

## Proposal: Internal AI Co-Pilot for Enterprise Data Access

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### Overview: What We're Building

We aim to build a secure, on-premise AI Co-Pilot that allows internal users to ask natural-language questions over company systems (Oracle DB, Excel, reports) and receive answers, insights, and summaries instantly. This will:

- Simplify data access (e.g., “What was Vendor A’s total spend in Q3?”)
- Generate reports (e.g., Excel sheets of high-variance SKUs)
- Validate data (e.g., flag GST mismatches)
- Summarize trends (e.g., procurement highlights)

All data and processing will remain securely inside our infrastructure.

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### Technology Stack

- **Database:** Oracle DB with Python access via `cx_Oracle`
  - **Vector Store:** Chroma, Weaviate, FAISS
  - **Embedding Models:** SentenceTransformers (MiniLM, BGE, etc.)
  - **LLMs:** Mistral, LLaMA 2, Falcon (quantized, open-source)
  - **Frameworks:** LangChain, LlamaIndex
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### Capabilities

- Query structured data from Oracle via SQL (manual or generated by LLM)
  - Perform semantic search over documents and past records
  - Generate Excel reports and summaries via Python
  - Answer contextual questions using Retrieval-Augmented Generation (RAG)
  - Keep everything local for maximum privacy and control
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### RAG Workflow Explained

#### *What We Actually Do:*

1. **Prepare Internal Data**
  - Pull data from Oracle using Python
  - Convert tables/reports to readable text chunks
2. **Embed & Index**
  - Use embedding models to convert text chunks into vectors

- Store vectors in Chroma/Weaviate
- 3. **User Query → Semantic Search**
  - User asks a question (“top 10 SKUs with variance”)
  - System retrieves relevant text chunks
- 4. **Prompt the LLM with Retrieved Chunks**
  - Pass both user question + retrieved context to LLM
  - Get accurate, grounded answer

#### *Why This Works*

- No fine-tuning needed
- Minimal hallucination (grounded answers)
- Secure and efficient

## Real-Time Oracle DB Integration

### *Can We Pull Data Live from Oracle?*

Using Python libraries:

- cx\_Oracle for direct SQL
- SQLAlchemy for ORM
- Select AI (if available) for natural language to SQL

### *Security Best Practices*

- Use service accounts with read-only access
- Encrypt connections (SSL/TLS)
- Store credentials securely
- Audit logs & IP firewalls

### *Hybrid Approach*

- Static embeddings for documents, reports
- Live SQL for real-time metrics

This ensures speed + freshness.

## Sample Query Flow

User: “What were total sales in Region A last week?”

1. LLM generates:

```
SELECT SUM(sales_amount) FROM sales_data WHERE region='A' AND sale_date
BETWEEN SYSDATE-7 AND SYSDATE;
```

2. Python executes, fetches result

3. Result passed back to LLM:

“Total sales in Region A last week were ₹1.25 million.”