

Multi-Agent Insurance System

Course Name: Agentic AI

Institution Name: Medicaps University – Datagami Skill Based Course

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Problem Statement & Objectives

1. Problem Statement

The insurance industry deals with highly complex policy documents, regulatory clauses, and compliance requirements. Manual analysis of policies is:

- Time-consuming
- Error-prone
- Difficult to scale
- Susceptible to inconsistency

Traditional AI systems struggle with **context overflow**, hallucinations, and lack of traceability when processing large insurance documents.

There is a need for a **collaborative, specialized multi-agent system** that can automate policy research and structured report generation while maintaining accuracy, compliance, and data security.

2. Project Objectives

The main objectives of this project are:

- Design and implement a **Multi-Agent Insurance System**
- Separate responsibilities between specialized agents
- Automate insurance policy research and reporting
- Ensure structured and traceable outputs
- Implement secure local LLM processing
- Maintain workflow state management
- Reduce hallucinations through structured context passing
- Provide a scalable and modular architecture

3. Scope of the Project

The project scope includes:

- Development of a **two-agent collaborative system**
- Integration of:
 - LLM for policy analysis
 - External API for data retrieval
 - MongoDB for persistence
- REST API-based interaction
- Sequential task orchestration using agent framework
- Error handling, retry mechanisms, and logging
- Structured output generation for insurance policy reports

Proposed Solution

The system implements a **Collaborative Multi-Agent Architecture** where tasks are divided between:

- Researcher Agent (Fact extraction & evidence grounding)
- Writer Agent (Synthesis & structured documentation)

Orchestration is handled using **CrewAI**, ensuring sequential execution and automatic context passing.

LLM inference is handled locally using **Ollama** to ensure privacy and PII protection.

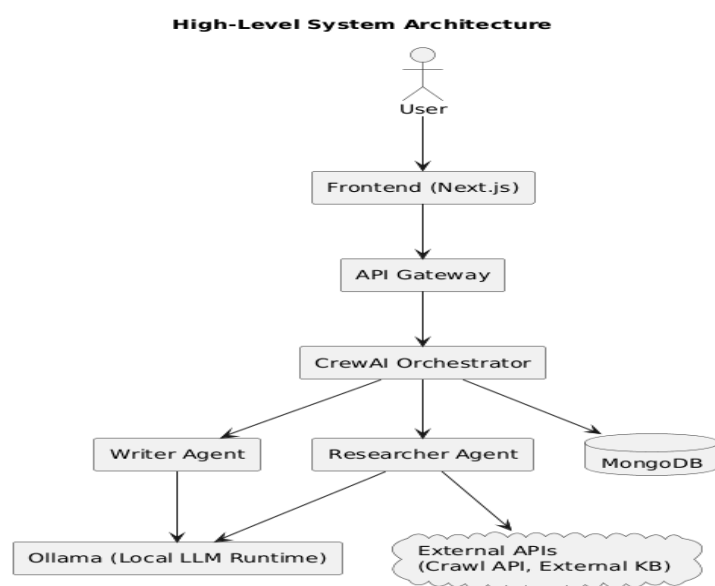
Data persistence is implemented using **MongoDB**.

1. Key Features

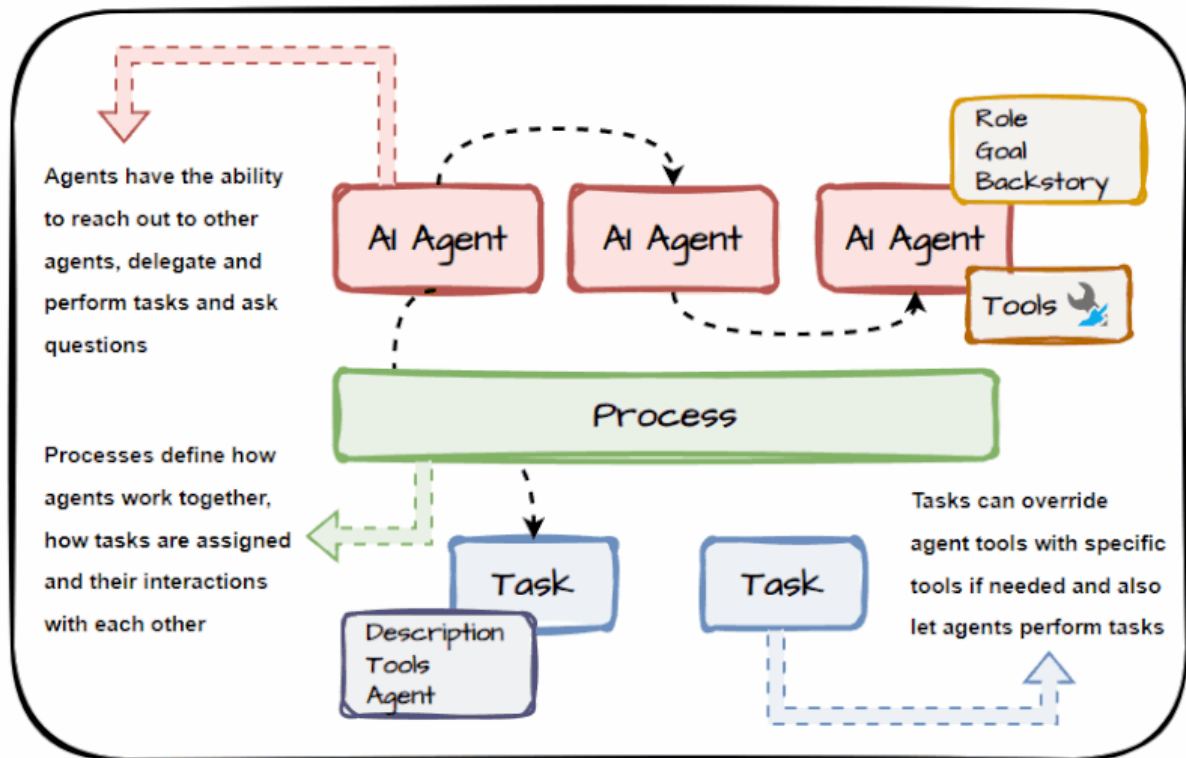
- Agent Specialization (Researcher & Writer)
- Sequential Task Orchestration
- Automatic Context Passing
- Structured JSON Evidence Generation
- RAG-based Information Retrieval
- Local LLM Processing (Privacy-first)
- Workflow State Management
- Retry with Exponential Backoff
- Circuit Breaker Pattern
- Semantic Caching
- REST API Endpoints
- Structured Report Generation

2. Overall Architecture / Workflow

High-Level Architecture



MULTI-AGENTIC SYSTEMS WITH CREW AI



System Layers

Client Layer

User submits insurance query

API Layer

Receives request and triggers workflow

Orchestration Layer

- Managed by CrewAI
- Handles task sequencing
- Transfers structured context

Agent Layer

- Researcher Agent
- Writer Agent

Data Layer

- MongoDB collections:

- workflow_states
- research_data
- final_outputs

External Services

- Ollama (LLM inference)
- Crawl API

Workflow Execution Steps

1. User submits insurance query
2. API creates workflow ID
3. Researcher Agent:
 - Queries LLM
 - Fetches external data
 - Validates findings
 - Generates structured JSON
4. Writer Agent:
 - Receives structured context
 - Synthesizes report
 - Applies formatting & compliance
5. Final output stored in database
6. Status updated to "Completed"
7. Structured response returned to user

3. Tools & Technologies Used

Layer	Technology
Agent Orchestration	CrewAI
LLM Integration	LangChain
LLM Runtime	Ollama
Database	MongoDB
Backend	FastAPI / Flask
Programming Language	Python 3.9+
External Data	Crawl API
Optional Caching	Redis
Logging	OpenTelemetry
Security	IAM Integration

Results & Output

1. Screenshots / Outputs

Deploy

Multi-Agent Insurance System

Enter your insurance query:

Give me policy for 20 year old male

Run Query

Research Agent Working...

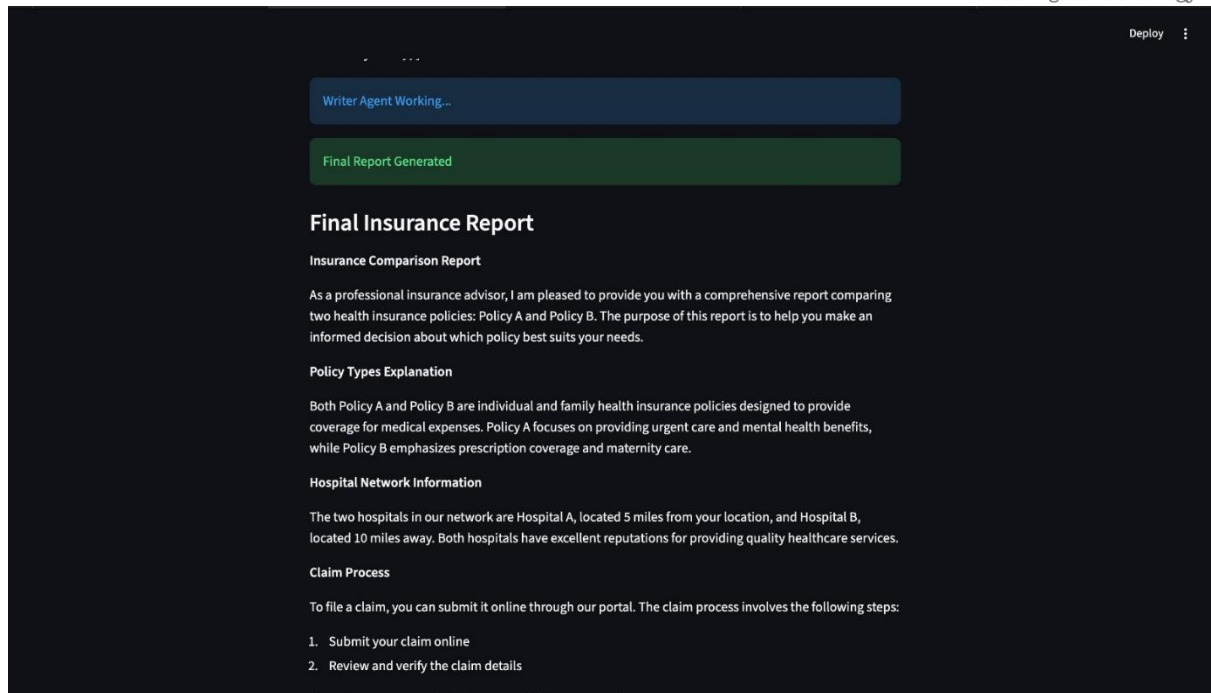
Research Completed

Research JSON Output

JSON formatting issue

```
{ "insurance_type": "health", "available_policy_types": ["individual", "family"], "network_hospitals": [{ "name": "Hospital A", "distance": 5 }, { "name": "Hospital B", "distance": 10 } ], "claim_process": { "steps": [ "submit claim online", "review and verify" ], "timeframe": "up to 30 days" }, "claim_rejection_reasons": [ "insufficient documentation", "ineligible treatment" ], "exclusions": [ "pre-existing conditions", "experimental treatments" ], "comparison_points": [{ "policy": "Policy A", "price": 100, "benefits": [ "urgent care", "mental health" ] }, { "policy": "Policy B", "price": 150, "benefits": [ "prescription coverage", "maternity care" ] } ] }
```

Waiting Agent Working...



The system successfully generates:

- Structured JSON research outputs
- Formatted insurance coverage reports
- Workflow status logs
- Stored research evidence with traceability

2. Reports / Dashboards / Models

The system produces:

- Policy Coverage Summary Reports
- Structured JSON Output Files
- Research Confidence Scores
- Workflow Logs
- Quality Metrics:
 - Completeness Score
 - Accuracy Score
 - Readability Score

MongoDB stores:

- Research evidence
- Final formatted outputs
- Workflow state transitions

3. Key Outcomes

- Successfully implemented a **collaborative multi-agent system**
- Reduced hallucination through task isolation
- Achieved structured, traceable outputs
- Ensured PII-safe processing using local LLM
- Built scalable modular architecture
- Implemented error handling & retry strategy
- Enabled state persistence and resume capability
- Demonstrated practical application of Agentic AI in Insurance

Conclusion

The Multi-Agent Insurance System demonstrates how **Agentic AI architectures** can transform complex insurance workflows.

By dividing responsibilities between:

- Researcher Agent (Fact extraction)
- Writer Agent (Synthesis & formatting)

the system significantly improves:

- Accuracy
- Traceability
- Modularity
- Scalability
- Compliance

The use of CrewAI for orchestration, Ollama for secure local LLM inference, and MongoDB for state persistence enabled us to build a production-ready architecture blueprint.

This project strengthened our understanding of:

- Multi-agent coordination
- RAG pipelines
- Workflow orchestration
- LLM integration
- State management
- Error handling strategies
- Secure AI system design

Future Scope & Enhancements

1. Add More Specialized Agents

- Compliance Validator Agent
- Fraud Detection Agent
- Quality Assurance Agent

2. Scalability Enhancements

- Horizontal scaling with microservices
- Load balancing
- Message queue (Kafka / RabbitMQ)

3. Performance Optimization

- Advanced semantic caching
- Async processing
- Database sharding

4. Monitoring & Observability

- Distributed tracing
- Centralized log aggregation

5. Security Enhancements

- OAuth2 / JWT Authentication
- Role-based Access Control
- End-to-End encryption

6. Feature Extensions

- PDF export generation
- Real-time dashboard
- Multi-language policy analysis
- Regulatory auto-updates
- Agent learning with feedback loop