

1. Project Overview

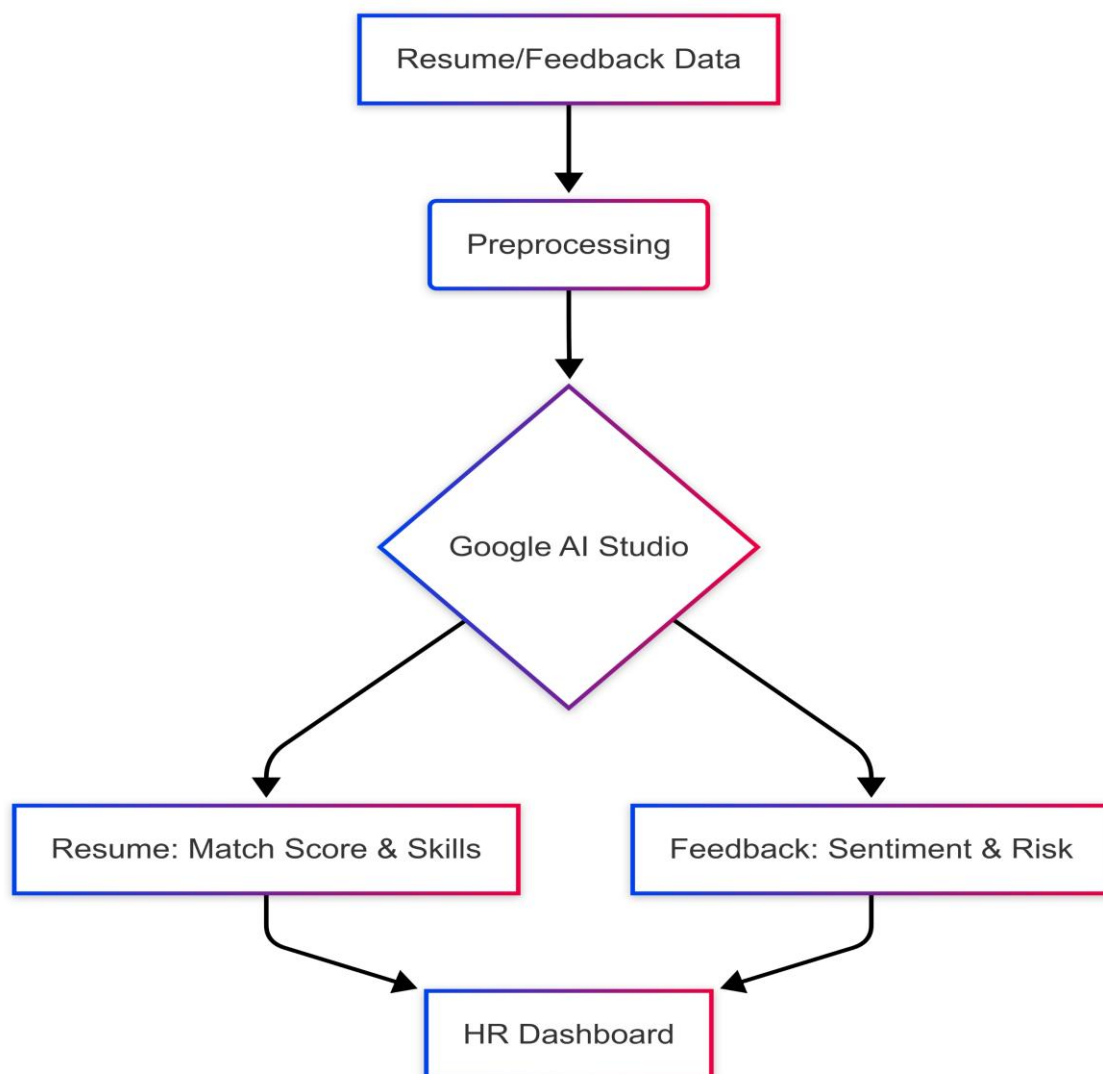
Objective: Automate HR processes using AI for resume screening and employee sentiment analysis.

Tools:

- **Google AI Studio** (Gemini Pro) for LLM integration.
 - **Python** for data processing and API development.
 - **FastAPI** for local deployment (bonus).
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2. Technical Approach

2.1 Workflow Diagram



2.2 Data Pipelines

- **Resume Screening:**
 1. **PDF Parsing:** Extract text using PyMuPDF.
 2. **Text Cleaning:** Remove irrelevant characters.
 3. **Prompt Engineering:** Structured input for LLM.
 4. **LLM Processing:** Match skills/experience with JD.
 - **Sentiment Analysis:**
 1. **Text Normalization:** Lowercase, remove stopwords.
 2. **Prompt Design:** Classify sentiment and suggest strategies.
 3. **LLM Processing:** Generate JSON output.
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3. Model Selection & Integration

- **LLM Choice:** Gemini Pro (Google AI Studio) for its 1M token context window.
 - **Integration:**
 - Use Google AI Studio's API for prompt testing.
 - Simulate responses locally for rapid prototyping.
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4. Prompt Engineering

4.1 Resume Screening Prompt

python

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```
def build_resume_prompt(resume_text, job_desc):
```

```
    return f"""
```

```
    Role: Senior HR Analyst
```

```
    Task: Analyze the resume against the job description.
```

```
    Output Format: JSON with keys [matched_skills, missing_skills, match_score].
```

```
    Resume:
```

```
    {resume_text[:10000]} # Truncate for token limits
```

```
    Job Description:
```

```
{job_desc[:5000]}
```

```
"""
```

4.2 Sentiment Analysis Prompt

python

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```
def build_sentiment_prompt(feedback):
```

```
    return f"""
```

```
    Role: Employee Engagement Expert
```

```
    Task: Analyze feedback for attrition risk and suggest improvements.
```

```
    Output Format: JSON with keys [sentiment, risk_score, strategies].
```

```
Feedback:
```

```
{feedback}
```

```
"""
```

5. Code Implementation

5.1 Resume Parser (resume_screening/parser.py)

python

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```
import pdfplumber
```

```
def parse_pdf(file_path):
```

```
    with pdfplumber.open(file_path) as pdf:
```

```
        return " ".join([page.extract_text() for page in pdf.pages])
```

5.2 FastAPI Endpoint (api/main.py)

python

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```
from fastapi import FastAPI
```

```
from pydantic import BaseModel
```

```
app = FastAPI()
```

```
class ResumeRequest(BaseModel):
    resume_text: str
    jd_text: str

@app.post("/screen")
async def screen_resume(request: ResumeRequest):
    # Simulated Gemini response
    return {
        "matched_skills": ["Python", "AWS"],
        "match_score": 85
    }
```

6. Deployment Steps

1. Google AI Studio Setup:

- Create prompts and test with sample data.
- Capture screenshots of inputs/outputs.

2. Local API Testing:

bash

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uvicorn api.main:app --reload

- Test via <http://localhost:8000/docs>.

3. Cloud Deployment (Bonus):

- Containerize with Docker.
 - Deploy to Google Cloud Run.
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7. Technical Report Content

Sections:

1. **Introduction:** Problem statement and business impact.
2. **Methodology:** Workflow, data pipelines, model selection.
3. **Implementation:** Code snippets, prompt design.

4. **Results:** Screenshots from Google AI Studio.
5. **Challenges:** PDF parsing errors, prompt tuning.
6. **Conclusion:** Scalability and future work.