End-to-End Job Data Analysis, Recommendation, and Resume Customization System

Data Scraping: Rowan

DevOps: Rowan, David, Sam

Gen Al: Kevin

ETL & Reporting: Jasper

Goal:

To develop an end-to-end automated system that empowers job seekers and analysts by leveraging advanced data processing, AI, and visualization technologies.

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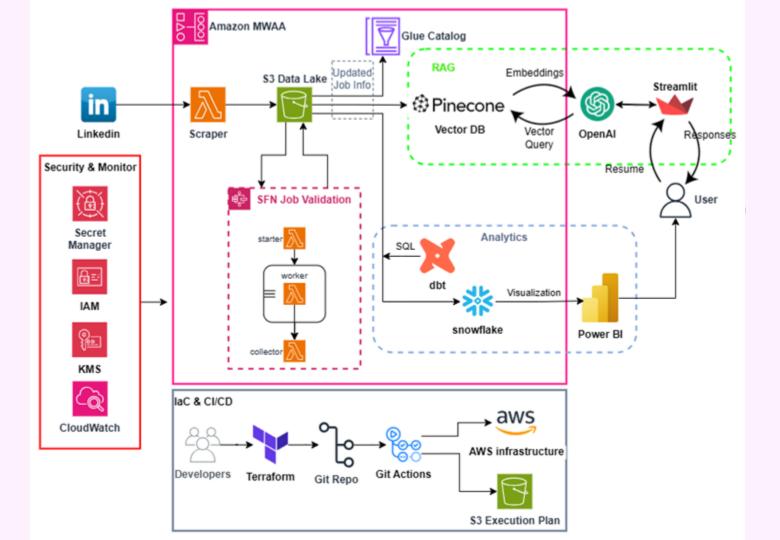
Data Lake + Gen AI + Analytics + Visualization

- Personalized Job Matching
- Customized Resumes and Cover Letters
- Job Market Analysis
- Seamless User Experience
- Interactive Visual Dashboards

Introduction and Project Overview

Pipeline

Spark on kubernetes?



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Gen Al

01

Data Scraping

LinkedIn Job Scraper Architecture

Technology Stack

- Python-based LinkedIn API
- Direct access to LinkedIn's Voyager endpoints
- AWS Lambda & S3 for deployment

Key Features

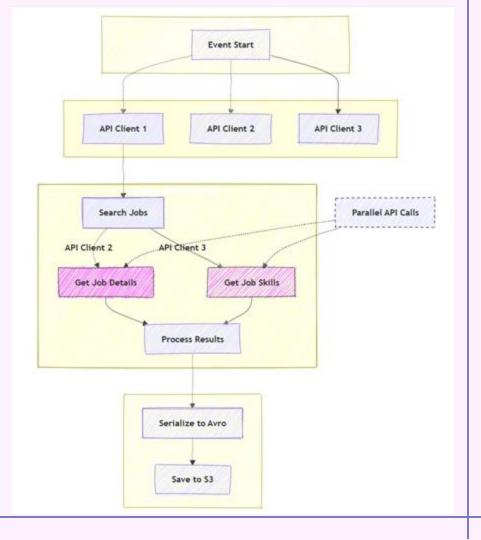
- Asynchronous processing using 3 accounts
- Rate limit handling & Data deduplication

Data Processing

- Job details and skills data extraction
- Storage in Avro format

Maintenance Note

- Cookie refresh required every 3 months
- Stored securely in S3



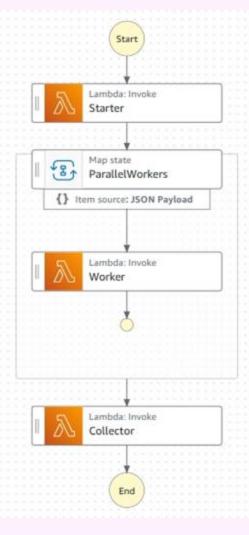
LinkedIn Job Validation

Architecture Overview

- Distributed Processing Architecture with AWS Step Functions
- Dynamic Worker Allocation
- Parallel Job Status Validation
- Automated Invalid Job Filtering

Key Benefits of Step Functions

- Native AWS Service Integration
- Automatic State Management
- Effortless to Build



02

Data Lake,
Processing &
Pipeline
Orchestration

Storage

Storage Architecture

- S3 as scalable data lake
- Snowflake for data warehousing

Why Avro Format?

- Strong support for fields evolution
- Store schema with data
- Built-in schema validation
- Efficient binary compression

```
AVRO SCHEMA = {
    "type": "record",
    "name": "JobPosting",
    "fields":
        {"name": "job_id", "type": "string"},
       {"name": "title", "type": "string"},
       {"name": "company", "type": "string"},
       {"name": "location", "type": "string"},
       {"name": "employment type", "type": "string"},
       {"name": "seniority level", "type": "string"},
        {"name": "industries", "type": {"type": "array", "items": "string"}},
       {"name": "job_functions", "type": {"type": "array", "items": "string"}},
        {"name": "workplace_type", "type": "string"},
        {"name": "description", "type": "string"},
        {"name": "skills", "type": {"type": "array", "items": "string"}},
        {"name": "job_url", "type": "string"},
       {"name": "reposted", "type": "boolean"},
       {"name": "posted time", "type": "long"},
       {"name": "expire time", "type": "long"},
        ("name": "apply url", "type": "string")
```

Glue Catalog

Table Configuration

- External table for Avro data
- Partitioned by year/month/day/keyword
- Managed through Terraform

Crawler Features

- Automated metadata updates
- S3 encryption enabled
- Schema change logging

job_id	title ♥	company	employment_type ▼	seniority_level '
4116368744	Data Engineer - Video Architecture -LiveNet (Live Streaming Networks)	TikTok	Full-time	Entry level
4117011223	Data Engineer	Sharp & Carter	Full-time	Entry level
4116917812	Systems Engineer	Thales	Full-time	
4030452252	Principal Security Engineer, AWS Security	Amazon Web Services (AWS)	Full-time	Director
4101864395	Azure DevOps Engineer	Ampstek	Contract	Mid-Senior level
4089387407	Senior Data Engineer	EPAM Systems	Full-time	Mid-Senior level

Data Processing

AWS Lambda

- LinkedIn data extraction via Python
- Data storage in Avro format
- 15-minute timeout sufficient
- Automatic scaling



Benefits

Serverless · Cost-efficient · Scalable

Pipeline Orchestration

Orchestration – Automation of daily tasks

• pipeline is triggered everyday to scrap jobs posted in the last 24 hours

Airflow

- Comprehensive Workflow Management & Flexibility
- Robust Monitoring & Error Handling
- Integration with wide range of services

Step Function

- AWS native service for workflow orchestration
- Limited to AWS services primarily, less flexible for non-AWS tasks





Why Include Step Functions Within Airflow?

Problem statement: During the job validation step, we need to split the workload and invoke a dynamic number of Lambdas to call the LinkedIn API from different IPs, avoiding being blocked.

Challenges with Airflow:

- **Static Nature of DAGs**: Airflow DAGs are predefined and static. Adjusting workflows dynamically is possible but requires workarounds that are not elegant.
- Limitations of LambdaInvokeFunctionOperator: Not Deferrable
 - Synchronous Invocation: Blocks Airflow workers while waiting (up to 15 minutes until Lambda timeout).
 - o **Asynchronous Invocation**: Cannot track Lambda execution status, leaving success/failure unknown.

Advantages of AWS Step Functions:

- Dynamic Workflow Support:
 - The first Lambda outputs the number of subsequent Lambdas required.
- Step Functions orchestrate and dynamically invoke these Lambdas, managing execution statuses seamlessly.
- Deferrable in Airflow:
 - Using the StepFunctionStartExecutionOperator in Airflow avoids occupying workers while waiting for Step Functions to complete. This enhances resource efficiency and scalability.

Airflow DAGs Demonstration in MWAA UI

<u>1d419daf-46dd-4d52-9e7d-c8fe4633b6f1.c7.ap-southeast-2.airflow.amazonaws.com</u>

Security:

MWAA

- **Login**: Users authenticate to MWAA through AWS credentials, with optional Multi-Factor Authentication (MFA) for enhanced security.
- User Roles: Permissions and roles within Airflow are managed using AWS IAM policies.
- **Authorization**: Access to AWS services is managed through the MWAA execution role, which defines the actions Airflow is allowed to perform on AWS resources.

Within airflow

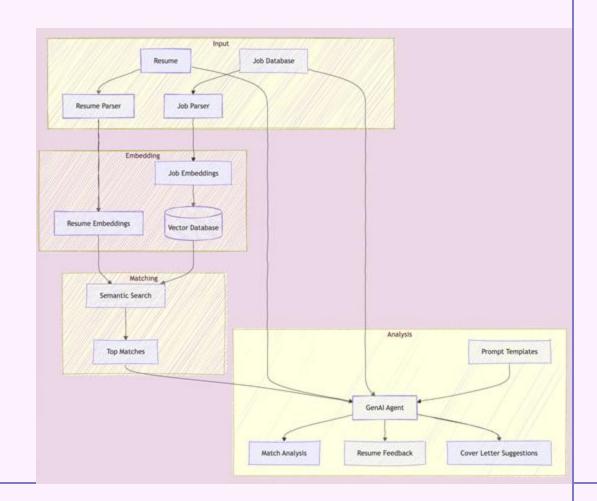
- Data Encryption in Airflow:
 - Configured AIRFLOW__CORE__FERNET_KEY to encrypt sensitive data such as variables, connections, and XComs within Airflow, ensuring confidentiality.
- AWS Secrets Manager as Secret Backend:
 - Integrated AWS Secrets Manager as the secret backend for Airflow to securely store and manage sensitive information like secrets, API keys, and connection details.
 - o Enables Airflow to securely access third-party services, such as:
 - Gmail: For sending notification emails,
 - OpenAl and Pinecone: For injecting new jobs and removing invalid jobs from the vector database,
 - Snowflake: For extracting data and loading it into the data warehouse.

03

Gen Al

Al powered job hunting

- Key steps
 - Retrieval
 - Embedding
 - Generation
 - Analysis
 - Evaluation
- Key features
 - Intelligent job matching
 - Resume analysis and feedback
 - Cover letter generation & support





- Highlight his experience with designing and implementing end-toend data solutions.
- Discuss specific projects or achievements that demonstrate his skills and expertise.
- Express his interest in the specific company and role, and how he can contribute to their success.

Top Matching Jobs

Data Engineer at Accenture Australia (Match: 100.00%)

Data Engineer (AWS + Databricks) at CareCone Group (Match: 75.00%)

Data Engineer at Accenture Australia (Match: 50.00%)

Data Engineer at Sense Recruitment (Match: 25.00%)

Analysis complete! Feel free to ask questions about the matches.

Jobs AI - Your Career Assistant



1. Overview of Candidate's Background and Job Match Potential:

Zhaoyu Guo is a data professional with over two years of experience in designing and implementing end-to-end data solutions. His expertise lies in scalable pipeline development, workflow optimization, interactive dashboard creation, and process automation using Python and cloud technologies. He has a strong background in programming languages, database and data warehouse management, big data, cloud and infrastructure, and tools and frameworks. His professional skills include communication, requirements gathering, teamwork, problem-solving, and stakeholder management. Given his skills and experience, he is a strong candidate for the Data Engineer positions.

- 2. Key Skills and Experience Relevant to the Matched Positions:
- · Proficiency in Python and SQL, which are required for data manipulation and pipeline development.
- . Experience in designing, developing, and maintaining scalable and efficient data pipelines.
- · Knowledge of cloud services such as AWS, GCP, and Azure.
- · Familiarity with data warehousing solutions like Snowflake and BigQuery.
- · Experience with big data technologies like Spark and Kafka.
- · Strong problem-solving skills and the ability to troubleshoot complex data issues.
- 3. Specific Recommendations for Top 3 Most Promising Matches:
- Data Engineer at Accenture: This role requires proficiency in Python, SQL, and cloud services, all of which Zhaoyu has. His

Ask a follow-up question about the jobs or resume advice

04

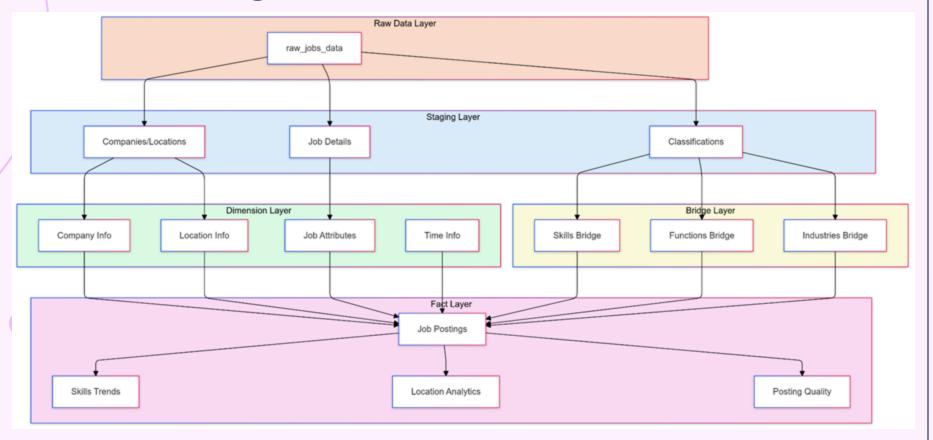
Data Transformation & Reporting

ELT pipeline



- Snowflake handles data storage and compute:
 - Stores raw data from sources (S3, JSON, AVRO)
 - Manages data marketplace access
 - Provides scalable compute for transformations
- DBT manages transformation layer:
 - Handles SQL transformations
 - Orchestrates data flows
 - Implements testing
 - Manages version control
 - Documents data lineage

Data Transform using DBT



Kimball's dimensional modeling

Fact Tables:

- 1. Main fact_job_postings:
 - One row per unique job posting ID
 - Links to all dimensions via surrogate keys
- 2. Analytical fact tables:
 - skill trends
 - location_analytics
 - posting_quality

Dimension Design:

- Company: Basic dimension with company metrics
- Location: Hierarchy (Country -> Region -> State -> City) + workplace type
- Employment: Type and seniority level standardization
- Time: Date dimension with year, month, quarter
- Bridge tables for many-to-many relationships:
 - job_skills
 - job_functions
 - job_industries

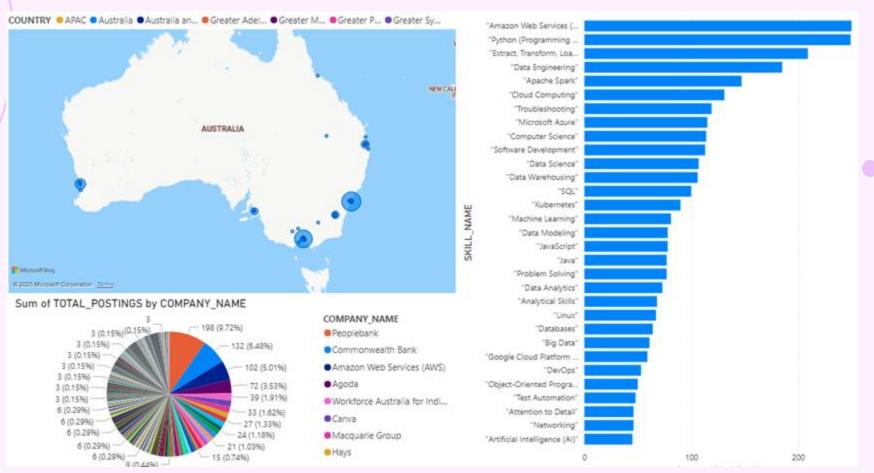
Visualization using Power BI

skill demand trends, hiring patterns by industry, location-based job insights, ...

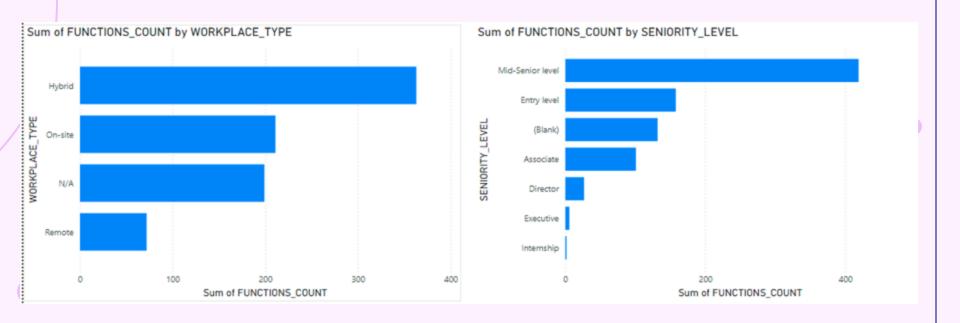
Power BI is preferred over Tableau

- Seamless integration with Excel, Azure and Microsoft 365 suite
- Direct connection to Excel / SharePoint data
- Data modelling capability and strong DAX language for complex calculations
- Extensive Power Query capabilities for data transformation
- Version control support (on Power BI Server)

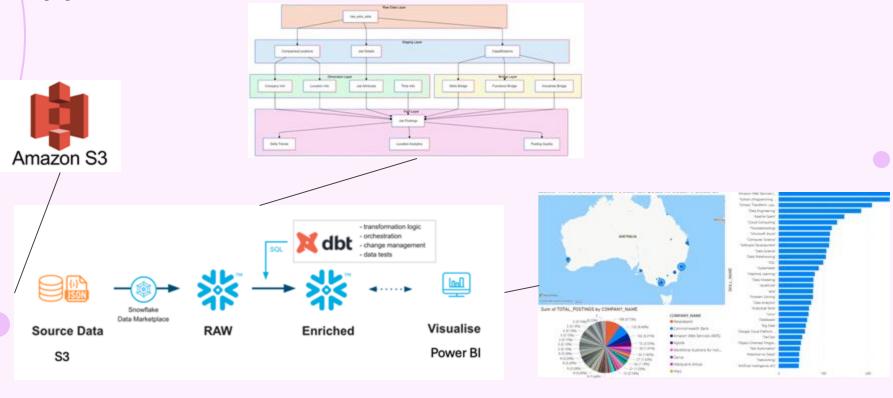
Visualization using Power BI



Visualization using Power BI



ELT pipeline



05

DevOps & Project Management

DevOps

- Terraform for Infrastructure as Code
- GitHub for version control and GitHub Actions for CI/CD
- Airflow for workflow orchestration and monitoring
- CloudWatch for infrastructure and application monitoring



Terraform

- Infrastructure as Code (IaC) automation
- Multi-cloud support
- Version-controlled infrastructure
- Consistent environments
- Easy to code & Reduced human error

Advantages over CloudFormation

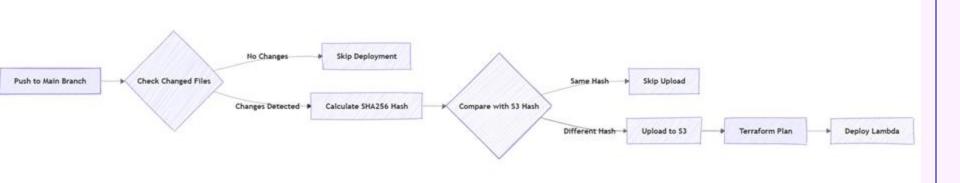
- Provider-agnostic (not AWS-only)
- HCL syntax (more readable than JSON/YAML)
- State management capabilities
- Faster deployment execution
- Better dependency handling

```
variable "instance type" {
  type = string
instance type = "t2.micro"
resource "aws_instance" "example" {
  instance_type = var.instance_type
```

CI/CD: GitHub Action

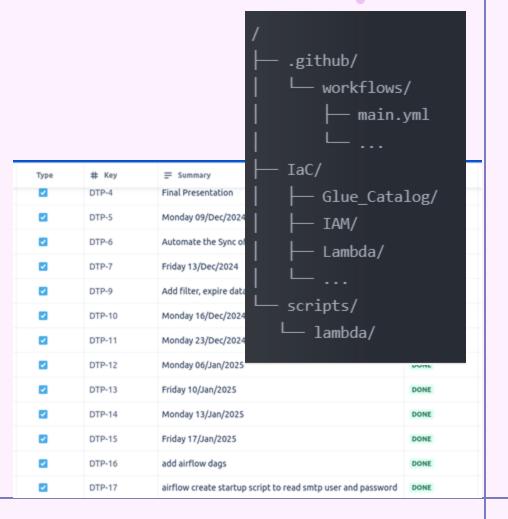
Lambda Function Depolyment





Code repo and Project Management

- Github for code repo and CI/CD
 - Scraper code and IaC
 - Data Modeling (dbt models)
 - RAG
 - Orchestration (Airflow dags)
- JIRA for Planning and Task Management
- Confluence for documentation, knowledge management, and collaboration



Thanks!

