

Research Agent System - Complete Flow Documentation

Version: 2.0.0 **Date:** 2025-01-09 **Status:** Production-Ready (ROI Model Architecture)

Table of Contents

1. [Overview](#)
 2. [13 ROI Model Types](#)
 3. [System Architecture](#)
 4. [WebSearchAgent](#)
 5. [DocumentAnalysisAgent](#)
 6. [API Layer](#)
 7. [Validation Pipeline](#)
 8. [Data Models](#)
 9. [Component Specifications](#)
 10. [Configuration & Setup](#)
 11. [Example Usage](#)
-

1. Overview

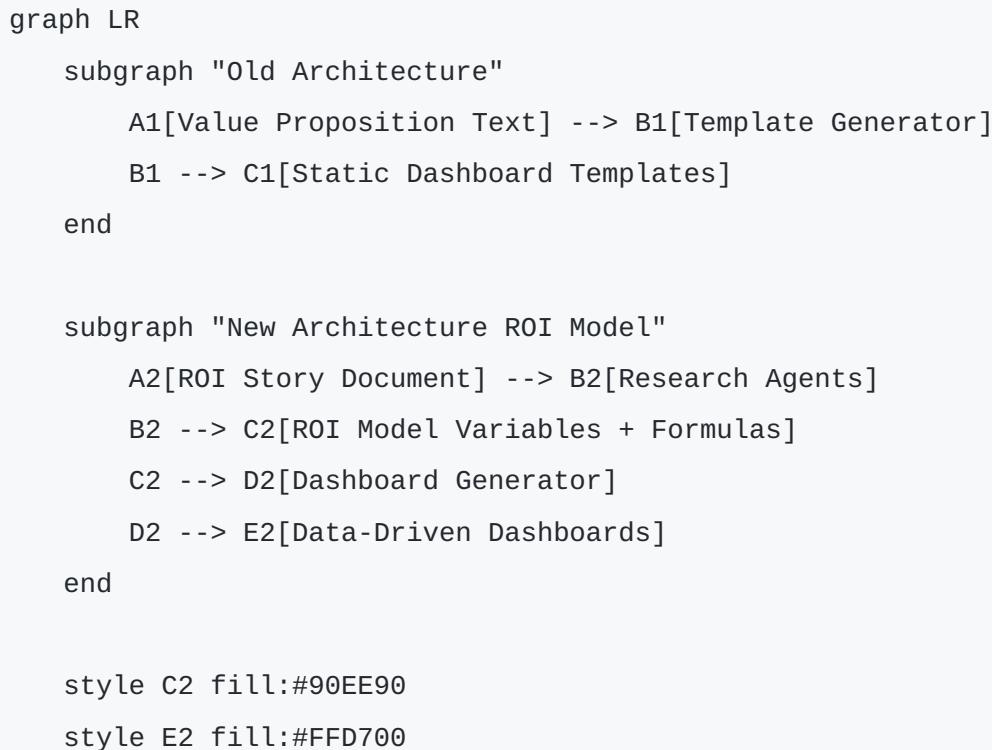
System Purpose

The Research Agent System provides two specialized AI agents for **ROI Model Data Extraction** in the Triton platform. These agents extract quantitative financial metrics, variables, and formulas that enable mathematical ROI calculations rather than qualitative value propositions.

- **WebSearchAgent:** Researches healthcare companies to extract ROI Model data (cost savings, metrics, formulas)
- **DocumentAnalysisAgent:** Analyzes ROI Story documents (PDF, DOCX, TXT) from S3 to extract ROI Model components

Both agents use Claude Sonnet 4 via AWS Bedrock with a sophisticated 4-layer validation pipeline and automatic retry logic.

Architectural Shift: Value Propositions → ROI Models



Concept	Old System	New System (ROI Model)
Primary Entity	Value Proposition (qualitative)	ROI Model (quantitative)
Input	Client documents/research	ROI Story (structured document)
Classification	None	13 ROI Model Types
Variables	Hardcoded	Dynamic variables extracted
Formulas	None	Mathematical formulas for calculations
Output	Static templates	ROI Model → Dashboard Templates

Key Features

- **Dual Research Modes:** Autonomous (15-25 searches) and Manual (5-15 searches)
 - **Multi-Tool Integration:** Google search, web scraper, S3 document reader
 - **ROI Model Extraction:** Extract variables, formulas, and assumptions for 13 ROI Model Types
 - **Robust Validation:** 4-layer validation with retry feedback
 - **Async Processing:** Background task execution with job tracking
 - **Mock Mode:** Fallback to mock data when API keys unavailable
-

2. 13 ROI Model Types

Overview

Research agents extract data to populate one of **13 distinct ROI Model Types**. Each type represents a different healthcare cost reduction or value creation strategy with specific:

- **Variables:** Baseline costs, targets, intervention parameters
- **Formulas:** ROI calculations, payback periods, savings projections
- **Data Requirements:** Claims types, member data, clinical outcomes
- **Industry Benchmarks:** Typical ranges for metrics

Type Classification Matrix

#	ROI Model Type	Category	Key Metrics	Typical ROI Range
1	Unit Price Reduction	Cost Reduction	Unit cost PMPM, contract savings %	150-300%
2	Site of Care Shift	Utilization	HOPD vs freestanding cost delta	200-400%
3	Provider Steering	Cost Reduction	Tier differential, steering rate	180-350%
4		Administrative		500-1000%

#	ROI Model Type	Category	Key Metrics	Typical ROI Range
	Payment Integrity		Overpayment recovery \$, edit hit rate	
5	Utilization Reduction	Utilization	Low-value service rate, reduction %	200-450%
6	Medical Management	Utilization	Chronic disease costs, outcomes	250-500%
7	Episode Optimization	Utilization	Episode cost, pathway adherence %	150-300%
8	OON Mitigation	Administrative	OON rate, cost differential	300-600%
9	Leakage Recapture	Utilization	Leakage %, in-system advantage	200-400%
10	Pharmacy Optimization	Cost Reduction	Pharmacy PMPM, specialty drug %	250-400%
11	Supply Chain Validation	Administrative	Device/implant cost variance	400-800%
12	Admin Error Reduction	Administrative	Duplicate payment \$, errors	600-1200%
13	Vendor Incentive ROI	Utilization	Incentive \$, steerage impact	100-250%

Research Agent Role

WebSearchAgent and DocumentAnalysisAgent extract:

- ROI Model Type Indicators:** Keywords, themes, and metrics suggesting which of 13 types applies
- Variable Values:** Baseline costs, population sizes, target metrics, intervention parameters

3. **Formula Components:** Cost calculations, savings formulas, ROI methodology
4. **Assumptions:** Industry benchmarks, default values, data source requirements
5. **Evidence:** Supporting sources, confidence scores, extracted page numbers

The extracted data feeds into the **ROI Model Builder Agent** (separate system) which constructs the complete ROI Model.

3. System Architecture

3.1 High-Level Architecture

```
graph TB
    subgraph "API Layer"
        API[FastAPI Application]
        WS_EP["/research/web-search"]
        DA_EP["/research/document-analysis"]
        STATUS_EP["/research/{job_id}"]
    end

    subgraph "Agent Layer"
        WSA[WebSearchAgent]
        DAA[DocumentAnalysisAgent]
        RETRY[Retry Wrapper]
        TEMPLATE[Agent Template]
    end

    subgraph "Tool Layer"
        GS[Google Search Tool]
        WS[Web Scraper Tool]
        S3[S3 Document Reader]
        MOCK[Mock Mode Fallback]
    end

    subgraph "LLM Layer"
        BEDROCK[AWS Bedrock]
        CLAUDE[Claude Sonnet 4]
    end
```

```
subgraph "Validation Layer"
    JSON_EXT[JSON Extraction]
    JSON_PARSE[JSON Parsing]
    PYDANTIC[Pydantic Validation]
    BUSINESS[Business Logic Validation]
end
```

```
API --> WS_EP
API --> DA_EP
API --> STATUS_EP
```

```
WS_EP --> WSA
DA_EP --> DAA
```

```
WSA --> RETRY
DAA --> RETRY
RETRY --> TEMPLATE
```

```
TEMPLATE --> GS
TEMPLATE --> WS
TEMPLATE --> S3
```

```
GS .fallback.-> MOCK
WS .fallback.-> MOCK
S3 .fallback.-> MOCK
```

```
TEMPLATE --> BEDROCK
BEDROCK --> CLAUDE
```

```
CLAUDE --> JSON_EXT
JSON_EXT --> JSON_PARSE
JSON_PARSE --> PYDANTIC
PYDANTIC --> BUSINESS
```

```
style API fill:#e1f5ff
style WSA fill:#fff3e0
```

```
style DAA fill:#fff3e0  
style BEDROCK fill:#f3e5f5  
style CLAUDE fill:#f3e5f5
```

3.2 Data Flow Overview (ROI Model Extraction)

```
sequenceDiagram  
    participant User  
    participant API  
    participant BackgroundTask  
    participant Agent  
    participant Tools  
    participant LLM  
    participant Validation  
  
    User->>API: POST /research/web-search<br/>(Extract ROI Model Data)  
    API->>API: Create job (pending)  
    API->>BackgroundTask: Schedule execution  
    API-->User: 202 Accepted (job_id)  
  
    BackgroundTask->>BackgroundTask: Update status: in_progress  
    BackgroundTask->>Agent: Create agent with retry  
    Agent->>Tools: Load tools  
    Tools-->>Agent: google_search, scrape_webpage  
  
    Agent->>LLM: Execute ROI Model research  
    loop For each search (15-25)  
        LLM->>Tools: google_search(ROI metrics query)  
        Tools-->>LLM: Search results  
        LLM->>Tools: scrape_webpage(url)  
        Tools-->>LLM: Page content with metrics  
    end  
  
    LLM-->>Agent: Raw response (ROI Model data)  
    Agent->>Validation: Validate ROI Model extraction  
  
    alt Validation Success
```

```

Validation-->Agent: Valid ROI Model data
Agent-->BackgroundTask: ROI Model extraction result
BackgroundTask-->BackgroundTask: Update status: completed
else Validation Failure
    Validation-->Agent: Errors + Feedback
    Agent-->LLM: Retry with feedback
end

User-->API: GET /research/{job_id}
API-->User: Status + ROI Model Data

```

4. WebSearchAgent

Purpose: ROI Model Data Extraction via Web Search

The WebSearchAgent researches healthcare companies to extract **ROI Model data components**:

- **ROI Model Type Indicators:** Keywords and themes suggesting which of 13 ROI Model Types applies
- **Quantitative Metrics:** Baseline costs, savings percentages, population sizes, PMPM values
- **Formula Components:** ROI calculation methods, payback period formulas, savings calculations
- **Clinical Outcomes:** HbA1c reductions, admission rate decreases, adherence improvements
- **Financial Evidence:** Contract savings, cost avoidance, total cost of care reductions
- **Target Audiences:** Health plans, employers, providers, TPAs
- **Industry Benchmarks:** Typical ranges for metrics in each ROI Model Type

4.1 Agent Creation Flow

```

flowchart TD
    START([Start: create_web_search_agent_with_retry]) --> GET_MODEL{Mod
    GET_MODEL -->|No| DEFAULT[Get default model from config]

```

```

GET_MODEL -->|Yes| USE_MODEL[Use provided model]

DEFAULT --> CREATE_TEMPLATE[Create WebSearchAgentTemplate]
USE_MODEL --> CREATE_TEMPLATE

CREATE_TEMPLATE --> SET_MODE{Research Mode}
SET_MODE -->|autonomous| AUTO["Set: 15-25 searches mode"]
SET_MODE -->|manual| MANUAL["Set: 5-15 searches mode"]

AUTO --> LOAD_INST[Load web_search_instructions.md]
MANUAL --> LOAD_INST

LOAD_INST --> ADD_CONTEXT[Append mode-specific context]
ADD_CONTEXT --> LOAD_TOOLS[Load Tools]

LOAD_TOOLS --> GS_TOOL[Google Search Tool]
LOAD_TOOLS --> WS_TOOL[Web Scraper Tool]

GS_TOOL --> CHECK_API{API Keys?}
WS_TOOL --> CHECK_API

CHECK_API -->|Present| REAL_MODE[Real API Mode]
CHECK_API -->|Missing| MOCK_MODE[Mock Mode]

REAL_MODE --> CREATE_AGENT[Create MareAgent]
MOCK_MODE --> CREATE_AGENT

CREATE_AGENT --> WRAP_RETRY[Wrap with retry logic]
WRAP_RETRY --> SET_MAX[Set max_retries=3]
SET_MAX --> RETURN([Return: Agent with retry])

style START fill:#e8f5e9
style RETURN fill:#e8f5e9
style REAL_MODE fill:#c8e6c9
style MOCK_MODE fill:#ffecb3

```

4.2 Execution Flow (Autonomous Mode) - ROI Model Data Extraction

```
flowchart TD
    START([Agent.run called]) --> BUILD_MSG[Build research message]
    BUILD_MSG --> COMPANY["Company: client_company_name"]
    BUILD_MSG --> INDUSTRY[Industry hint if provided]
    BUILD_MSG --> CONTEXT[Additional context if provided]

    COMPANY --> EXECUTE[Execute LLM with tools]
    INDUSTRY --> EXECUTE
    CONTEXT --> EXECUTE

    EXECUTE --> PHASE1["Phase 1: Company Information"]
    PHASE1 --> S1["google_search('company overview')"]
    S1 --> S2["google_search('company mission')"]
    S2 --> S3["scrape_webpage(company_website)"]

    S3 --> PHASE2["Phase 2: Product & Solutions"]
    PHASE2 --> S4["google_search('products solutions')"]
    S4 --> S5["google_search('platform features')"]

    S5 --> PHASE3["Phase 3: ROI Model Components"]
    PHASE3 --> S6["google_search('ROI cost savings')"]
    S6 --> S7["google_search('financial metrics PMPM')"]

    S7 --> PHASE4["Phase 4: Clinical Outcomes & Metrics"]
    PHASE4 --> S8["google_search('clinical outcomes data')"]
    S8 --> S9["google_search('quality metrics study')"]

    S9 --> PHASE5["Phase 5: Competitive Landscape"]
    PHASE5 --> S10["google_search('competitors')"]
    S10 --> S11["google_search('vs competitor ROI')"]

    S11 --> PHASE6["Phase 6: Market Presence"]
    PHASE6 --> S12["google_search('news press release')"]
    S12 --> S13["google_search('awards partnerships')"]
```

```

S13 --> PHASE7["Phase 7: Customer ROI Evidence"]
PHASE7 --> S14["google_search('case study ROI')"]
S14 --> S15["google_search('customer results metrics')"]

S15 --> CHECK_COUNT{Total searches >= 15?}
CHECK_COUNT -->|No| MORE["google_search(additional ROI data)"]
MORE --> CHECK_COUNT
CHECK_COUNT -->|Yes| SYNTHESIZE[Synthesize ROI Model data into JSON]

SYNTHESIZE --> RETURN_JSON[Return WebSearchResult with ROI Model dat
RETURN_JSON --> END([End: Raw LLM response])

style START fill:#e3f2fd
style END fill:#e3f2fd
style PHASE1 fill:#fff3e0
style PHASE2 fill:#fff3e0
style PHASE3 fill:#ffcccc
style PHASE4 fill:#fff3e0
style PHASE5 fill:#fff3e0
style PHASE6 fill:#fff3e0
style PHASE7 fill:#fff3e0

```

4.3 Execution Flow (Manual Mode) - ROI Model Data Extraction

```

flowchart TD
START([Agent.run called with prompts]) --> PARSE_PROMPTS[Parse user
PARSE_PROMPTS --> P1["Prompt 1: Research solution"]
PARSE_PROMPTS --> P2["Prompt 2: Find clinical outcomes"]
PARSE_PROMPTS --> P3["Prompt 3: Identify target segments"]

P1 --> EXECUTE1[Execute searches for Prompt 1]
EXECUTE1 --> S1["google_search(query 1)"]
S1 --> S2["google_search(query 2)"]
S2 --> SCRAPE1["scrape_webpage(url 1)"]

```

```

P2 --> EXECUTE2[Execute searches for Prompt 2]
EXECUTE2 --> S3["google_search(query 3)"]
S3 --> S4["google_search(query 4)"]

P3 --> EXECUTE3[Execute searches for Prompt 3]
EXECUTE3 --> S5["google_search(query 5)"]

SCRAPE1 --> CHECK_COUNT{Total searches >= 5?}
S4 --> CHECK_COUNT
S5 --> CHECK_COUNT

CHECK_COUNT -->|No| MORE["Additional searches if needed"]
MORE --> CHECK_COUNT

CHECK_COUNT -->|Yes| SYNTHESIZE[Synthesize findings]
SYNTHESIZE --> ADDRESS_P1[Address Prompt 1 findings]
SYNTHESIZE --> ADDRESS_P2[Address Prompt 2 findings]
SYNTHESIZE --> ADDRESS_P3[Address Prompt 3 findings]

ADDRESS_P1 --> COMPILE[Compile WebSearchResult]
ADDRESS_P2 --> COMPILE
ADDRESS_P3 --> COMPILE

COMPILE --> RETURN_JSON[Return JSON]
RETURN_JSON --> END([End: Raw LLM response])

style START fill:#e3f2fd
style END fill:#e3f2fd
style P1 fill:#e3f2fd
style P2 fill:#e3f2fd
style P3 fill:#e3f2fd

```

4.4 Tool Integration Flow

```

flowchart TD
    subgraph "Google Search Tool"
        GS_START[google_search called] --> GS_CHECK{TAVILY_API_KEY set?}

```

```

GS_CHECK -->|Yes| GS_REAL[Call Tavily API]
GS_CHECK -->|No| GS_MOCK[Return mock results]

GS_REAL --> GS_PARSE[Parse results]
GS_MOCK --> GS_PARSE

GS_PARSE --> GS_RETURN["Return SearchResult[]"]
end

subgraph "Web Scraper Tool"
WS_START[scrape_webpage called] --> WS_CHECK{FIRECRAWL_API_KEY S
WS_CHECK -->|Yes| WS_FIRE[Call Firecrawl API]
WS_CHECK -->|No| WS_BASIC[Basic scraping mode]

WS_FIRE --> WS_CONVERT[Convert to markdown]
WS_BASIC --> WS_CONVERT

WS_CONVERT --> WS_RETURN[Return page content]
end

AGENT[LLM Agent Execution] --> GS_START
AGENT --> WS_START

GS_RETURN --> AGENT_CTX[Add to agent context]
WS_RETURN --> AGENT_CTX

AGENT_CTX --> NEXT_SEARCH{More searches needed?}
NEXT_SEARCH -->|Yes| AGENT
NEXT_SEARCH -->|No| SYNTHESIZE[Synthesize results]

style GS_REAL fill:#c8e6c9
style GS_MOCK fill:#ffecb3
style WS_FIRE fill:#c8e6c9
style WS_BASIC fill:#ffecb3

```

4.5 Retry Logic with Validation Feedback

```
flowchart TD
    START([run_with_retry called]) --> ATTEMPT_1["Attempt 1/3"]

    ATTEMPT_1 --> EXEC_1[Execute agent.run]
    EXEC_1 --> EXTRACT_1[Extract JSON from response]

    EXTRACT_1 --> JSON_OK_1{JSON extracted?}
    JSON_OK_1 -->|No| ERROR_1[Build error feedback]
    JSON_OK_1 -->|Yes| PARSE_1[Parse JSON]

    PARSE_1 --> PARSE_OK_1{Valid JSON syntax?}
    PARSE_OK_1 -->|No| ERROR_1
    PARSE_OK_1 -->|Yes| PYDANTIC_1[Pydantic validation]

    PYDANTIC_1 --> PYD_OK_1{Pydantic valid?}
    PYD_OK_1 -->|No| ERROR_1
    PYD_OK_1 -->|Yes| BUSINESS_1[Business logic validation]

    BUSINESS_1 --> BUS_OK_1{Business rules pass?}
    BUS_OK_1 -->|Yes| SUCCESS([Return validated result])
    BUS_OK_1 -->|No| ERROR_1

    ERROR_1 --> ATTEMPT_2["Attempt 2/3"]
    ATTEMPT_2 --> FEEDBACK_2["Append feedback to message:<br/>- Previous"]

    FEEDBACK_2 --> EXEC_2[Execute agent.run with feedback]
    EXEC_2 --> EXTRACT_2[Extract JSON]

    EXTRACT_2 --> JSON_OK_2{JSON extracted?}
    JSON_OK_2 -->|No| ERROR_2[Build error feedback]
    JSON_OK_2 -->|Yes| PARSE_2[Parse JSON]

    PARSE_2 --> PARSE_OK_2{Valid JSON syntax?}
    PARSE_OK_2 -->|No| ERROR_2
    PARSE_OK_2 -->|Yes| PYDANTIC_2[Pydantic validation]
```

```

PYDANTIC_2 --> PYD_OK_2{Pydantic valid?}
PYD_OK_2 -->|No| ERROR_2
PYD_OK_2 -->|Yes| BUSINESS_2[Business logic validation]

BUSINESS_2 --> BUS_OK_2{Business rules pass?}
BUS_OK_2 -->|Yes| SUCCESS
BUS_OK_2 -->|No| ERROR_2

ERROR_2 --> ATTEMPT_3["Attempt 3/3"]
ATTEMPT_3 --> FEEDBACK_3["Append cumulative feedback:<br/>- All prev

FEEDBACK_3 --> EXEC_3[Execute agent.run with full feedback]
EXEC_3 --> EXTRACT_3[Extract JSON]

EXTRACT_3 --> JSON_OK_3{JSON extracted?}
JSON_OK_3 -->|No| FAILURE[Raise RuntimeError]
JSON_OK_3 -->|Yes| PARSE_3[Parse JSON]

PARSE_3 --> PARSE_OK_3{Valid JSON syntax?}
PARSE_OK_3 -->|No| FAILURE
PARSE_OK_3 -->|Yes| PYDANTIC_3[Pydantic validation]

PYDANTIC_3 --> PYD_OK_3{Pydantic valid?}
PYD_OK_3 -->|No| FAILURE
PYD_OK_3 -->|Yes| BUSINESS_3[Business logic validation]

BUSINESS_3 --> BUS_OK_3{Business rules pass?}
BUS_OK_3 -->|Yes| SUCCESS
BUS_OK_3 -->|No| FAILURE

style START fill:#e8f5e9
style SUCCESS fill:#c8e6c9
style FAILURE fill:#ffccdd2
style FEEDBACK_2 fill:#ffff9c4
style FEEDBACK_3 fill:#ffecb3

```

5. DocumentAnalysisAgent

Purpose: ROI Model Data Extraction from ROI Story Documents

The DocumentAnalysisAgent analyzes **ROI Story documents** (client-uploaded PDFs, DOCX, TXT) to extract **ROI Model data components**:

- **ROI Model Type Indicators:** Document themes suggesting which of 13 ROI Model Types applies
- **Quantitative Metrics:** Baseline costs, target savings, population sizes, financial projections
- **Formula Components:** ROI calculations, cost methodologies referenced in documents
- **Clinical Metrics:** Patient outcomes, quality measures, utilization rates
- **Financial Data:** Intervention costs, implementation timelines, payback periods
- **Assumptions:** Stated assumptions, industry benchmarks mentioned
- **Data Requirements:** Identified data sources needed for calculations

5.1 Agent Creation Flow

```
flowchart TD
    START([Start: create_document_analysis_agent_with_retry]) --> GET_MODEL
    GET_MODEL -->|No| DEFAULT[Get default model]
    GET_MODEL -->|Yes| USE_MODEL[Use provided model]

    DEFAULT --> CREATE_TEMPLATE[Create DocumentAnalysisAgentTemplate]
    USE_MODEL --> CREATE_TEMPLATE

    CREATE_TEMPLATE --> LOAD_INST[Load document_analysis_instructions.md]
    LOAD_INST --> LOAD_TOOL[Load S3 Document Reader Tool]

    LOAD_TOOL --> CHECK_AWS{AWS credentials?}
    CHECK_AWS -->|Present| REAL_S3[Real S3 mode]
    CHECK_AWS -->|Missing| MOCK_S3[Mock mode]

    REAL_S3 --> CREATE_AGENT[Create MareAgent]
    MOCK_S3 --> CREATE_AGENT
```

```

CREATE_AGENT --> WRAP_RETRY[Wrap with retry logic]
WRAP_RETRY --> SET_MAX[Set max_retries=3]
SET_MAX --> RETURN([Return: Agent with retry])

style START fill:#e8f5e9
style RETURN fill:#e8f5e9
style REAL_S3 fill:#c8e6c9
style MOCK_S3 fill:#ffecb3

```

5.2 Document Analysis Execution Flow (ROI Story Data Extraction)

```

flowchart TD
START([Agent.run called with ROI Story documents]) --> PARSE_DOCS[Pa
  PARSE_DOCS --> DOC1["Document 1: s3://path/roi_story.pdf"]
  PARSE_DOCS --> DOC2["Document 2: s3://path/case_study.pdf"]
  PARSE_DOCS --> DOC3["Document 3: s3://path/financial_projections.pdf"]

  DOC1 --> READ1["read_document('s3://path/roi_story.pdf')"]
  DOC2 --> READ2["read_document('s3://path/case_study.pdf')"]
  DOC3 --> READ3["read_document('s3://path/financial_projections.pdf')"]

  READ1 --> EXTRACT1[Extract text from PDF]
  READ2 --> EXTRACT2[Extract text from PDF]
  READ3 --> EXTRACT3[Extract text from PDF]

  EXTRACT1 --> ANALYZE1[Analyze Document 1]
  EXTRACT2 --> ANALYZE2[Analyze Document 2]
  EXTRACT3 --> ANALYZE3[Analyze Document 3]

  ANALYZE1 --> FIND_ROI1[Extract ROI Model components]
  ANALYZE1 --> FIND_VAR1[Extract variables PMPM, costs]
  ANALYZE1 --> FIND_FM1[Extract financial metrics]

  ANALYZE2 --> FIND_ROI2[Extract ROI Model components]

```

```

ANALYZE2 --> FIND_VAR2[Extract variables]
ANALYZE2 --> FIND_FM2[Extract financial metrics]

ANALYZE3 --> FIND_ROI3[Extract ROI Model components]
ANALYZE3 --> FIND_VAR3[Extract variables]
ANALYZE3 --> FIND_FM3[Extract financial metrics]

FIND_ROI1 --> SYNTHESIZE[Synthesize all findings]
FIND_VAR1 --> SYNTHESIZE
FIND_FM1 --> SYNTHESIZE
FIND_ROI2 --> SYNTHESIZE
FIND_VAR2 --> SYNTHESIZE
FIND_FM2 --> SYNTHESIZE
FIND_ROI3 --> SYNTHESIZE
FIND_VAR3 --> SYNTHESIZE
FIND_FM3 --> SYNTHESIZE

SYNTHESIZE --> DEDUPE[Deduplicate across documents]
DEDUPE --> CONF_SCORE[Calculate confidence scores]
CONF_SCORE --> COMPILE[Compile DocumentAnalysisResult with ROI Model]

COMPILE --> RETURN_JSON[Return JSON]
RETURN_JSON --> END([End: Raw LLM response])

style START fill:#e3f2fd
style END fill:#e3f2fd
style ANALYZE1 fill:#fff3e0
style ANALYZE2 fill:#fff3e0
style ANALYZE3 fill:#fff3e0

```

4.3 Document Type Processing

```

flowchart TD
    START[read_document called] --> PARSE_PATH[Parse S3 path]
    PARSE_PATH --> EXTRACT_EXT[Extract file extension]

    EXTRACT_EXT --> CHECK_TYPE{File type?}

```

```
CHECK_TYPE -->|.pdf| PDF_FLOW[PDF Processing]
CHECK_TYPE -->|.docx| DOCX_FLOW[DOCX Processing]
CHECK_TYPE -->|.txt| TXT_FLOW[TXT Processing]
CHECK_TYPE -->|other| ERROR[Unsupported format]

PDF_FLOW --> DOWNLOAD_PDF[Download from S3]
DOWNLOAD_PDF --> PYPDF2[PyPDF2 extraction]
PYPDF2 --> EXTRACT_TEXT_PDF[Extract text from pages]
EXTRACT_TEXT_PDF --> COMBINE_PDF[Combine pages]

DOCX_FLOW --> DOWNLOAD_DOCX[Download from S3]
DOWNLOAD_DOCX --> PYTHON_DOCX[python-docx extraction]
PYTHON_DOCX --> EXTRACT_TEXT_DOCX[Extract paragraphs]
EXTRACT_TEXT_DOCX --> COMBINE_DOCX[Combine paragraphs]

TXT_FLOW --> DOWNLOAD_TXT[Download from S3]
DOWNLOAD_TXT --> READ_TXT[Read as UTF-8]

COMBINE_PDF --> RETURN[Return extracted text]
COMBINE_DOCX --> RETURN
READ_TXT --> RETURN
ERROR --> RETURN_EMPTY[Return empty string]

RETURN --> END([End: Document text])
RETURN_EMPTY --> END

style START fill:#e8f5e9
style END fill:#e8f5e9
style PDF_FLOW fill:#ffccbc
style DOCX_FLOW fill:#c5cae9
style TXT_FLOW fill:#dcedc8
style ERROR fill:#ffcdd2
```

4.4 Tool Integration (S3 Document Reader)

```
graph TD
    subgraph "S3 Document Reader Tool"
        TOOL_START[read_document called] --> VALIDATE_PATH{Valid S3 path}
        VALIDATE_PATH -->|No| MOCK_PATH["Treat as mock/test path"]
        VALIDATE_PATH -->|Yes| CHECK_AWS{AWS configured?}

        CHECK_AWS -->|No| MOCK_MODE[Return mock document content]
        CHECK_AWS -->|Yes| CONNECT_S3[Connect to S3]

        CONNECT_S3 --> DOWNLOAD[Download document]
        DOWNLOAD --> DETECT_TYPE[Detect file type]

        DETECT_TYPE --> PDF_PROC["PDF: PyPDF2"]
        DETECT_TYPE --> DOCX_PROC["DOCX: python-docx"]
        DETECT_TYPE --> TXT_PROC["TXT: UTF-8 read"]

        PDF_PROC --> EXTRACT[Extract text]
        DOCX_PROC --> EXTRACT
        TXT_PROC --> EXTRACT

        MOCK_PATH --> MOCK_CONTENT[Generate mock content]
        MOCK_MODE --> MOCK_CONTENT

        MOCK_CONTENT --> RETURN[Return text content]
        EXTRACT --> RETURN
    end

    AGENT[LLM Agent Execution] --> TOOL_START
    RETURN --> AGENT_CTX[Add document content to context]

    AGENT_CTX --> MORE_DOCS{More documents?}
    MORE_DOCS -->|Yes| AGENT
    MORE_DOCS -->|No| ANALYZE[Analyze all documents]

    style CONNECT_S3 fill:#c8e6c9
```

```
style MOCK_MODE fill:#ffecb3  
style MOCK_CONTENT fill:#ffecb3
```

4.5 Retry Logic (Same as WebSearchAgent)

```
flowchart LR  
    START([Attempt 1]) --> VALIDATE_1[4-Layer Validation]  
    VALIDATE_1 --> SUCCESS_1{Pass?}  
    SUCCESS_1 -->|Yes| DONE([Return result])  
    SUCCESS_1 -->|No| FEEDBACK_1[Build error feedback]  
  
    FEEDBACK_1 --> ATTEMPT_2([Attempt 2])  
    ATTEMPT_2 --> VALIDATE_2[4-Layer Validation]  
    VALIDATE_2 --> SUCCESS_2{Pass?}  
    SUCCESS_2 -->|Yes| DONE  
    SUCCESS_2 -->|No| FEEDBACK_2[Add to error feedback]  
  
    FEEDBACK_2 --> ATTEMPT_3([Attempt 3])  
    ATTEMPT_3 --> VALIDATE_3[4-Layer Validation]  
    VALIDATE_3 --> SUCCESS_3{Pass?}  
    SUCCESS_3 -->|Yes| DONE  
    SUCCESS_3 -->|No| FAIL([Raise RuntimeError])  
  
    style START fill:#e3f2fd  
    style DONE fill:#c8e6c9  
    style FAIL fill:#ffcdd2
```

5. API Layer

5.1 API Request Flow (Sequence Diagram)

```
sequenceDiagram  
    autonumber  
    participant Client  
    participant FastAPI
```

```
participant BackgroundTask
participant Agent
participant Tools
participant LLM
participant Validation
participant JobStore

Client->>FastAPI: POST /research/web-search<br/>{company_name, mode,
FastAPI->>FastAPI: Validate request<br/>(check manual mode has prompt)
FastAPI->>FastAPI: Generate job_id<br/>"research_web_abc123"
FastAPI->>JobStore: Create job record<br/>{status: "pending", progress: 0}
FastAPI->>BackgroundTask: Schedule execute_web_search_research
FastAPI-->>Client: 202 Accepted<br/>{job_id, estimated_time: 120s}
```

Note over BackgroundTask: Async execution begins

```
BackgroundTask->>JobStore: Update {status: "in_progress", progress: 0}
BackgroundTask->>Agent: create_web_search_agent_with_retry
Agent->>Tools: Load tools (google_search, scrape_webpage)
Tools-->>Agent: Tools ready
```

```
BackgroundTask->>JobStore: Update {progress: 20%}
BackgroundTask->>Agent: Build research message
BackgroundTask->>JobStore: Update {progress: 30%}
```

```
BackgroundTask->>Agent: agent.run(message)
```

```
loop For each retry attempt (max 3)
    Agent->>LLM: Execute with tools
    loop For each search (15-25 searches)
        LLM->>Tools: google_search(query)
        Tools-->>LLM: Search results
        LLM->>Tools: scrape_webpage(url)
        Tools-->>LLM: Page content
    end
    LLM-->>Agent: Raw response text
```

```

Agent->>Validation: Extract JSON
Validation->>Validation: Parse JSON
Validation->>Validation: Pydantic validation
Validation->>Validation: Business logic validation

alt Validation Success
    Validation-->>Agent: Valid WebSearchResult
    Note over Agent: Break retry loop
else Validation Failure
    Validation-->>Agent: Error feedback
    Agent->>LLM: Retry with feedback
end
end

Agent-->>BackgroundTask: WebSearchResult
BackgroundTask->>JobStore: Update {status: "completed", progress: 10}

Note over Client: Client polls for status

Client->>FastAPI: GET /research/{job_id}
FastAPI->>JobStore: Query job
JobStore-->>FastAPI: Job data
FastAPI-->>Client: 200 OK<br/>{status: "completed", result}

```

5.2 Job State Machine

```

stateDiagram-v2
[*] --> Pending: API creates job
Pending --> InProgress: Background task starts

InProgress --> Completed: Agent succeeds
InProgress --> Failed: Agent fails<br/>(after 3 retries)
InProgress --> Failed: Unexpected error

Completed --> [*]
Failed --> [*]

```

```

note right of Pending
    progress_percent: 0%
    status: "pending"
end note

note right of InProgress
    progress_percent: 10% → 30% → 90%
    status: "in_progress"
end note

note right of Completed
    progress_percent: 100%
    status: "completed"
    result: WebSearchResult or DocumentAnalysisResult
end note

note right of Failed
    status: "failed"
    error: Error message
end note

```

5.3 Background Task Progression

```

flowchart LR
    START([Background task starts]) --> P10["10%: Mark in_progress"]
    P10 --> P20["20%: Create agent"]
    P20 --> P30["30%: Build message"]
    P30 --> P90["90%: Agent execution<br/>This takes 60-120 seconds"]
    P90 --> P100["100%: Validation & save"]
    P100 --> DONE([Mark completed])

    style START fill:#e3f2fd
    style P90 fill:#fff9c4
    style DONE fill:#c8e6c9

```

5.4 API Endpoints

Method	Endpoint	Description	Response
POST	/research/web-search	Initiate web search research	202 Accepted + job_id
POST	/research/document-analysis	Initiate document analysis	202 Accepted + job_id
GET	/research/{job_id}	Get job status and results	200 OK + job data
GET	/research/	List jobs with filters	200 OK + paginated list
GET	/research/stats/summary	Get aggregate statistics	200 OK + stats
POST	/research/validate	Validate research result	200 OK + validation report

6. Validation Pipeline

6.1 Four-Layer Validation Process

```
flowchart TD
    START([Raw LLM response]) --> LAYER1["Layer 1: JSON Extraction"]

    LAYER1 --> EXTRACT[Extract JSON from text]
    EXTRACT --> MARKDOWN{Contains markdown?}
    MARKDOWN -->|Yes| REGEX["Regex: `\\`json(.*)`\\`"]
    MARKDOWN -->|No| FIND["Find first { to last }"]

    REGEX --> JSON_STR[JSON string]
    FIND --> JSON_STR

    JSON_STR --> LAYER2["Layer 2: JSON Parsing"]
    LAYER2 --> PARSE["json.loads"]
```

```
PARSE --> SYNTAX_OK{Valid JSON syntax?}
SYNTAX_OK -->|No| ERROR_PARSE[JSON syntax error]
SYNTAX_OK -->|Yes| LAYER3["Layer 3: Pydantic Validation"]

LAYER3 --> PYDANTIC[Pydantic model validation]
PYDANTIC --> TYPE_OK{All types valid?}
TYPE_OK -->|No| ERROR_PYDANTIC["Type/constraint error"]
TYPE_OK -->|Yes| FIELD_OK{All required fields?}
FIELD_OK -->|No| ERROR_PYDANTIC
FIELD_OK -->|Yes| LAYER4["Layer 4: Business Logic"]

LAYER4 --> BIZ_RULES[Business rule validation]

BIZ_RULES --> CHECK_SEARCHES{Minimum searches?}
CHECK_SEARCHES -->|No| WARN_SEARCHES["Warning: Too few searches"]
CHECK_SEARCHES -->|Yes| CHECK_VP{Min value props?}

CHECK_VP -->|No| ERROR_VP["Error: Missing value props"]
CHECK_VP -->|Yes| CHECK_DESC{Description length?}

CHECK_DESC -->|No| ERROR_DESC["Error: Description too short"]
CHECK_DESC -->|Yes| CHECK_CONF{Confidence reasonable?}

CHECK_CONF -->|No| WARN_CONF["Warning: Low confidence"]
CHECK_CONF -->|Yes| CHECK_SOURCES{Sufficient sources?}

CHECK_SOURCES -->|No| WARN_SOURCES["Warning: Few sources"]
CHECK_SOURCES -->|Yes| SUCCESS([Validation passed])

ERROR_PARSE --> RETRY{Attempt < max?}
ERROR_PYDANTIC --> RETRY
ERROR_VP --> RETRY
ERROR_DESC --> RETRY

RETRY -->|Yes| FEEDBACK[Build error feedback]
RETRY -->|No| FAIL([Validation failed])
```

```

FEEDBACK --> APPEND[Append feedback to message]
APPEND --> RERUN[Re-run agent with feedback]
RERUN --> START

WARN_SEARCHES --> SUCCESS
WARN_CONF --> SUCCESS
WARN_SOURCES --> SUCCESS

style START fill:#e3f2fd
style SUCCESS fill:#c8e6c9
style FAIL fill:#ffcdd2
style LAYER1 fill:#fff3e0
style LAYER2 fill:#e1f5ff
style LAYER3 fill:#f3e5f5
style LAYER4 fill:#e8f5e9

```

6.2 Validation Rules

WebSearchResult Validation

```

flowchart TD
    START[WebSearchResult validation] --> CHECK_MODE{Research mode?}

    CHECK_MODE -->|autonomous| AUTO_CHECK["searches_performed >= 15"]
    CHECK_MODE -->|manual| MANUAL_CHECK["searches_performed >= 5"]

    AUTO_CHECK --> VP_CHECK["value_propositions.length >= 1"]
    MANUAL_CHECK --> VP_CHECK

    VP_CHECK --> DESC_CHECK["company_overview.description.length >= 50"]
    DESC_CHECK --> SOURCES_CHECK["sources.length >= 1"]
    SOURCES_CHECK --> CONF_CHECK["0.0 <= confidence_score <= 1.0"]

    CONF_CHECK --> HIGH_CONF["Any high-confidence value props?"]
    HIGH_CONF -->|No| WARN_HIGH["Warning: No high-confidence VPs"]
    HIGH_CONF -->|Yes| DONE

```

```
WARN_HIGH --> DONE([Validation report])
```

```
style START fill:#e3f2fd  
style DONE fill:#c8e6c9
```

DocumentAnalysisResult Validation

```
flowchart TD  
START[DocumentAnalysisResult validation] --> DOCS_CHECK["documents_a  
DOCS_CHECK --> VP_CHECK["extracted_value_propositions.length >= 1"]  
VP_CHECK --> VP_NAME["Each VP name.length >= 10"]  
VP_NAME --> VP_DESC["Each VP description.length >= 30"]  
  
VP_DESC --> CONF_CHECK["0.0 <= overall_confidence <= 1.0"]  
  
CONF_CHECK --> LOW_CONF{Confidence < 0.5?}  
LOW_CONF -->|Yes| WARN_CONF["Warning: Low confidence"]  
LOW_CONF -->|No| HIGH_VP  
  
HIGH_VP["Any high-confidence VPs?"]  
HIGH_VP -->|No| WARN_VP["Warning: No high-confidence extractions"]  
HIGH_VP -->|Yes| DONE  
  
WARN_CONF --> DONE  
WARN_VP --> DONE([Validation report])  
  
style START fill:#e3f2fd  
style DONE fill:#c8e6c9
```

7. Data Models

7.1 WebSearchResult Model Hierarchy

```
classDiagram

    class WebSearchResult {
        +int searches_performed
        +List<str> queries
        +CompanyOverview company_overview
        +List<ValuePropositionEvidence> value_propositions
        +List<ClinicalOutcomeEvidence> clinical_outcomes
        +ROIFramework roi_framework
        +CompetitivePositioning competitive_positioning
        +List<str> target_audiences
        +List<str> sources
        +str research_mode
        +float confidence_score
        +List<str> missing_information
        +List<str> assumptions_made
        +datetime research_timestamp
    }

    class CompanyOverview {
        +str name
        +str description
        +str mission
        +List<str> target_markets
        +str website
    }

    class ValuePropositionEvidence {
        +str name
        +str description
        +str evidence_type
        +List<str> supporting_sources
        +str confidence
    }
```

```

class ClinicalOutcomeEvidence {
    +str outcome
    +str metric_value
    +str evidence_type
    +str source
    +str confidence
}

class ROIFramework {
    +str typical_roi_range
    +str payback_period
    +List~str~ cost_savings_areas
    +str evidence_quality
    +List~str~ supporting_sources
}

class CompetitivePositioning {
    +List~str~ main_competitors
    +List~str~ unique_advantages
    +List~str~ market_differentiators
    +str market_position
}

WebSearchResult *-- CompanyOverview
WebSearchResult *-- ValuePropositionEvidence
WebSearchResult *-- ClinicalOutcomeEvidence
WebSearchResult *-- ROIFramework
WebSearchResult *-- CompetitivePositioning

```

7.2 DocumentAnalysisResult Model Hierarchy

```

classDiagram
    class DocumentAnalysisResult {
        +int documents_analyzed
        +List~str~ document_names
        +List~ExtractedValueProposition~ extracted_value_propositions
    }

```

```
+List~ExtractedClinicalOutcome~ clinical_outcomes
+List~ExtractedFinancialMetric~ financial_metrics
+List~str~ target_audiences
+List~ExtractedCompetitiveAdvantage~ competitive_advantages
+str additional_context
+float overall_confidence
+List~str~ missing_information
+datetime analysis_timestamp
}

class ExtractedValueProposition {
    +str name
    +str description
    +Dict metrics
    +str source_document
    +List~int~ page_numbers
    +str confidence
}

class ExtractedClinicalOutcome {
    +str outcome
    +str metric_value
    +str source_document
    +List~int~ page_numbers
    +str confidence
}

class ExtractedFinancialMetric {
    +str metric_name
    +str value
    +str context
    +str source_document
    +List~int~ page_numbers
}

class ExtractedCompetitiveAdvantage {
    +str advantage
}
```

```

        +str supporting_evidence
        +str source_document
    }

DocumentAnalysisResult *-- ExtractedValueProposition
DocumentAnalysisResult *-- ExtractedClinicalOutcome
DocumentAnalysisResult *-- ExtractedFinancialMetric
DocumentAnalysisResult *-- ExtractedCompetitiveAdvantage

```

7.3 API Request/Response Models

```

classDiagram

class WebSearchRequest {
    +str client_company_name
    +str research_mode
    +str industry_hint
    +List~str~ prompts
    +str additional_context
    +int max_searches
}

class DocumentAnalysisRequest {
    +List~str~ document_ids
    +str additional_context
}

class ResearchJobResponse {
    +str job_id
    +str status
    +str message
    +str research_type
    +datetime created_at
    +int estimated_completion_seconds
}

class ResearchJobStatusResponse {
    +str job_id

```

```

+str status
+str research_type
+int progress_percent
+datetime created_at
+datetime started_at
+datetime completed_at
+dict result
+str error
}

class ResearchJobListResponse {
    +int total
    +int page
    +int page_size
    +List<ResearchJobStatusResponse> jobs
}

class ResearchStatsResponse {
    +int total_jobs
    +int web_search_jobs
    +int document_analysis_jobs
    +int completed_jobs
    +int failed_jobs
    +float average_duration_seconds
    +float success_rate
}

ResearchJobListResponse *-- ResearchJobStatusResponse

```

8. Component Specifications

8.1 WebSearchAgent

File: agents/web_search_agent.py

Key Functions:

- `create_web_search_agent()` - Creates base agent
- `create_web_search_agent_with_retry()` - Wraps with retry logic
- `extract_json_from_response()` - Extracts JSON from LLM output

Configuration:

```
research_mode: "autonomous" | "manual"
max_retries: int = 3
model: AWS Bedrock Claude Sonnet 4
tools: [GoogleSearchTool, WebScraperTool]
```

Research Modes:

- **Autonomous:** 15-25 searches across 7 research areas
- **Manual:** 5-15 searches following user prompts

Instructions: `agents/templates/web_search_instructions.md` (376 lines)

8.2 DocumentAnalysisAgent

File: `agents/document_analysis_agent.py`

Key Functions:

- `create_document_analysis_agent()` - Creates base agent
- `create_document_analysis_agent_with_retry()` - Wraps with retry logic
- `extract_json_from_response()` - Extracts JSON from LLM output

Configuration:

```
max_retries: int = 3
model: AWS Bedrock Claude Sonnet 4
tools: [S3DocumentReaderTool]
```

Supported Document Types:

- **PDF:** PyPDF2 extraction
- **DOCX:** python-docx extraction

- **TXT**: UTF-8 text reading

Instructions: `agents/templates/document_analysis_instructions.md` (345 lines)

8.3 Tool Specifications

Google Search Tool

File: `tools/google_search_tool.py`

API: Tavily Search API (AI-optimized search)

Mock Mode: Returns 3 mock search results when `TAVILY_API_KEY` not set

Function Signature:

```
def google_search(  
    query: str,  
    max_results: int = 5,  
    include_raw_content: bool = False  
) -> List[Dict[str, Any]]
```

Returns:

```
[{  
    "title": str,  
    "url": str,  
    "snippet": str,  
    "content": str, # if include_raw_content=True  
    "score": float  
}]
```

Web Scraper Tool

File: `tools/web_scraper_tool.py`

API: Firecrawl API (structured web scraping)

Mock Mode: Returns placeholder content when `FIRECRAWL_API_KEY` not set

Function Signature:

```
def scrape_webpage(url: str) -> str
```

Returns: Markdown-formatted page content

S3 Document Reader Tool

File: tools/s3_document_reader.py

Dependencies: boto3, PyPDF2, python-docx

Mock Mode: Returns mock document content when AWS not configured

Function Signature:

```
def read_document(storage_path: str) -> str
```

Path Format: s3://bucket/key or bucket/key

Returns: Extracted text from document

9. Configuration & Setup

9.1 Environment Variables

```
# Model Configuration
DEFAULT_MODEL_PROVIDER=aws_bedrock
DEFAULT_MODEL_NAME=us.anthropic.claude-sonnet-4-20250514-v1:0

# AWS Bedrock (Required)
AWS_PROFILE=your-profile
AWS_REGION=us-east-1

# Research Tool APIs (Optional - uses mock mode if not set)
TAVILY_API_KEY=tvly-...          # For Google search
FIRECRAWL_API_KEY=fc-...         # For web scraping
```

```
# API Server  
API_HOST=0.0.0.0  
API_PORT=8000  
  
# Logging  
LOG_LEVEL=INFO  
DEBUG_MODE=false
```

9.2 Installation

```
# Create virtual environment  
python3 -m venv venv  
source venv/bin/activate  # Windows: venv\Scripts\activate  
  
# Install dependencies  
pip install -r requirements.txt  
  
# Configure environment  
cp .env.example .env  
# Edit .env with your credentials
```

9.3 Running the System

Start API Server:

```
uvicorn app:app --reload --port 8000
```

Access Documentation:

- Swagger UI: <http://localhost:8000/docs>
- ReDoc: <http://localhost:8000/redoc>

Run Tests:

```
python tests/test_research_api.py
```

10. Example Usage

10.1 Web Search Research (Autonomous Mode)

Request:

```
curl -X POST "http://localhost:8000/research/web-search" \
-H "Content-Type: application/json" \
-d '{
  "client_company_name": "Livongo Health",
  "research_mode": "autonomous",
  "industry_hint": "diabetes management",
  "additional_context": "Focus on ROI and clinical outcomes"
}'
```

Response (202 Accepted):

```
{
  "job_id": "research_web_abc123",
  "status": "pending",
  "message": "Web search research initiated for Livongo Health",
  "research_type": "web_search",
  "created_at": "2025-12-08T10:30:00Z",
  "estimated_completion_seconds": 120
}
```

Check Status:

```
curl "http://localhost:8000/research/research_web_abc123"
```

Response (Completed):

```
{
  "job_id": "research_web_abc123",
  "status": "completed",
```

```

"research_type": "web_search",
"progress_percent": 100,
"result": {
    "searches_performed": 18,
    "queries": ["Livongo diabetes", "Livongo ROI", ...],
    "company_overview": {
        "name": "Livongo Health",
        "description": "Digital health company focused on...",
        "mission": "Empower people with chronic conditions",
        "target_markets": ["Health Plans", "Employers"],
        "website": "https://livongo.com"
    },
    "value_propositions": [
        {
            "name": "Cost Reduction through Prevention",
            "description": "Reduce diabetes-related costs...",
            "evidence_type": "explicit",
            "supporting_sources": ["https://livongo.com/roi"],
            "confidence": "high"
        }
    ],
    "confidence_score": 0.85
}
}

```

10.2 Document Analysis

Request:

```

curl -X POST "http://localhost:8000/research/document-analysis" \
-H "Content-Type: application/json" \
-d '{
    "document_ids": [
        "s3://triton-docs/client123/roi_sheet.pdf",
        "s3://triton-docs/client123/case_study.pdf"
    ],
    "additional_context": "Diabetes management for health plans"
}'

```

Response (202 Accepted):

```
{  
  "job_id": "research_doc_xyz789",  
  "status": "pending",  
  "message": "Document analysis initiated for 2 documents",  
  "research_type": "document_analysis",  
  "estimated_completion_seconds": 60  
}
```

Result:

```
{  
  "documents_analyzed": 2,  
  "document_names": ["roi_sheet.pdf", "case_study.pdf"],  
  "extracted_value_propositions": [{  
    "name": "340% ROI in 24 Months",  
    "description": "Achieve 340% return on investment...",  
    "metrics": {"roi": "340%", "payback_months": 14},  
    "source_document": "roi_sheet.pdf",  
    "page_numbers": [1, 2],  
    "confidence": "high"  
  }],  
  "financial_metrics": [{  
    "metric_name": "24-Month ROI",  
    "value": "340%",  
    "context": "Based on 1000-member population",  
    "source_document": "roi_sheet.pdf",  
    "page_numbers": [1]  
  }],  
  "overall_confidence": 0.95  
}
```

10.3 Python Client Example

```
import requests
import time

BASE_URL = "http://localhost:8000"

# Initiate research
response = requests.post(f"{BASE_URL}/research/web-search", json={
    "client_company_name": "Livongo Health",
    "research_mode": "autonomous",
    "industry_hint": "diabetes management"
})

job_id = response.json()["job_id"]
print(f"Research job created: {job_id}")

# Poll for completion
while True:
    status_response = requests.get(f"{BASE_URL}/research/{job_id}")
    status = status_response.json()

    print(f"Status: {status['status']} ({status.get('progress_percent', 0)}%)")

    if status["status"] == "completed":
        result = status["result"]
        print("\n✓ Research complete!")
        print(f" - Searches: {result['searches_performed']}")
        print(f" - Company: {result['company_overview']['name']}")
        print(f" - Value props: {len(result['value_propositions'])}")
        print(f" - Confidence: {result['confidence_score']}")
        break
    elif status["status"] == "failed":
        print("\nx Research failed: {status['error']}")
        break

    time.sleep(2)
```

Summary

This research agent system provides:

□ **Comprehensive Web Research:** 15-25 searches across 7 key research areas □
□ **Document Intelligence:** Extract value propositions from client materials □**Robust Validation:** 4-layer validation with automatic retry □**Production-Ready API:** Async processing with job tracking □**Mock Mode Support:** Test without external API dependencies □**Rich Data Models:** Structured output with confidence scores

Total Implementation:

- 3,073 lines of production code
 - 2 specialized agents
 - 3 tool integrations
 - 6 REST API endpoints
 - 4-layer validation pipeline
 - 15+ Mermaid diagrams (in this document)
-

Document Version: 1.0.0 **Last Updated:** 2025-12-08 **Status:** Complete □

 Mermaid diagrams are shown as code blocks in this PDF.

For interactive diagrams, use VS Code Markdown Preview Enhanced or visit the [HTML version](#).