

# Job AI Agent — Full Workflow & Techniques (Workbook)

A practical, production-ready workbook describing nodes, techniques, data schemas, prompts, and example code snippets to build a job-search + auto-apply AI agent. Use this as a single-source design and implementation guide.

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## 1. Goals & Scope

**Primary goal:** Build an autonomous/semi-autonomous agent that: finds job postings matching user preferences, extracts and normalizes job data, ranks jobs, personalizes resume sections & cover letters, fills application forms (where allowed), logs applications, and surfaces results via a dashboard with human-in-the-loop checkpoints.

**Constraints & safety:** Respect terms-of-service for job portals. Prefer official APIs / third-party aggregators (SerpAPI, JSearch, Indeed API). Use Selenium/Playwright only for internal/personal accounts and rate-limit actions. Never bypass captchas.

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## 2. High-level Architecture

```
[User Profile & Resume JSON] --> [Job Search Node] --> [Job Fetch Node] -->
[Extractor Node]
    --> [Matcher & Ranker Node] --> [Resume Personalizer Node] --> [Cover
Letter Node]
    --> [Form Filler Node] --> [Application Tracker DB] --> [Dashboard]
```

Human-in-loop checkpoints: Review Matches -> Confirm Apply -> Solve Captcha

Supporting infra: FastAPI backend, LangGraph / LangChain orchestration, Vector DB (Milvus/Pinecone/Weaviate), Redis for queues, Postgres for audit logs, S3 for files.

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## 3. Nodes (Detailed)

### 3.1 Job Search Node

**Purpose:** Discover job listings given keywords, location, filters.

**Techniques:** - Use SerpAPI (LinkedIn/Indeed engines) or official Indeed API - Google Custom Search as fallback - Search query templates with synonyms expansion - Rate limiting and pagination

**Inputs:** user\_keywords, location, remote\_flag, experience\_level **Outputs:** list of job-result objects: {title, company, short\_snippet, job\_url, source, posted\_date}

**Notes:** store request metadata (query, timestamp) for auditing.

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### 3.2 Job Fetch / Page Loader Node

**Purpose:** Fetch full job page HTML or API JSON for a job\_url.

**Techniques:** - Prefer API endpoints (Greenhouse, Lever) - If not available, use Playwright (headful) or Selenium to render JS pages - Respect robots.txt, throttle requests

**Outputs:** raw\_html, final\_url, http\_status, screenshots(optional)

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### 3.3 Extractor Node (Job Description Parser)

**Purpose:** Convert raw\_html -> structured JD fields

**Techniques:** - Use `unstructured` to extract main text - LLM-based extraction (prompt to extract sections) - Regexp + heuristic fallback - Named-entity recognition (spaCy) for skills, locations

**Output schema:**

```
{
  "job_id": "",
  "title": "",
  "company": "",
  "location": "",
  "employment_type": "",
  "salary": "",
  "posted_date": "",
  "description": "",
  "responsibilities": [],
  "required_skills": [],
  "nice_to_have": [],
  "application_url": ""
}
```

**LLM prompt pattern (short):**

Extract and return JSON with fields: title, company, location, posted\_date, responsibilities (list), required\_skills (list), nice\_to\_have (list), description (cleaned text) from the following job text: <JOB\_TEXT>

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### 3.4 Matcher & Ranker Node

**Purpose:** Score and rank jobs for the user.

**Techniques:** - Embedding similarity (OpenAI/HF embeddings) between job description and user's resume/profile - Rule boosts: must-have skills, location match, remote preference - Heuristic scoring: recency, seniority match - Learning-to-rank (optional) using historical user clicks/applies

**Output:** Ranked job list with `score`, `matching_skills`, `explainability` (why ranked)

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### 3.5 Resume Personalizer Node

**Purpose:** Generate tailored resume sections or full resume variants for a job

**Techniques:** - Input: base\_resume JSON, extracted JD - Use LLM to rewrite bullets, reorder skills, generate tailored summary, and select projects - Ensure fidelity: do not invent facts — rely on user-provided achievements & projects - Output schema (sections): `summary`, `skills`, `experience_bullets`, `projects`, `achievements`, `keywords`

**Sample prompt fragment:**

```
Given this resume JSON and the job requirements, produce JSON containing:
personalized_summary (2-3 sentences), top_8_skills (ordered), 5 experience
bullets rewritten to match JD, 2 highlighted projects. Keep facts only from
resume.
```

**Safety:** Add a verification pass: every generated bullet must cite origin field (e.g., project id or experience id).

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### 3.6 Cover Letter Generator Node

**Purpose:** Generate a personalized cover letter using JD + resume personalization outputs

**Techniques:** - LLM generation with structured template and tone settings - Include 2-3 bullets aligning achievements to job needs - Produce a short paragraph for why candidate fits & interest

**Output:** `cover_letter_text`, `cover_letter_html`, `highlights` (bullets used)

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### 3.7 Form Filler / Auto-Apply Node

**Purpose:** Automate application submission via APIs or browser automation

**Techniques & strategy:** 1. **API-first:** Check if the job source exposes a submit API (Greenhouse/Lever). If yes, encode payload and POST. 2. **Selenium/Playwright:** If no API, spin a browser session (headful), locate fields (LLM-assisted), fill, upload resume, and submit. - HTML field detection approach: - Use

heuristics (name/id/labels) + LLM to map fields -> profile attributes - Example mapping JSON: `{ "email": "xpath://...", "resume": "xpath://..." }` 3. **Human-in-loop:** For captchas or multi-step forms, pause & notify user with screenshot and resume after manual solve.

**File uploads:** Use `input[type=file].send_keys(local_path)` in Selenium.

**Safety & throttling:** 1 apply/sec per account is too fast — use 1 apply / 1–5 minutes depending on source.

**Audit log:** store request/response, status, timestamp, screenshot.

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### 3.8 Application Tracker & DB

**Purpose:** Record every job found and applied to, track status, follow-ups, interview pipeline.

**Schema (Postgres):** - jobs (job\_id, user\_id, title, company, url, posted\_date, fetched\_at) - applications (application\_id, job\_id, user\_id, resume\_variant\_id, cover\_letter\_id, status, applied\_at, response) - tasks (captcha\_pending, manual\_review)

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### 3.9 Dashboard / UX Node

**Purpose:** Present matches, let users review, edit personalized sections, confirm apply, show application history.

**Tech stack:** React + Tailwind (or create-agent-chat-app UI) + FastAPI backend

**Features:** - Job feed with inline highlights (matching\_skills) - Edit preview of personalized resume bullets & cover letter - One-click apply (with confirm modal) - Notifications for captcha/manual steps

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## 4. Data Schemas (JSON) — Base Resume Example

```
{
  "user_id": "u_123",
  "name": "Jugal Lachhwani",
  "email": "jugal@example.com",
  "phone": "",
  "summary": "...",
  "skills": ["Python", "TensorFlow", "FastAPI", "LangChain", "Selenium"],
  "experiences": [
    {
      "id": "e1",
      "title": "ML Engineer",
      "company": "X",
      "start": "2024-01",
      "end": "2024-12",
      "bullets": [
        "Built...",
        "Deployed..."
      ]
    }
  ],
  "projects": [
    {
      "id": "p1",
      "title": "Email Agent",
      "desc": "...",
      "achievements":
```

```
[{"reduced time by 40%"}]
```

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## 5. Prompt Templates (Practical)

### 5.1 Extractor prompt (clean JSON):

You are a JSON extractor. Parse the job text and return only JSON with fields: title, company, location, posted\_date, responsibilities(list), required\_skills(list), nice\_to\_have(list), description. Do not add extra commentary.  
<JOB\_TEXT>

### 5.2 Resume Personalizer prompt:

Input: base\_resume\_json, job\_json  
Task: Produce JSON: {personalized\_summary, top\_skills, experience\_bullets, projects\_highlighted}  
Rules: Use only facts from base\_resume\_json. Map each generated bullet to a source id in the resume. Keep bullets <= 20 words.

### 5.3 Field mapping prompt (for Form Filler):

Given the following HTML snippet, identify which input fields correspond to name, email, phone, resume upload, linkedin, github. Reply with JSON mapping {field\_label: css\_selector\_or\_xpath, type}. Only output JSON.  
<HTML\_SNIPPET>

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## 6. Example LangGraph / LangChain Node Map (Pseudo)

- job\_search\_node (python) -> paginated
  - fetch\_node (async) -> concurrency limit
  - extractor\_node (llm + unstructured)
  - embeddings\_node (generate embeddings for JD & resume)
  - ranker\_node (scoring function)
  - personalize\_node (LLM)
  - apply\_node (api or selenium)
  - logger\_node (db)
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## 7. Code Snippets (Short)

### 7.1 SerpAPI search (python)

```
import requests
params = { 'engine': 'linkedin_jobs', 'q': 'machine learning engineer',
'location': 'Bangalore', 'api_key': SERPAPI_KEY }
r = requests.get('https://serpapi.com/search', params=params)
jobs = r.json()['jobs_results']
```

### 7.2 Playwright fetch (python)

```
from playwright.sync_api import sync_playwright
with sync_playwright() as p:
    browser = p.chromium.launch(headless=False)
    page = browser.new_page()
    page.goto(job_url)
    html = page.content()
```

### 7.3 Selenium upload

```
upload = driver.find_element(By.XPATH, "//input[@type='file']")
upload.send_keys(r"C:\path\resume.pdf")
```

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## 8. Testing & Metrics

**E2E tests:** mock job feeds, verify extractor outputs, verify mapping accuracy, test apply flow in staging accounts.

**KPIs:** - Match precision @ top5 - Apply success rate - Time per application (human-in-loop) - False-positive resumes generated (factual mismatch rate)

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## 9. Deployment & Scaling

- Containerize nodes (Docker). Use kubernetes for scaling fetch/extractor.
- Use Redis queue for concurrency control.
- Use vector DB for embeddings; keep cold-start index warm for common skills.
- Rotate API keys and respect rate limits.

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## 10. Legal & Ethical Checklist

- Respect TOS of sources. Prefer API partners.

- Never circumvent CAPTCHA or bot protections.
  - Transparently inform users when human intervention is required.
  - Provide an audit trail of generated text and sources.
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## 11. Next Steps & Templates

- ☒ Choose source (SerpAPI vs official API)
  - ☒ Build minimal proof-of-concept: Job Search -> Extractor -> Ranker -> Personalize resume -> Preview UI
  - ☒ Add apply automation for one source (e.g., Greenhouse) in API mode
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## 12. Files to generate (when implementing)

- resume\_schema.json
  - job\_schema.json
  - langgraph\_workflow.json
  - fastapi\_app/ endpoints
  - frontend/ React components: JobFeed, ResumeEditor, ApplyModal
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## 13. Quick Checklist for Implementation

1. Get SerpAPI key or Indeed API credentials
  2. Build base resume JSON
  3. Implement Job Search node (save to DB)
  4. Implement Job Fetch & Extractor
  5. Implement Embeddings and Ranker
  6. Implement Resume Personalizer & Cover Letter generator
  7. Implement Apply node (API first; Selenium fallback)
  8. Build Dashboard & human-in-loop flows
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*End of workbook.*

If you want, I can now: - Generate the LangGraph workflow JSON for this workbook - Scaffold the FastAPI + Selenium repo with code for 3 nodes - Create the React dashboard skeleton

Which one should I generate next?