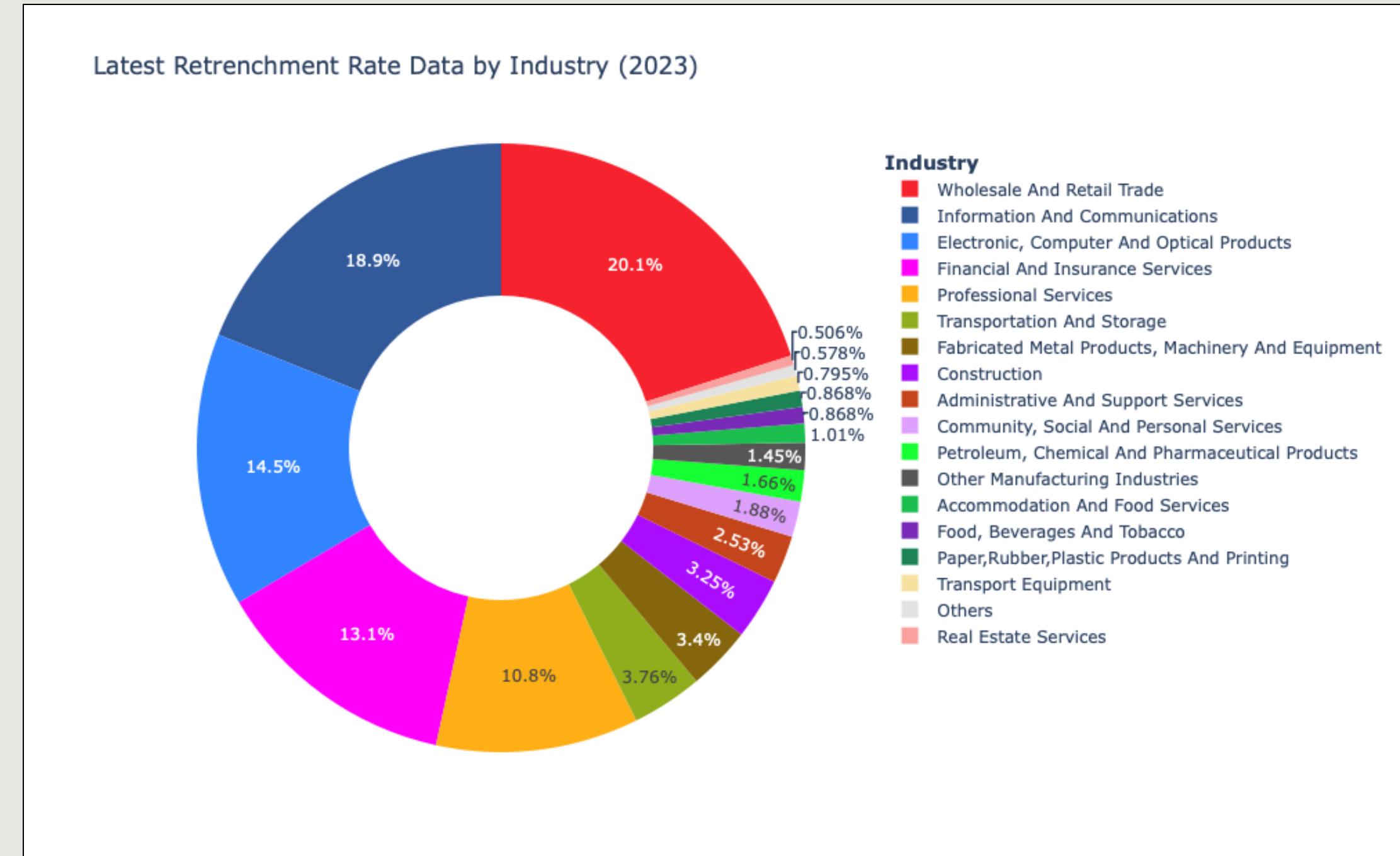
## FINDINGS (JOB SECURITY)

### Analysis

Pie chart to examine proportion and absolute retrenchment of each industry in latest year

### Insights

- With a focus on immediate job security, values from 2023 reflect current job market condition across industries
- Majority of more 'stable' industries are within the Manufacturing sector
- Within the Services sector, industries such as Accoms. & Food Services; Admin & Support Services; Community, Social & Personal Services are doing better



**Where can job-seekers look for  
opportunities within the various sectors or  
industries?**

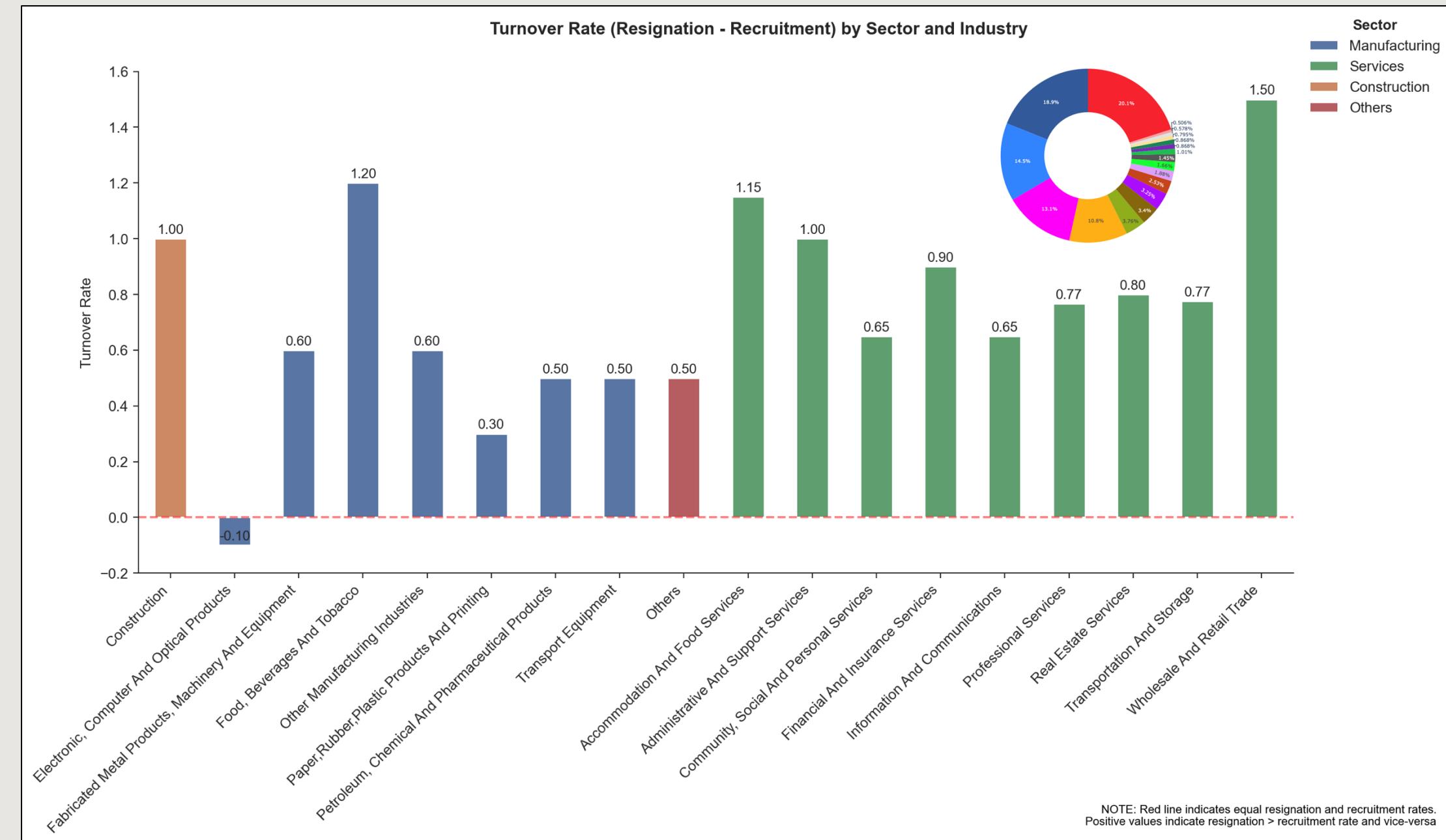
## FINDINGS (JOB OPPORTUNITIES)

### Analysis

Comparison of latest turnover rate across industries to elucidate job opportunities

### Insights

- Resignations and recruitment signals job availability, consider overall turnover rate to guide job search (positive = opportunities available)
- Wholesale & Retail Trade has low resignation but high recruitment, indicating industry growth
- However, previous analysis shows high retrenchment, implying restructuring within industry — reducing redundancy, focus on more strategic roles



**Caveat:** Important to note that positive turnover (i.e., high recruitment) not entirely indicative of residential employment. Job-seekers should be mindful of prevailing market conditions and labour regulations / laws which are not considered in this analysis

**How can job-seekers find employment that  
aligns with their expectations and offers  
satisfaction?**

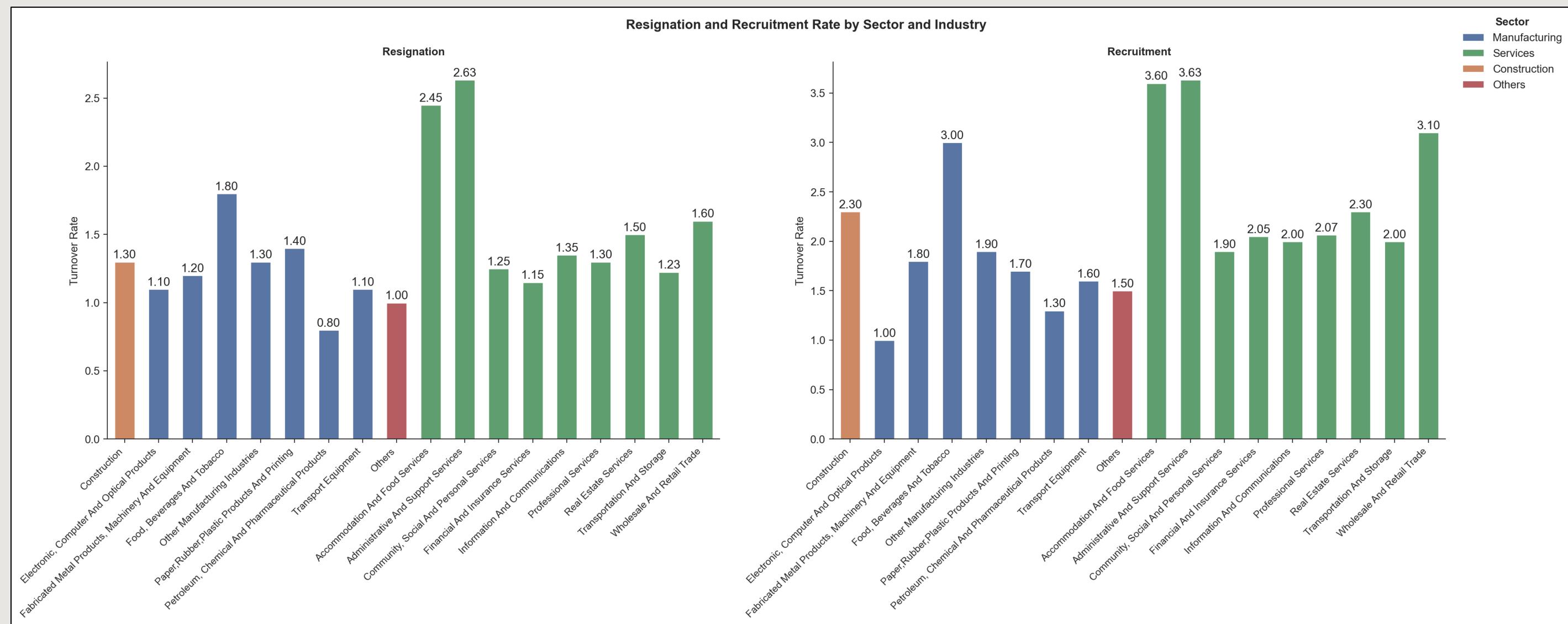
## FINDINGS (JOB SATISFACTION)

### Analysis

Breakdown of turnover into resignation and recruitment across industries for granularity as resignations serve as proxy for satisfaction, indicating labour-management relations

### Insights

- High correlation between resignation and recruitment ( $r = 0.913, p \approx 0.00$ )
- Accoms. & Food Services and Admin. & Support Services industries indicative of low job satisfaction
- Other than resignations, important to consider other indicators



#### Caveats:

- Important to note that positive turnover (i.e., high recruitment) not entirely indicative of residential employment. Job-seekers should be mindful of prevailing market conditions and labour regulations / laws which are not considered in this analysis
- Low resignations not entirely indicative of high satisfaction. Job-seekers to keep in mind that there are other indicators of satisfaction which are not explored in this analysis.

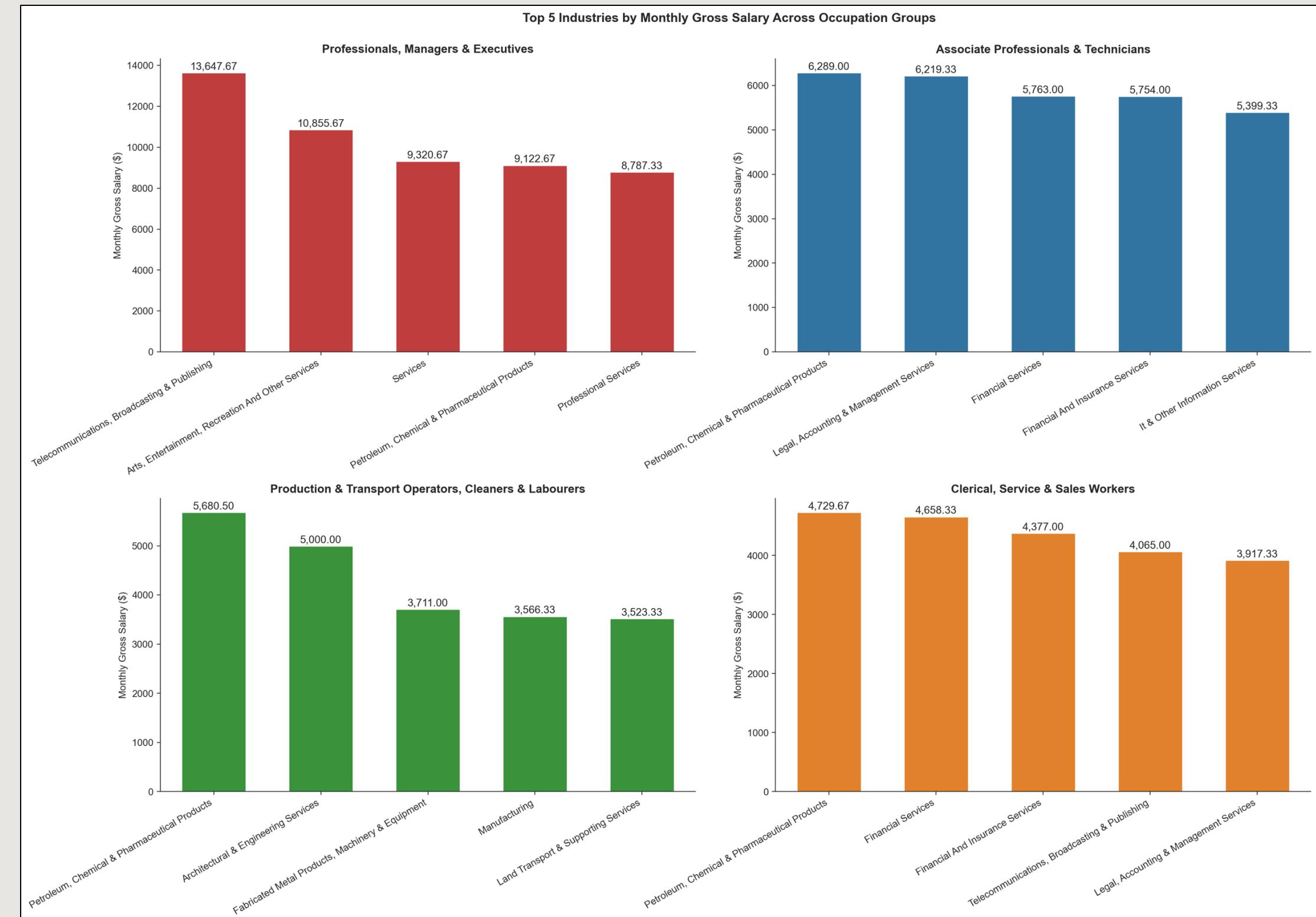
## FINDINGS (WAGES)

### Analysis

Comparison of top-earning industries across occupation groups to identify opportunities

### Insights

- PMEs have highest wages generally, especially from Telecommunications, Broadcasting & Publishing industry
  - Sales Manager
  - ICT Sales & Services Professional
  - Systems Analyst
- Across remaining occupation groups, Petroleum industry the most lucrative
  - Engineering Technician
  - Supervisor / Foreman
  - Office / Information / Production Clerk



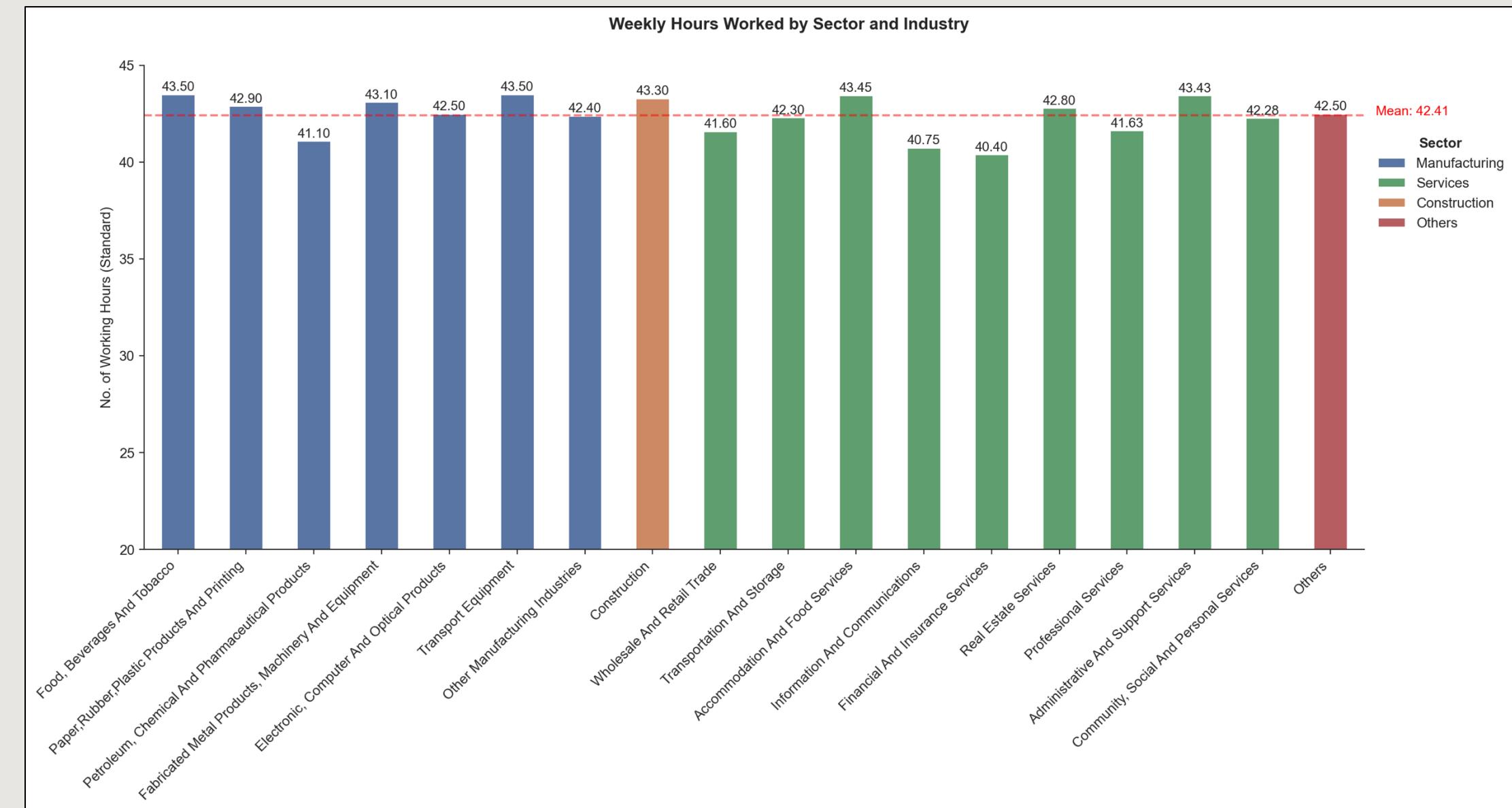
## FINDINGS (WORK-LIFE BALANCE)

### Analysis

Comparison of weekly working hours across industries and sectors to work-life balance

### Insights

- Fairly consistent working hours across industries, Services sector has the lowest mean working hours (42.1 hours)
- E.g., IT, Financial & Insurance Services industries work 2 hours less than average
- Transport Equipment, Construction, and Administrative & Support Services industries work 1 hour more than average



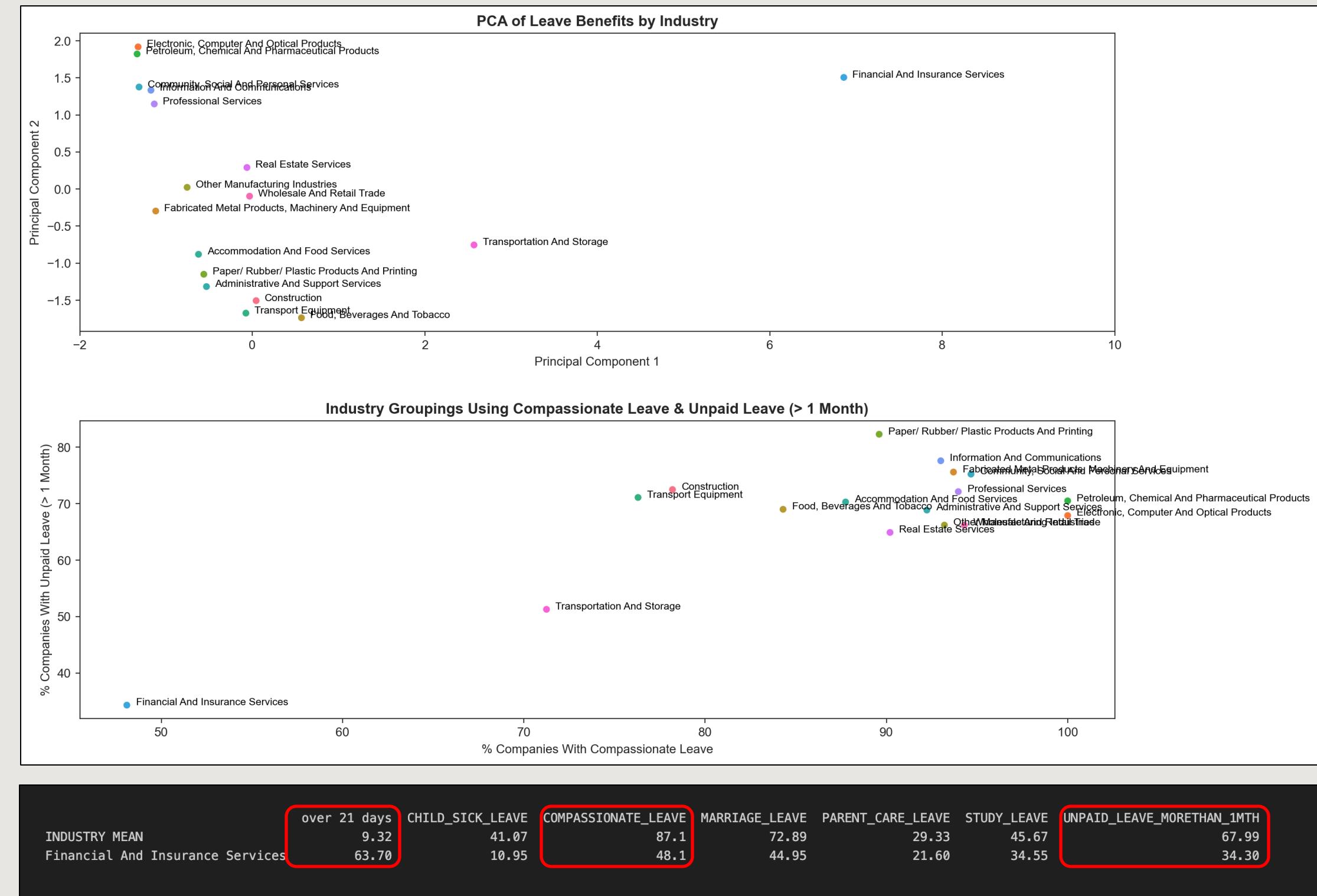
## FINDINGS (LEAVE PACKAGE)

### Analysis

PCA simplifies information on different leave types across industries into two principal components (PCs) for easy comparison of relative industry groupings by leave package

### Insights

- Compassionate and unpaid leave biggest differentiators, can be used as a start-point to examine differences in leave packages
- E.g., Financial & Insurance Services industry low compassionate and unpaid leave but also highest annual leave compared to mean
- Job-seekers can use this approach to compare leave package across industries based on groups



## FINDINGS (FLEXIBLE WORKING ARRANGEMENTS)

### Analysis

Comparison of FWAs across industries to determine industries with greatest flexibility

### Insights

- Most companies in Singapore offer WFH
- Compressed work week not yet the norm, with full-time home-working and job-sharing being rare across all industries
- Part-time work prevalent in Services sector
- Staggered hours and tele-work (remote working) considerably higher in tertiary industries such as IT, Financial and Insurance Services, Electronic & Computer Products



## KEY TAKEAWAYS

1

Singapore is doing well compared to other countries

Despite concerns and fears, data shows Singapore's job market outperforms other countries (things could be worse)

2

Job-seekers must balance job security and satisfaction expectations

In a tight job market, job-seekers should prioritise job security and be ready to seize opportunities when conditions improve

3

IT sector (us) probably not doing very well, benefits are good though

2<sup>nd</sup> highest retrenchment levels in 2023, little resignations and thin recruitment opportunities 😞

## CONCLUSION

# What Now?

Despite economic challenges, Singapore's job market demonstrates resilience and performs well compared to other countries. While the pursuit of job benefits is important, job-seekers need to carefully and strategically balance with job security in both the short and long-term.

While the labour market within Singapore may be tight now, the potential for growth opportunities exists. Staying adaptable and prepared will be crucial for capitalising on future opportunities when markets recover and rebound.



# THANK YOU