

Web Analytics Final Project (Group 13)

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Project Idea

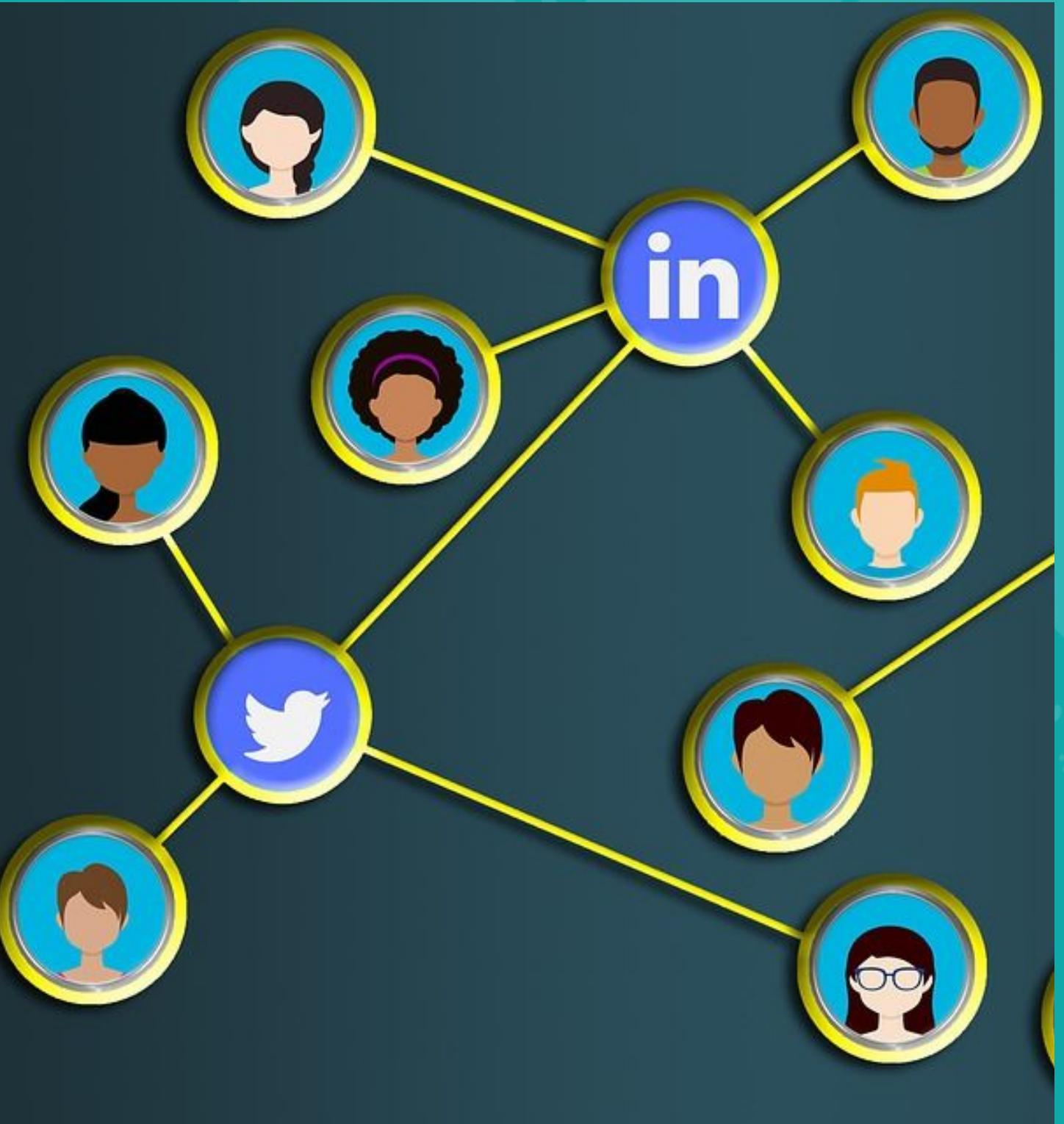
→ Create a network and career-growth platform for students and recent graduates

- ◆ 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



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



Home / Results

KEYWORD



PROGRAM TYPE >

COUNTRY >

CITY >

TERMS >

AREA OF STUDY >

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:

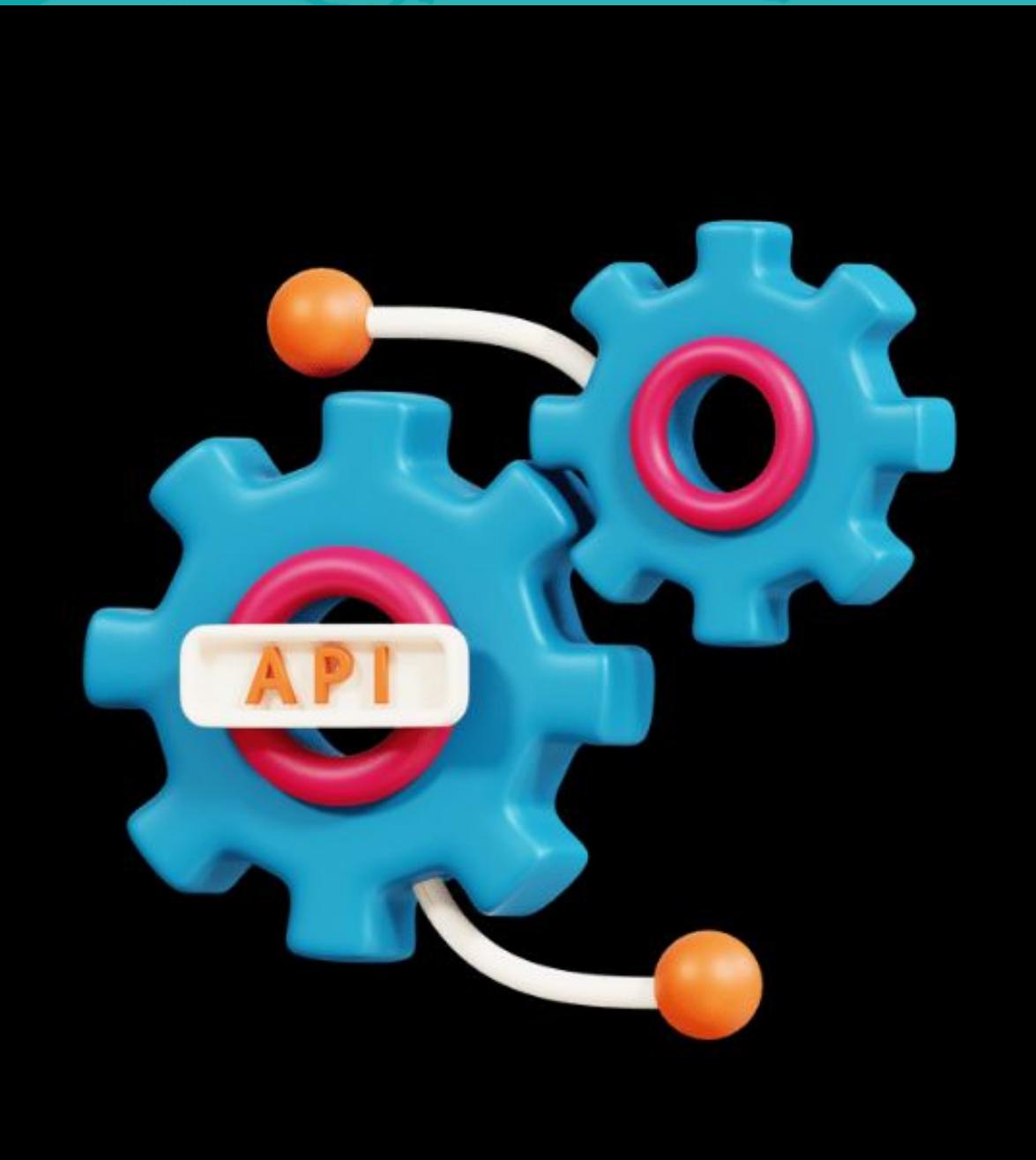
- ◆ **Base url:** <https://api.adzuna.com/v1/api/jobs>
- ◆ **Example of request:**
gb/search/1?app_id=my_app_id&app_key= my_app_key&results_per_page=50&where= London&what=Engineer
- ◆ **We send a get request to the API**
 - Every time we make a query the response is returned as an object serialized using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

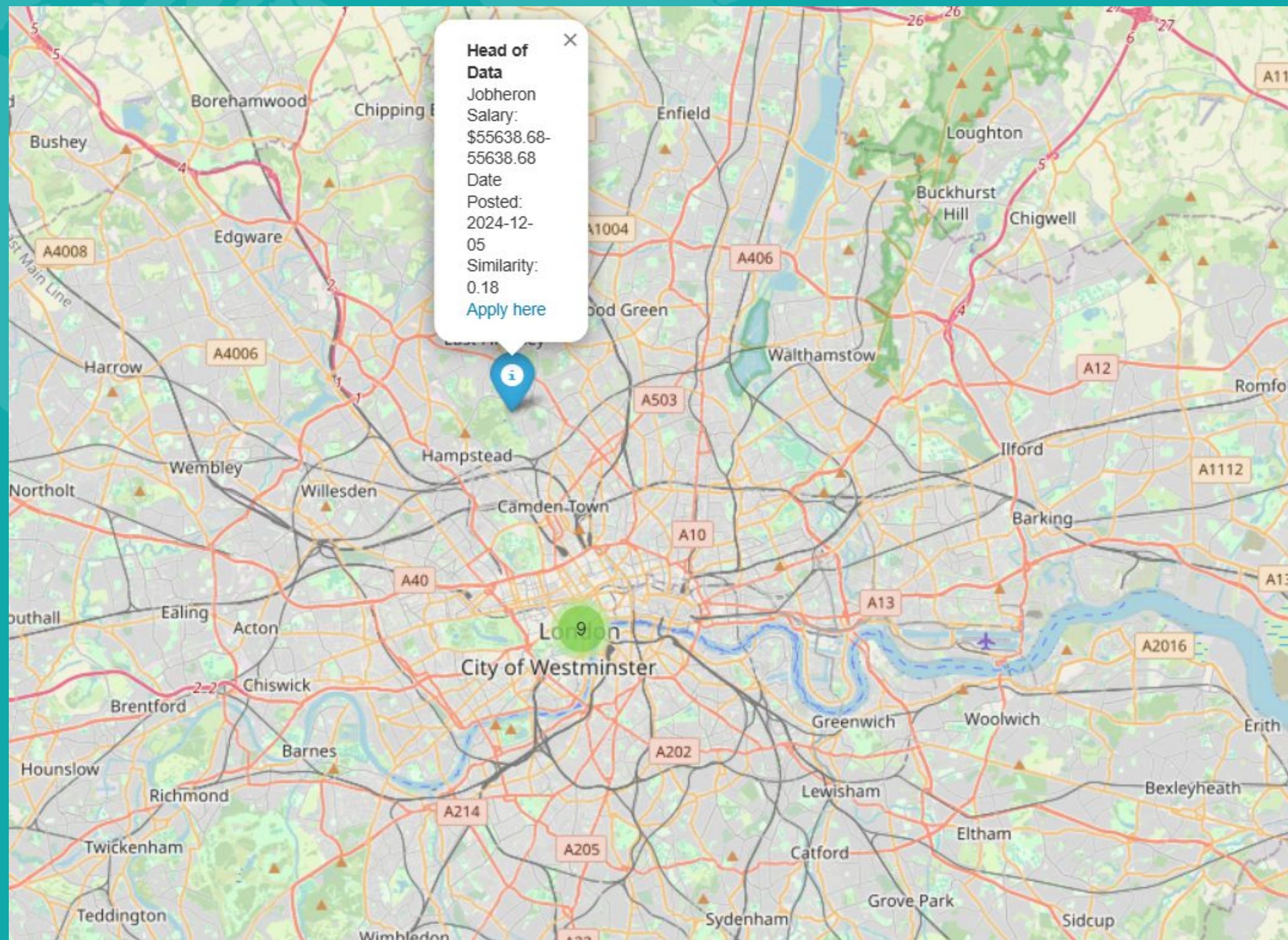
Data Visualization

- 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	
<input type="text"/>	Search
<input type="button"/> job_trends_us.csv	<input type="button"/>
<input type="button"/> raw_job_data (1).csv	<input type="button"/>
<input type="button"/> median_job_data (2).csv	<input type="button"/>
<input type="button"/> job_data.csv	<input type="button"/>
<input type="button"/> median_job_data.csv	<input type="button"/>
<input type="button"/> updated_historical_salary_data - ...	<input type="button"/>
<input type="button"/> updated_raw_job_data - updated...	<input type="button"/>
<input type="button"/> updated_median_job_data - upd...	<input type="button"/>

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)

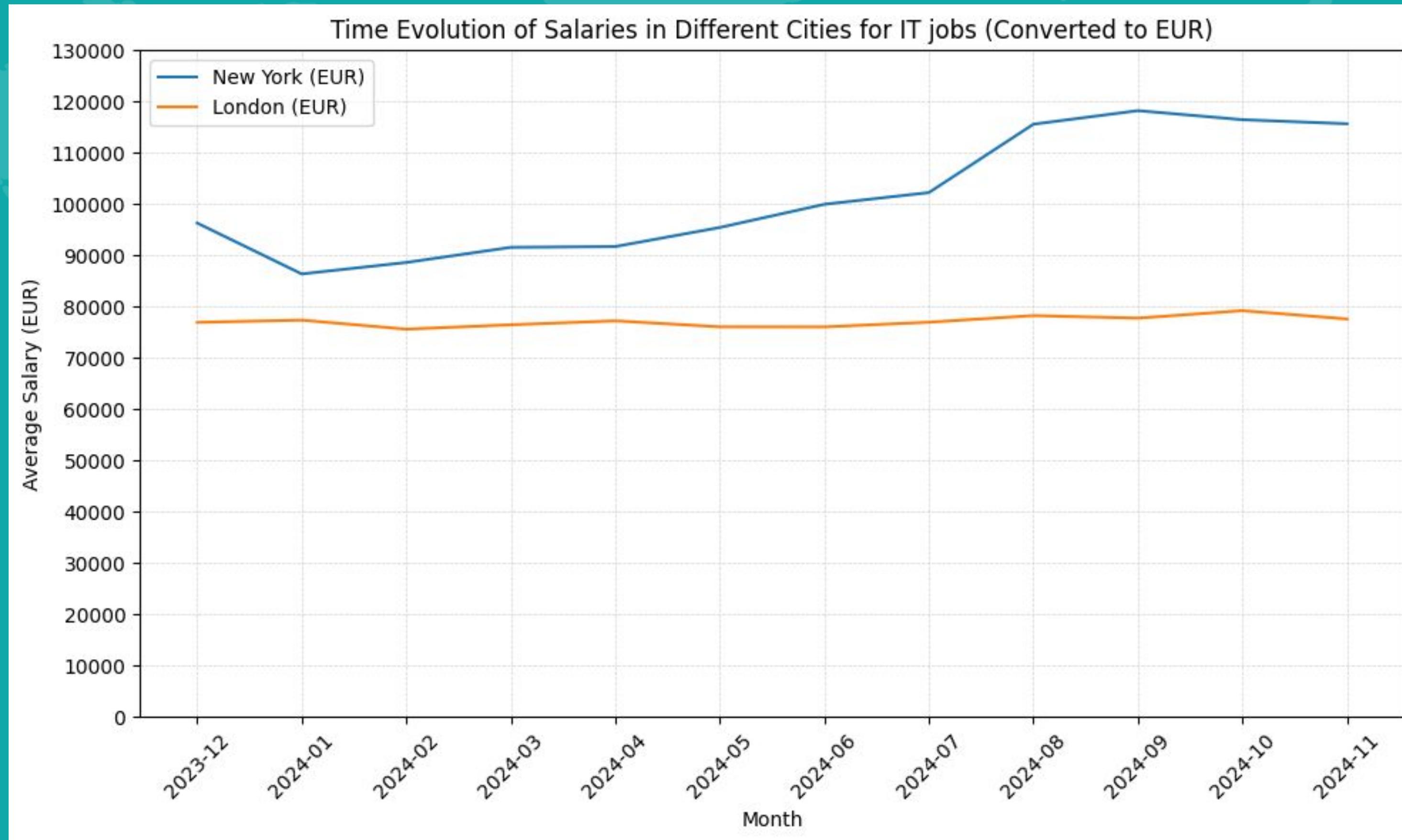


Data Visualization

- 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?

