

DocuVault - Document Management System with RAG Chatbot

Project Overview

DocuVault is a comprehensive Django-based Document Management System (DMS) designed for secure, efficient, and intelligent document handling. It features role-based access control, version tracking, robust collaboration tools, and a state-of-the-art **AI-powered RAG (Retrieval-Augmented Generation) chatbot** that allows users to query their documents in natural language.

Features

AI-Powered RAG Chatbot

The system integrates a sophisticated RAG pipeline to provide intelligent document interaction:

- **Intelligent Document Search:** Ask questions about your documents in natural language.
- **Context-Aware Responses:** Utilizes the **Qwen2.5-7B** Large Language Model (LLM) for high-quality, accurate answers.
- **Source Citations:** Transparently provides citations, showing exactly which document pages were used to generate the answer.
- **Conversation Memory:** Maintains context across multi-turn conversations for natural follow-up questions.
- **Document Indexing:** Automatic vectorization of PDF documents using **Qwen3-Embedding-0.6B**.
- **GPU Acceleration:** Fully supports CUDA for fast inference on supported hardware.
- **Permission-Aware:** Respects document access levels; users can only query documents they are authorized to view.
- **Hallucination Prevention:** Strictly adheres to the provided context to minimize false information.

User Management & Authentication

- **Guest Users:** Limited access to view public documents and register accounts.
- **Regular Users:** Full document management capabilities for their own files.
- **Admin Users:** Complete system administration and oversight.
- **Role-Based Access Control (RBAC):** Granular permission system with custom roles and hierarchical levels (1-100).
- **Secure Authentication:** Robust login and registration system.
- **User Profiles:** Rich profiles with avatars, bios, and activity history.

Document Management

- **Upload & Storage:** Securely upload and store documents with support for various file types.
- **Access Control Levels:**
 - **Public:** Accessible to all users.
 - **Private:** Restricted to the owner.

- **Role-Based:** Accessible to users with a specific role level or higher.
- **Custom:** Shared with specific individual users.
- **Version Control:** comprehensive history of document versions with change notes.
- **Metadata Tracking:**
 - Title, description, category, and tags.
 - File size, type, view counts, and download counts.
- **Document Locking:** Prevents concurrent edits to ensure data integrity.
- **Soft Delete:** Safety mechanism to recover accidentally deleted documents.

Search & Discovery

- **Advanced Search:** Full-text search capabilities with powerful filters.
- **Organization:**
 - **Categories:** Hierarchical system with color-coding and icons.
 - **Tags:** Flexible tagging for cross-category organization.
- **Filtering & Sorting:** Filter by owner, date, access level; sort by views, title, date, etc.

Collaboration

- **Comments & Replies:** Threaded discussions directly on documents.
- **Sharing:**
 - Direct sharing with specific users.
 - **Shareable Links:** Generate temporary, secure links with password protection, expiration dates, and access limits.
- **Notifications:** Real-time alerts for shares, comments, mentions, and permission changes.

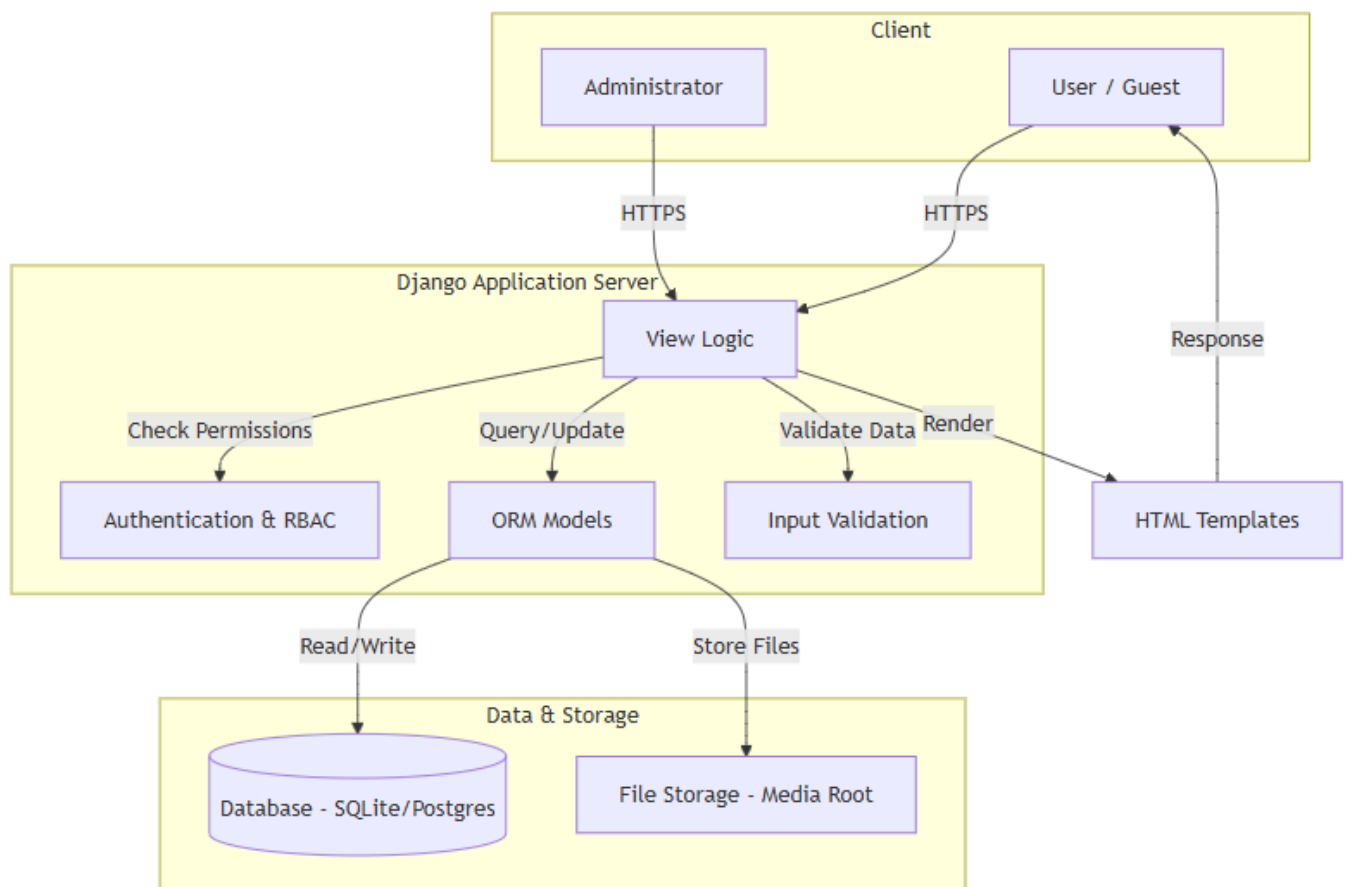
Additional Tools

- **Favorites/Bookmarks:** Quick access to frequently used documents.
 - **Activity Logging:** Detailed audit trail of all user actions.
 - **Dashboard:** Personalized overview of recent activity and stats.
 - **Bulk Operations:** Upload multiple documents simultaneously.
 - **File Preview:** View documents before downloading.
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System Architecture

DMS Architecture Diagram



Project Structure

```
DocuVault/
├── config/                # Django project settings (settings, urls, wsgi)
├── documents/            # Main Django app
│   ├── models.py         # Database models
│   ├── views.py          # View functions
│   ├── rag_views.py       # RAG chatbot views
│   ├── urls.py           # URL routing
│   ├── forms.py          # Form definitions
│   ├── admin.py          # Admin interface
│   ├── rag/              # RAG system modules (See RAG Architecture below)
│   ├── templates/        # HTML templates
│   └── migrations/       # Database migrations
├── media/                # User uploaded files
│   ├── documents/        # Document files
│   ├── avatars/          # User avatars
│   └── rag/              # RAG data (Vector DB)
├── static/               # Static files (CSS, JS)
├── fix_dependencies.py    # Automated installer
├── test_rag.py           # RAG testing script
└── manage.py             # Django management
```

RAG Module Architecture

The RAG system is modular and self-contained within `documents/rag/` :

```
documents/rag/
├── __init__.py          # Module exports
├── config.py            # Central configuration (RAGConfig)
├── document_processor.py # PDF loading and chunking (DocumentProcessor)
├── embeddings.py        # Embedding generation (EmbeddingManager)
├── vector_store.py      # ChromaDB vector database (VectorStore)
├── llm_manager.py       # LLM loading and inference (LLMManager)
├── retriever.py         # Query processing and retrieval (Retriever)
└── conversation.py     # Chatbot with memory (RAGChatbot)
```

Technical Details

Technology Stack

- **Backend:** Django 4.2.26
- **Database:** SQLite (default) / PostgreSQL (production ready)
- **Frontend:** HTML, Vanilla CSS, JavaScript
- **AI/ML Frameworks:**
 - **LangChain** 0.1.20
 - **Transformers** (HuggingFace)
 - **Sentence-Transformers**
 - **PyTorch** (with CUDA support)
- **Vector Database:** ChromaDB (Persistent storage)

RAG Components & Models

The RAG system is built on high-performance open source models and cloud-based inference:

1. Large Language Model (LLM):

- **Model:** `llama-3.1-8b-instant` (via Groq)
- **Provider:** Groq API
- **Role:** Generates natural language answers based on retrieved context with ultra-low latency.
- **Note:** Requires a `GROQ_API_KEY` environment variable.

2. Embeddings:

- **Model:** `all-MiniLM-L6-v2`
- **Library:** Sentence-Transformers
- **Role:** Converts text into semantic vector representations for similarity search.
- **Performance:** Fast and efficient model optimized for semantic search tasks.

3. Vector Store:

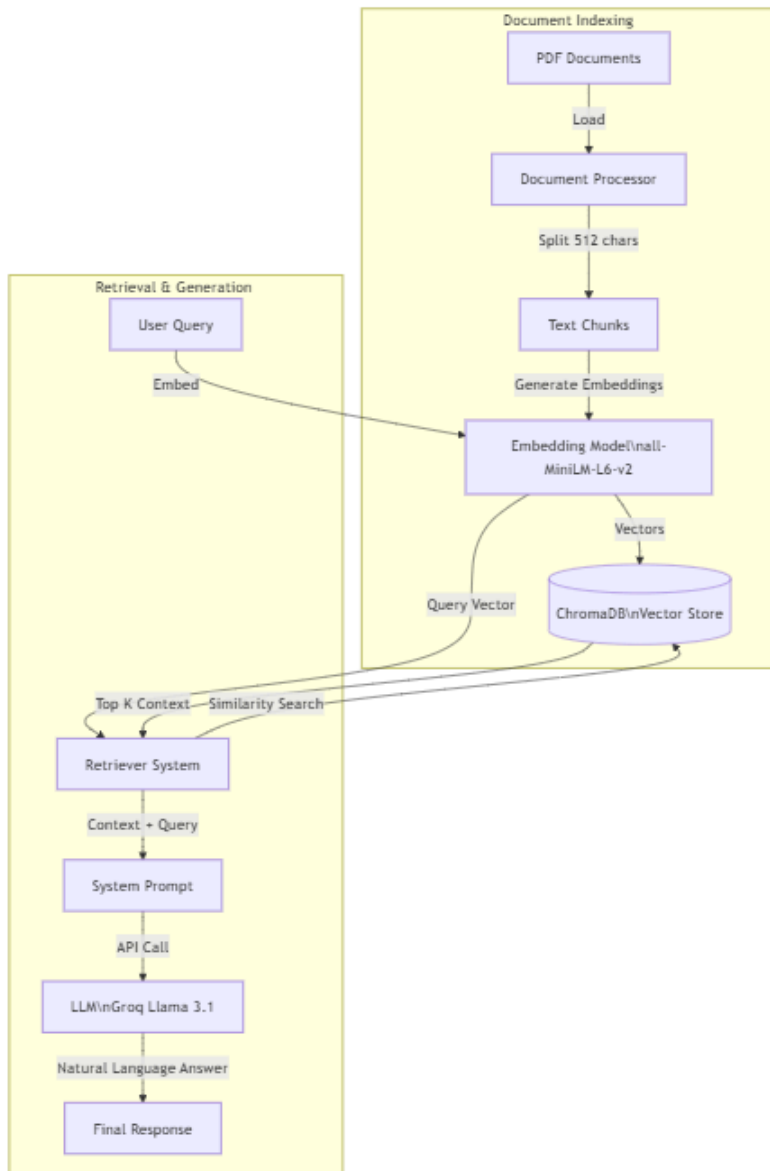
- **Technology:** ChromaDB
- **Role:** Stores document embeddings and metadata for fast retrieval.

4. Configuration (`config.py`):

- **Chunk Size:** 512 characters
- **Chunk Overlap:** 100 characters
- **Similarity Threshold:** 0.15
- **Max New Tokens:** 512

RAG Workflow

RAG Flowchart



1. **Ingestion:** PDFs are loaded via `DocumentProcessor`.
2. **Chunking:** Text is split into overlapping chunks (512 chars).
3. **Embedding:** `EmbeddingManager` converts chunks to vectors using `all-MiniLM-L6-v2`.
4. **Storage:** Vectors are stored in `VectorStore` (ChromaDB).
5. **Retrieval:** User query is embedded; `Retriever` fetches relevant chunks.
6. **Generation:** `LLMManager` sends the query + context to Groq's `llama-3.1-8b-instant` for response generation.
7. **Memory:** `RAGChatbot` manages conversation history for context.

Database Models

Core Models

1. **User** (Extended AbstractUser): Custom user model with role, bio, avatar, and department.
2. **Role**: Hierarchical roles (levels 1-100) for RBAC.
3. **Document**: Central model for files, including access levels, versioning, and counters.
4. **Category**: Hierarchical organization with colors and icons.
5. **Tag**: Many-to-many tagging system.

Supporting Models

6. **DocumentVersion**: Tracks history of document changes.
7. **DocumentComment**: Threaded comments on documents.
8. **SharedLink**: Manages temporary, secure sharing links.
9. **Favorite**: User bookmarks.
10. **ActivityLog**: Comprehensive audit trail.
11. **Notification**: User alerts system.

RAG System Models

12. **ChatSession**: Manages user chatbot sessions.
 13. **ChatMessage**: Stores individual messages and their source citations.
 14. **DocumentEmbedding**: Tracks indexing status and metadata for RAG.
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Permission & Security

Access Levels

- **Public**: Open to all.
- **Private**: Owner only.
- **Role-Based**: Restricted by role hierarchy (e.g., Level 30+).
- **Custom**: Explicitly granted to specific users.

Security Features

- **Authentication**: Django's secure password hashing and session management.
 - **CSRF Protection**: Enabled on all forms.
 - **Input Validation**: Strict file upload validation (size/type) and XSS protection.
 - **SQL Injection**: Prevented via Django ORM.
 - **Audit Logging**: All sensitive actions are logged.
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URL Structure

- **Auth**: `/register/`, `/login/`, `/logout/`

- **Core:** / (Home), /dashboard/
 - **Documents:** /documents/ , /documents/create/ , /documents/<id>/ (view/edit/delete/download)
 - **RAG Chatbot:** /chatbot/
 - **Collaboration:** /documents/<id>/comments/create/ , /documents/<id>/share/
 - **Admin:** /admin/users/ , /admin/roles/
 - **Profile:** /profile/ , /profile/edit/
-

Installation & Setup

Prerequisites

- **Python:** 3.10+
- **RAM:** 8GB+ (16GB recommended for RAG).
- **Groq API Key:** Required for the chatbot functionality.
- **Storage:** ~5GB free space (mostly for dependencies and vector DB).

Quick Start

1. Clone Repository:

```
git clone https://github.com/ft-prince/DocuVault.git
cd DocuVault
```

2. Environment Setup:

```
pip install uv

uv venv
# Windows
venv\Scripts\activate
# Linux/Mac
source venv/bin/activate

uv sync
```

3. Install Dependencies:

```
# Automated
python fix_dependencies.py

# OR Manual
pip install -r requirements.txt
pip install -r requirements_rag.txt

# OR just use uv
uv sync
```

4. Set Environment Variables:

Create a `.env` file in the project root and add your Groq API key:

```
GROQ_API_KEY=your_api_key_here
```

5. Database Migration:

```
python manage.py migrate
```

6. Create Admin:

```
python manage.py createsuperuser
```

7. Run Server:

```
python manage.py runserver
```

- App: <http://127.0.0.1:8000/>
- Admin: <http://127.0.0.1:8000/admin/>
- Chatbot: <http://127.0.0.1:8000/chatbot/>

Windows Automated Setup

Use the provided scripts for a one-click setup:

- PowerShell: `.\setup_windows.ps1`
- CMD: `setup_windows.bat`

RAG Model Setup

The system uses `all-MiniLM-L6-v2` for embeddings, which will be downloaded automatically on the first run (~80MB). The LLM is accessed via the Groq API, so no large local model download is required.

Usage Examples

RAG Chatbot

```
from documents.rag import RAGChatbot, RAGConfig
from documents.rag.document_processor import DocumentProcessor

# Initialize
config = RAGConfig()
chatbot = RAGChatbot(config=config)
chatbot.initialize()

# Index Documents
processor = DocumentProcessor(config=config)
docs = processor.load_documents_from_directory('./my_docs')
chatbot.index_documents(docs)

# Query
answer, sources = chatbot.query("What is the project timeline?")
print(answer)
```

Document Management (Code)

```
# Create a restricted document
doc = Document.objects.create(
    title='Confidential Report',
    file='report.pdf',
    owner=user,
    access_level='role',
    required_role_level=50 # Only for senior roles
)

# Check permissions
if doc.can_view(user):
    print("Access Granted")
```

Configuration Options (RAG)

Modify `documents/rag/config.py` to tune performance:

Parameter	Default	Description
<code>CHUNK_SIZE</code>	512	Characters per chunk.
<code>CHUNK_OVERLAP</code>	100	Overlap between chunks to preserve context.
<code>N_RESULTS</code>	6	Number of chunks to retrieve per query.

Parameter	Default	Description
SIMILARITY_THRESHOLD	0.15	Minimum relevance score for retrieval.
MAX_NEW_TOKENS	512	Maximum length of the LLM response.
TEMPERATURE	0.2	Creativity of the response (0.0-1.0).

Future Enhancements

- ☐ **Multi-language Support:** Expand RAG capabilities to non-English documents.
- ☐ **Document Summarization:** Auto-generate summaries upon upload.
- ☐ **Async Indexing:** Use Celery for background document processing.
- ☐ **Real-time Collaboration:** WebSocket integration for live editing/viewing.
- ☐ **Cloud Storage:** AWS S3 / Google Drive integration.
- ☐ **Hybrid Search:** Combine keyword search with semantic search.
- ☐ **Multi-modal Support:** RAG support for images and tables within PDFs.

Project Implementation Timeline

The project development is structured into distinct phases to ensure a robust and scalable delivery.

Phase	Description	Date	Key Deliverables	Phase Status
Phase 1	Planning & Architecture	Oct 01 - Oct 14, 2025	<ul style="list-style-type: none">✔ Requirement Analysis✔ Database Schema Design✔ Technology Stack Selection✔ UI/UX Wireframes	✔ Completed
Phase 2	Core DMS Development	Oct 15 - Nov 11, 2025	<ul style="list-style-type: none">✔ User Authentication & RBAC✔ Document CRUD Operations✔ File Upload & Storage System✔ Basic Search Functionality	✔ Completed
Phase 3	RAG System Integration	Nov 12 - Nov 30, 2025	<ul style="list-style-type: none">✔ PDF Processing & Chunking Pipeline✔ Vector Database (ChromaDB) Setup✔ LLM Integration (Groq)✔ Chatbot Interface Implementation	✔ Completed

Phase	Description	Date	Key Deliverables	Phase Status
Phase 4	Advanced Features & UI	Dec 01 - Dec 14, 2025	<ul style="list-style-type: none">🔍 Advanced Search & Filtering👥 Collaboration Tools (Comments, Sharing)📊 Dashboard & Analytics🎨 Frontend Styling & Responsiveness	🔄 In Progress
Phase 5	Testing & Optimization	Dec 15 - Dec 21, 2025	<ul style="list-style-type: none">⌚ Unit & Integration Testing⌚ Performance Tuning (RAG Latency)⌚ Security Audits⌚ User Acceptance Testing (UAT)	⌚ Pending
Phase 6	Deployment & Handover	Dec 22 - Dec 28, 2025	<ul style="list-style-type: none">⌚ Production Environment Setup⌚ CI/CD Pipeline Configuration📄 Final Documentation⌚ Project Handover	⌚ Pending

License

This project is open source and available for educational and commercial use.

Contributing

Contributions are welcome! Please submit a Pull Request.