

# Web Analytics Final Project (Group 13)

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# Project Idea

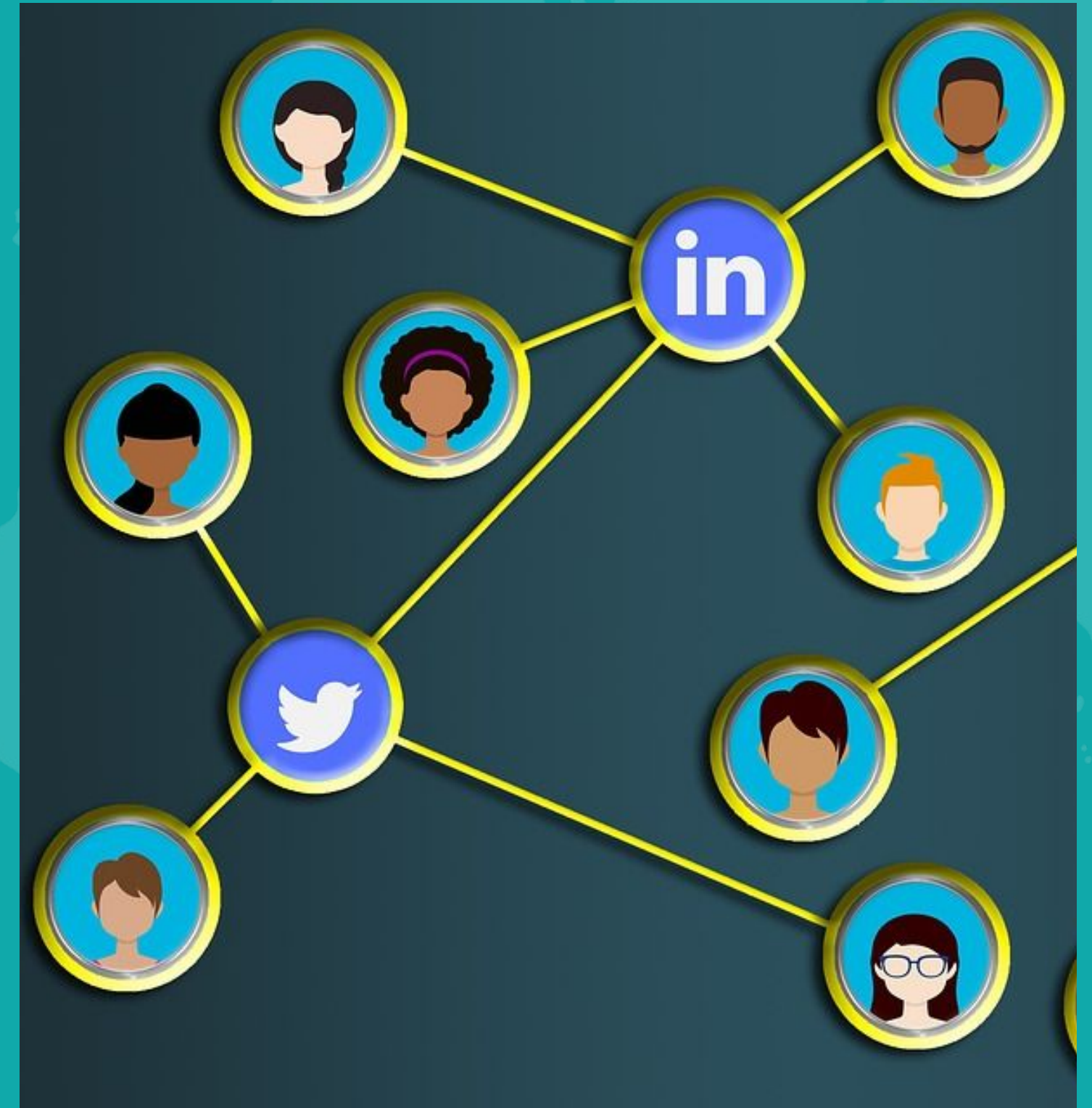
- **Create a network and career-growth platform for students and recent graduates**
  - ◆ Offers personalized job matches and tracks application history
  - ◆ Provides detailed feedback/advice to help users improve their chances for each applications (skills, keywords, other users related to job)





# Goal and Motivation

- To encourage students and recent graduates to network
- To help students find jobs and opportunities based off their skills and profile
- Students struggle with gaining experience related to their major and skills
- Helps address the need for students to find relevant jobs and opportunities that fit their studies and skill set



# State of the Art

→ Existing Platforms and APIs:

- ◆ LinkedIn
- ◆ Indeed
- ◆ Glassdoor

**Key Features:** filter by role, location, experience level; real time data

The LinkedIn logo, featuring the word "Linked" in white and "in" in white inside a white square, all on a blue background.The Indeed logo, featuring the word "indeed" in white lowercase letters with a small white arc above the "i", all on a dark blue background.The Glassdoor logo, featuring a green stylized "G" icon above the word "glassdoor" in green lowercase letters, all on a white background.



# State of the Art

## → Data analysis techniques

- ◆ **Natural Language Processing (NLP):**  
Extract skills, keywords, and trends from job descriptions
  - **Example:** presence of the word “Python” in engineering job descriptions
- ◆ **Machine Learning:** Matches users to jobs; improves recommendations with time



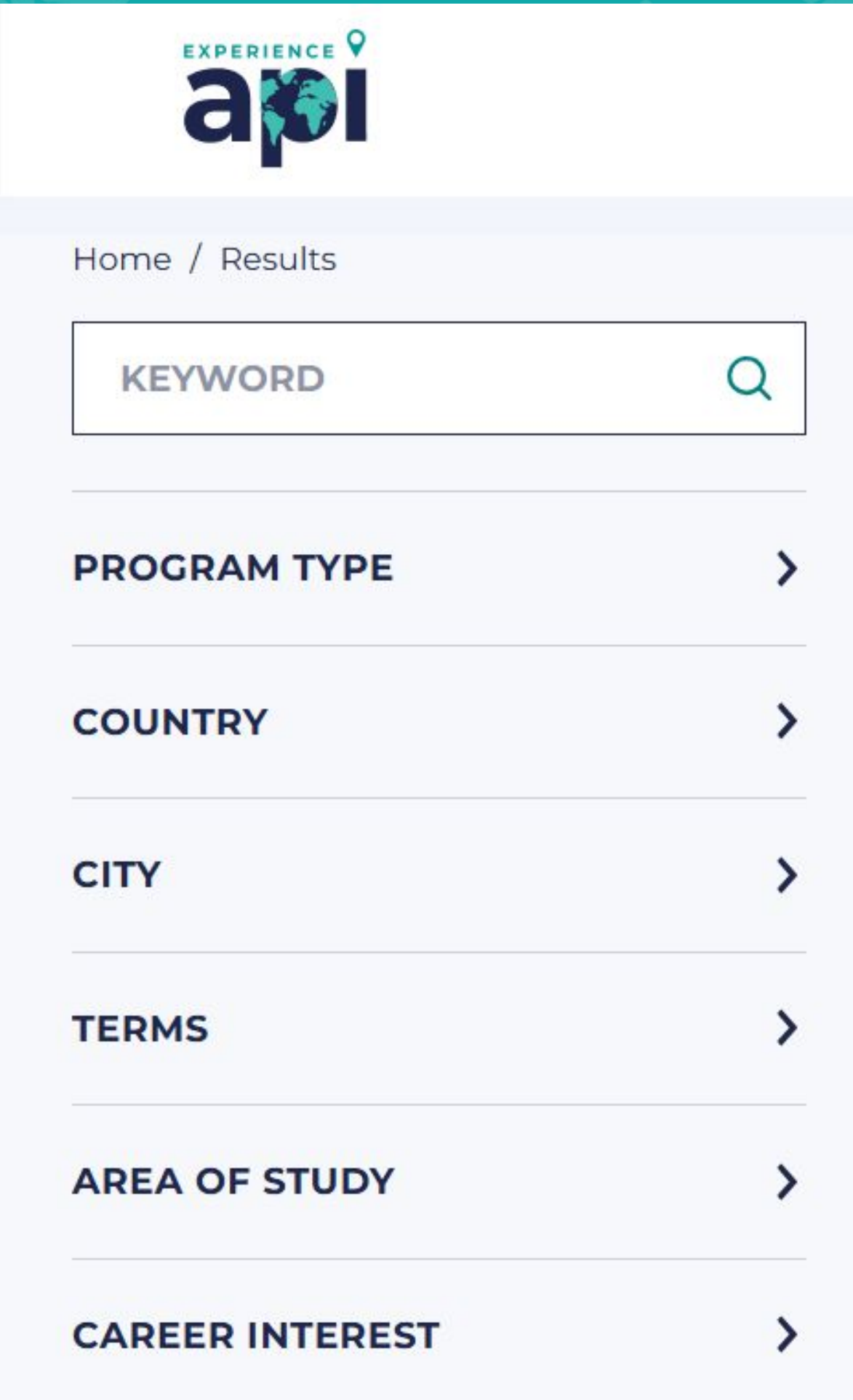
# State of the Art

## → Visualization and Insights

- ◆ **Interactive Dashboards:** Customizable filters for users preferences
- ◆ **Company Insights:** Salary reports, interview experiences, total job openings

## → Emerging techniques

- ◆ **Graph Analysis:** Modeling relationships between job categories, companies, industries
- ◆ **Chatbots:** Guide users, refine search, enhance user engagement



The screenshot displays the 'EXPERIENCE api' web application. At the top, the logo features the word 'EXPERIENCE' in a small font above 'api' in a large, bold font, with a location pin icon over the 'i'. Below the logo is a breadcrumb trail 'Home / Results'. A search bar with the placeholder text 'KEYWORD' and a magnifying glass icon is positioned next to it. Below the search bar, there are seven filter categories, each with a horizontal line and a right-pointing chevron: 'PROGRAM TYPE', 'COUNTRY', 'CITY', 'TERMS', 'AREA OF STUDY', and 'CAREER INTEREST'.



# Our Data Source

→ **Adzuna API:** It provides real-time job listings

## **The data includes:**

- Job title
- Location
- Salary information
- Company
- Job description



# Data Gathering Techniques

## → API Requests:

- ◆ **Base url:** `https://api.adzuna.com/v1/api/jobs`
- ◆ **Example of request:**  
`gb/search/1?app_id=my_app_id&app_key=my_app_key&results_per_page=50&where=London&what=Engineer`
- ◆ **We send a get request to the API**
  - Every time we make a query the response is returned as an object serialized using JSON

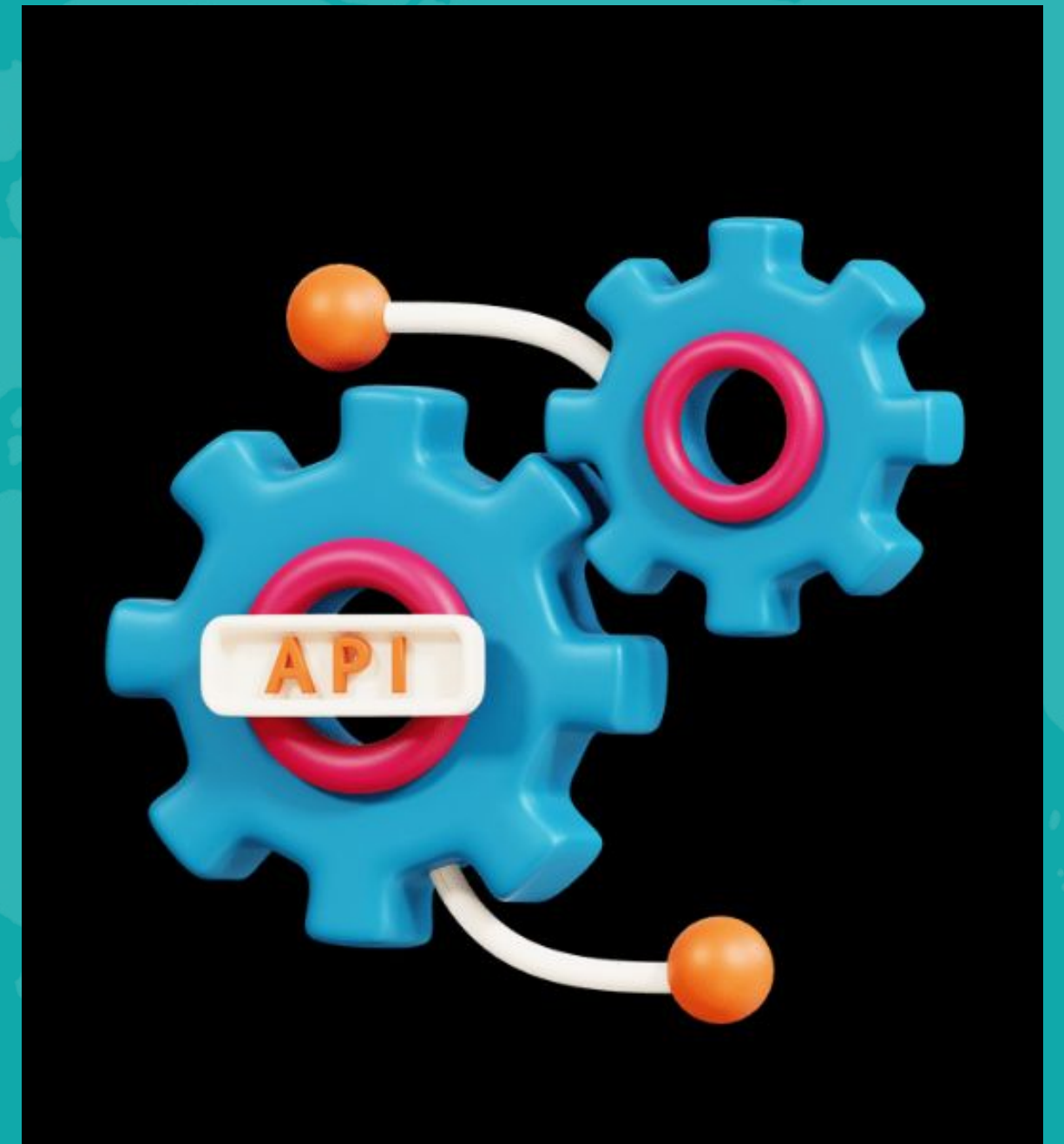




# Data Gathering Techniques

## → How it works:

- ◆ **Base URL:** Specifies the endpoint for job data
- ◆ **Query Parameters:**
  - **app\_id:** Your unique application ID
  - **app\_key:** Your unique API key
  - **results\_per\_page:** Number of job results to retrieve per request
  - **where:** Location (e.g., London)
  - **what:** Job title or keyword (e.g., Engineer)



# Data Gathering Techniques

## → Data Processing Workflow

### 1. Data Extraction:

- Extract the JSON element from the API response for further processing

### 2. Data Cleaning:

- Handle missing values
- Convert numeric fields (e.g., salaries) to integers or floats for accurate calculations

### 3. Data Storage:

- Use Pandas to save the cleaned data into CSV files

```
{
  "__CLASS__": "Adzuna::API::Response::JobGeoData",
  "locations": [
    {
      "__CLASS__": "Adzuna::API::Response::LocationJobs",
      "count": 10978,
      "location": {
        "__CLASS__": "Adzuna::API::Response::Location",
        "area": [
          "UK",
          "West Midlands",
          "Birmingham"
        ],
        "display_name": "Birmingham, West Midlands"
      }
    },
    .... more child locations here ...
  ]
}
```



# Problems found/Possible issues

- **API limitation:** 1000 requests/hour
- **Inconsistent Data:** Some job listings lack salary information, or latitude and longitude information
- **API errors:** Some errors (error 503) could occur, requiring retries
- **Presence of duplicates**
- **Jobs offers might not be available for certain regions**

|   | title                              |
|---|------------------------------------|
| 0 | DAML smart contract Developer      |
| 1 | Dual Fuel Smart Meter Engineer     |
| 2 | Dual Fuel Smart Meter Engineer     |
| 3 | Dual Fuel Smart Meter Engineer     |
| 4 | Dual Fuel Smart Meter Engineer     |
| 5 | Controls Engineer                  |
| 6 | Sales Engineer/Pre - Sales         |
| 7 | Technical Escalation Engineer      |
| 8 | Customer Support Engineer - Onsite |
| 9 | Customer Support Engineer          |

# Data Analysis Techniques

- We used natural language processing (NLP) to calculate a similarity score between the user's profile and the job description:
- ◆ We converted the job descriptions and the user's profile in numerical vectors based on word frequency and importance
- ◆ Then we compared the vector representing the user's profile and the job descriptions





# Problems faced/Potential issues

## → Stop words used for NLP:

- ◆ This implies that we can only calculate similarity scores for languages of chosen stop words (here English)

## → Data aggregation:

- ◆ Careful handling of the dataset to ensure no duplicates or errors and making it appropriate for visualisation

### Stop Words

- a
- of
- on
- I
- for
- with
- the
- at
- from
- in
- to

# Problems faced/Potential issues

## → Difficulty to observe long time trends

- ◆ The only tool given to us by the API is an history feature that returns the salary for given job category and location over a time period

**Ex: it-jobs in New-York the response is :**

```
"month": {  
  "2024-07": 65025.13,  
  "2024-09": 78193.44,  
  "2024-03": 71314.31,  
  "2024-04": 69694.3,  
  "2024-06": 62789.18,  
  "2024-10": 78704.29,  
  "2024-05": 57984.3,  
  "2024-02": 70137.13,  
  "2024-11": 75506.25,  
  "2024-01": 69238.5,  
  "2024-08": 74562.62,  
  "2023-12": 70545  
},
```





# Data Visualization

→ We created a Folium map that shows the jobs having the best similarities for a given user

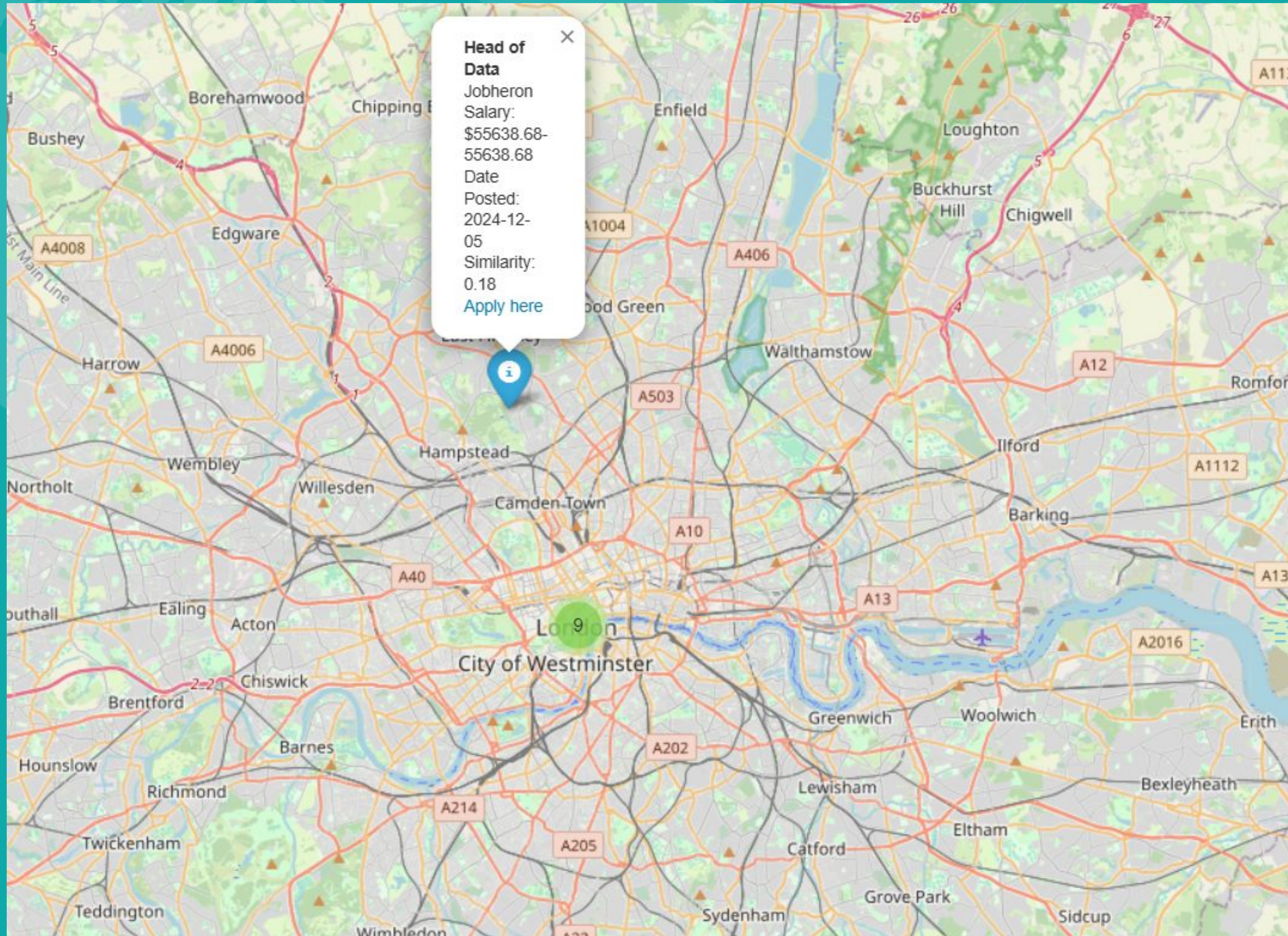
◆ Example for the following user:

```
user_profile = {  
    "field": "Data",  
    "experience": "Senior",  
    "location": "London",  
    "min_salary": 30000,  
    "max_salary": 150000,  
    "category": "IT Jobs",  
    "date_posted_within_days": 7  
}
```





# Data Visualization





# Data Visualization

- Creation of a Google dashboard:
  - A barplot to visualize the median salaries for certain job offers in different cities
  - A boxplot to visualize the salaries for certain job offers in different cities

| Data                                 |  |  |
|--------------------------------------|--|--|
| Search                               |  |  |
| job_trends_us.csv                    |  |  |
| raw_job_data (1).csv                 |  |  |
| median_job_data (2).csv              |  |  |
| job_data.csv                         |  |  |
| median_job_data.csv                  |  |  |
| updated_historical_salary_data - ... |  |  |
| updated_raw_job_data - updated...    |  |  |
| updated_median_job_data - upd...     |  |  |

# Data Visualization (Software Engineer)

Job Title: Software Engineer

(1) ▾

Median Salary by City (USD)



Boxplot of Salary by City (USD)





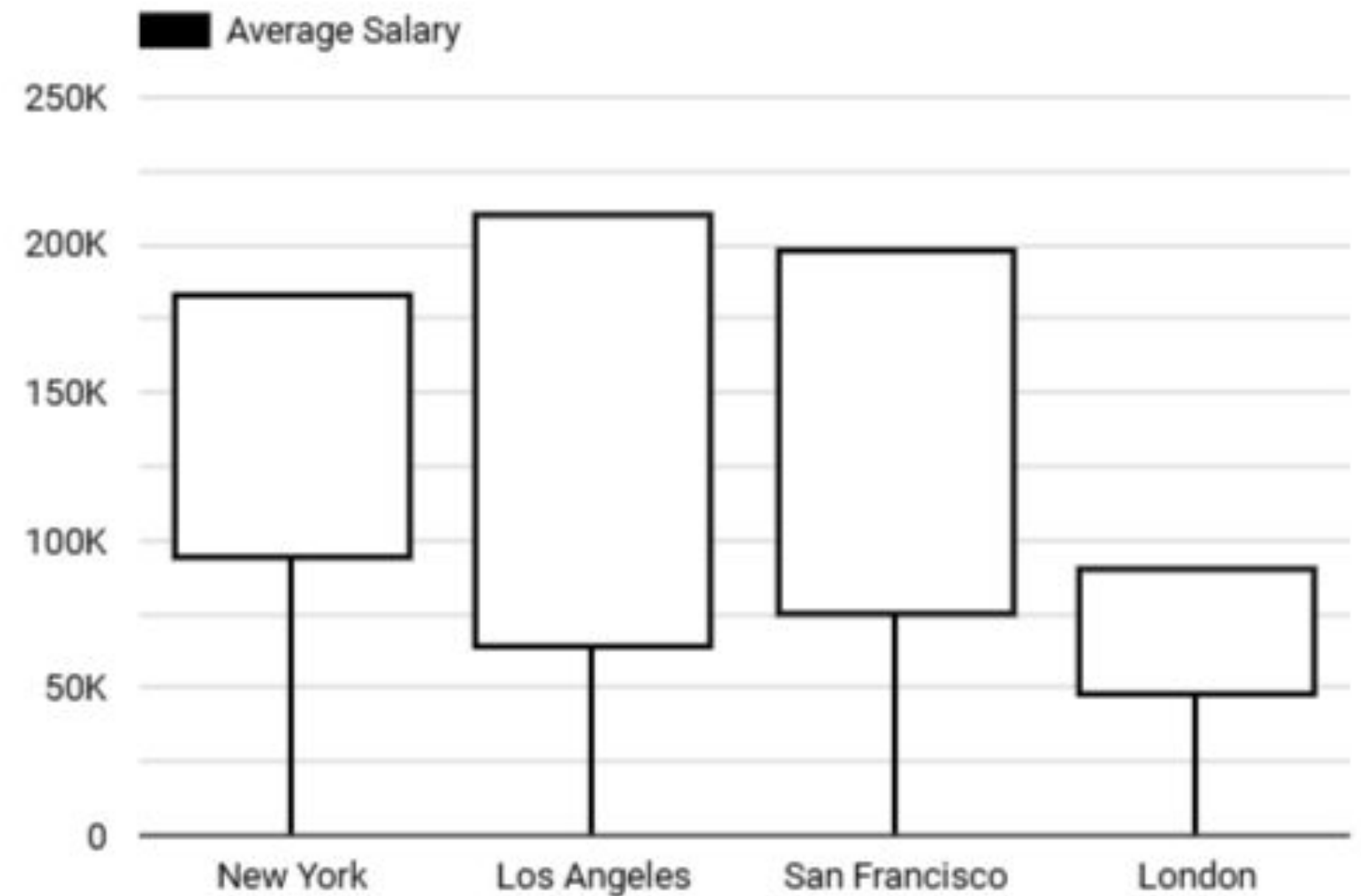
# Data Visualization (Cybersecurity Specialist)

Job Title: Cybersecurity Specialist (1) ▾

Median Salary by City (USD)



Boxplot of Salary by City (USD)



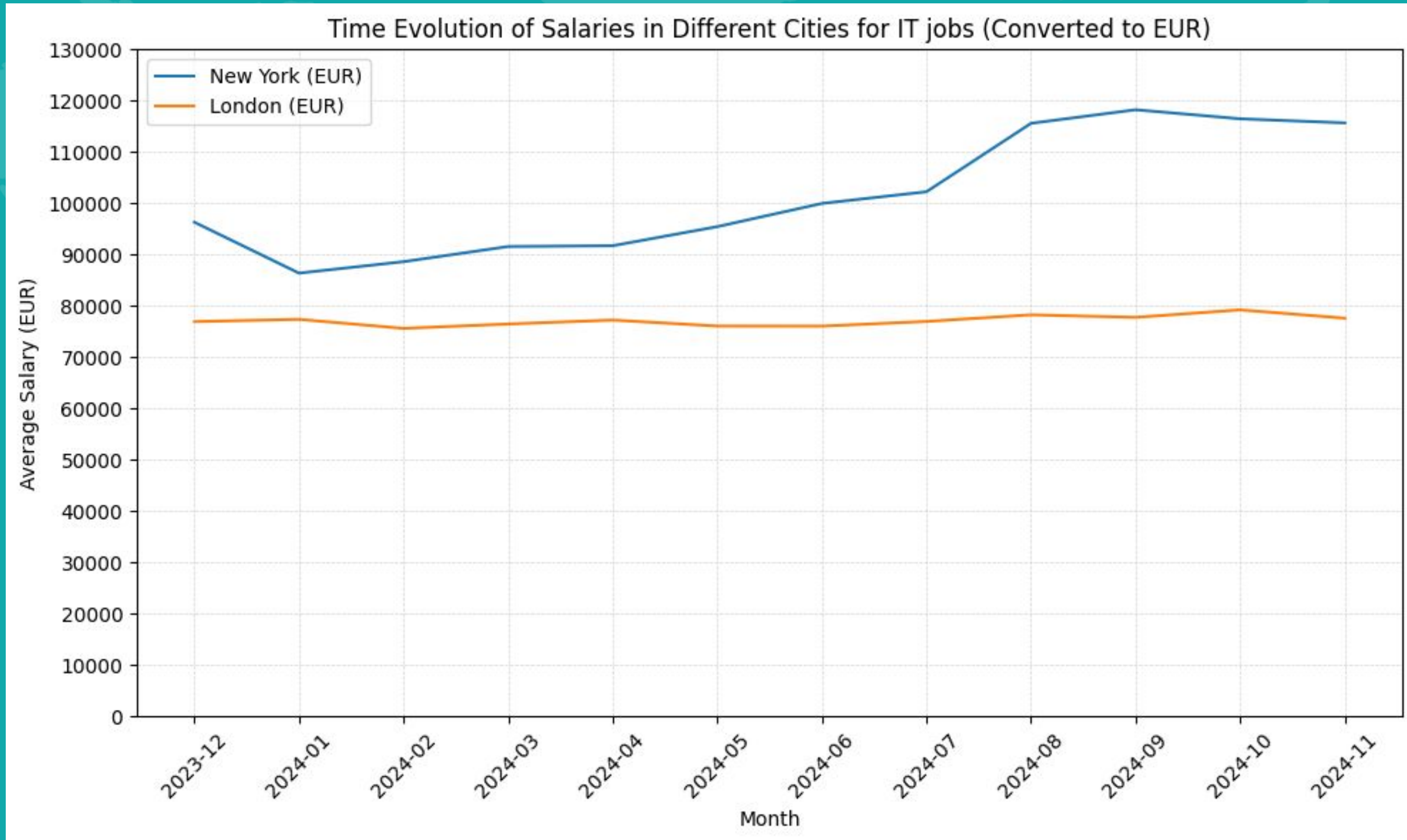
# Data Visualization

- We also created a plot that represents the time evolution of salaries for a given job category and some given location

**For example: it-jobs in London and New York**



# Data Visualization



# Did we meet our goals?

→ Yes, most of them.



- A platform that allows user to find the best job
- Provides useful information on jobs
- Sadly we don't have accurate trends over time for a given/specific job or detailed feedback for applicants



# Main Takeaways

→ **The salary distribution varies a lot across cities**

- ◆ We can see specific data trends for overall jobs and salaries, which is more clear compared to other state-of-the-art solutions (e.g., LinkedIn)
  - We also provide exact locations/geographic info

→ **Importance of NLP for job matching**

→ **The data gathered provides useful insights for everyday users**

- ◆ See specific locations and salaries

# Future Work

## → **Expand data collection**

- ◆ Expand to other countries around the world
- ◆ Other languages besides English

## → **Incorporate more data sources:**

- ◆ Maybe incorporate data from different APIs

## → **Machine learning**

- ◆ Techniques such as more complex NLP
  - Personalized job recommendations given a specific prompt



# Any questions?

