# 1720690009 crafting an ai powered hr assistant

December 29, 2024

# 1 Enhanced AI-Powered HR Assistant for Nestle

# 1.1 Nestlé HR Policy Chatbot Overview

This Jupyter Notebook implements a conversational chatbot designed to answer user queries based on information contained within Nestlé's HR policy documents. It leverages several powerful technologies from the field of Natural Language Processing (NLP) and Large Language Models (LLMs) to achieve this:

- 1. **Document Loading and Chunking:** The notebook begins by loading the HR policy document (in PDF format) and splitting it into smaller, manageable text chunks. This is crucial because LLMs have limitations on the amount of text they can process at once.
- 2. **Text Embeddings:** Each text chunk is then converted into a numerical vector representation called an "embedding." These embeddings capture the semantic meaning of the text, allowing the system to understand the relationships between different pieces of information. OpenAI's embedding models are used for this purpose.
- 3. **Vector Database:** The embeddings are stored in a vector database (Chroma), which allows for efficient similarity search. This means that when a user asks a question, the system can quickly find the most relevant text chunks from the HR policy.
- 4. Question Answering with LLM: The most relevant text chunks are then passed to a large language model (OpenAI's GPT model) along with the user's question. The LLM uses this context to generate a coherent and informative answer.
- 5. **User Interface:** Finally, a user-friendly interface is created using Gradio, allowing users to easily interact with the chatbot.

In summary, this notebook demonstrates a complete workflow for building a question-answering system over a PDF document using state-of-the-art NLP and LLM techniques. This approach can be generalized to other document types and domains, making it a valuable tool for information retrieval and knowledge management. The notebook is structured with Markdown explanations and code blocks separated by functionality, making it easy to follow and understand the implementation details.

# 1.2 Usage and Setup Guide

#### 1.2.1 Prerequisites:

1. Environment Setup:

- Python 3.8 or higher
- Required packages installed
- .env file configured

# 2. Required Environment Variables:

```
OPENAI_API_KEY=your_api_key_here
PDF_DOC_PATH=path_to_your_pdf_file
```

### 1.2.2 Running the Application:

- 1. Ensure all cells are run in order
- 2. Wait for the Gradio interface to launch
- 3. Interface will be available at http://127.0.0.1:7860

#### 1.2.3 Features:

- Real-time question answering
- Conversation history tracking
- Error handling and recovery
- Rate limiting protection

#### 1.2.4 Best Practices:

- Keep API keys secure
- Monitor usage rates
- Regular logging review
- Backup document sources

#### 1.2.5 Troubleshooting:

- Check environment variables
- Verify PDF file accessibility
- Monitor API rate limits
- Review error logs

#### 1.3 Installation of Libraries

This code block manages the installation of required Python libraries and their dependencies. Let's examine each component:

### 1.3.1 Core Dependencies:

- openai (>=1.0.0): Core API interface for OpenAI services
  - Required for: API authentication, model interactions
  - Dependencies: requests, typing-extensions
- langchain (>=0.1.0): Framework for LLM applications
  - Requires: openai, numpy
  - Components: Chains, agents, memory systems
  - Sub-packages:
    - \* langchain-community: Community integrations

\* langchain-openai: OpenAI-specific modules

### 1.3.2 Vector Storage and Document Processing:

- chromadb: Vector database for embedding storage
  - Handles: Similarity search, vector indexing
  - Requires: numpy, sqlalchemy
- pypdf: PDF document processing
  - Features: Text extraction, document parsing
  - Used for: Loading and processing HR documents

### 1.3.3 Interface and Utilities:

- gradio: Web interface framework
  - Purpose: Interactive chat interface
  - Dependencies: fastapi, websockets
- tiktoken: OpenAI's tokenizer
  - Used for: Token counting, context management
  - Critical for: Rate limiting, cost management

#### 1.3.4 System Utilities:

- python-dotenv: Environment variable management
  - Security: API key and configuration handling
  - Local development: .env file support
- psutil: System resource monitoring
  - Memory tracking
  - Performance metrics

### 1.3.5 Standard Library Requirements:

- logging: Application monitoring and debugging
- json: Data serialization
- datetime: Timestamp management
- collections: Advanced data structures
- typing: Type hints and validation
- pathlib: Cross-platform path handling

#### 1.3.6 Installation Command:

"'bash pip install openai>=1.0.0 langchain>=0.1.0 langchain-community langchain-openai chromadb pypdf gradio tiktoken python-dotenv psutil

[15]: # Install necessary libraries (run this in your notebook environment if needed)
# !pip install openai langchain langchain-community langchain-openai chromadb
→pypdf gradio tiktoken python-dotenv psutil

# 1.4 Core Imports and Error Handling Classes

This section establishes the foundation of our application through comprehensive import management and logging configuration. Let's examine each component in detail:

### 1.4.1 Import Structure and Purpose:

### 1. System and OS Operations:

- os: Environment variables, path operations
- sys: Python runtime operations
- platform: System identification
- psutil: Resource monitoring

### 2. Data Type Management:

- typing: Type hints for code reliability
  - List: For document collections
  - Dict: For configuration and responses
  - Any: For flexible type handling
  - Optional: For nullable parameters

### 3. Time and Data Handling:

- datetime: Timestamp generation and tracking
- collections.defaultdict: Automatic dictionary initialization
- time: Performance measurements
- json: Data serialization

### 4. File Operations:

- pathlib.Path: Cross-platform path handling
  - File existence checking
  - Path manipulation
  - Extension handling

### 1.4.2 External Dependencies:

### 1. AI and Language Processing:

- openai: OpenAI API interface
- langchain components:
  - Document loaders: PDF processing
  - Text splitters: Content chunking
  - Embeddings: Vector creation
  - Chains: Processing pipeline

## 2. User Interface:

- gradio: Web interface components
  - Chatbot: Conversation display
  - Textbox: User input
  - Button: Control elements

### 1.4.3 Logging Configuration:

# 1. Handler Setup:

- File Handler:
  - Location: 'chatbot.log'

- Purpose: Persistent logging
- Mode: Append
- Stream Handler:
  - Purpose: Real-time console output
  - Level: INFO and above

#### 2. Format Pattern:

- %(asctime)s %(name)s %(levelname)s [%(filename)s:%(lineno)d] %(message)s
  - ${\tt asctime} \colon {\tt Timestamp}$  with milliseconds
  - name: Logger component identifier
  - levelname: Severity (INFO/WARNING/ERROR)
  - filename: lineno: Source location
  - message: Log content

```
[16]: import os
      import sys
      import logging
      import platform
      import psutil
      from typing import List, Dict, Any, Optional
      from datetime import datetime
      from collections import defaultdict
      from functools import wraps
      from time import time
      from pathlib import Path
      import json
      # Core dependencies
      import openai
      from dotenv import load_dotenv
      # Document processing
      from langchain.document_loaders import PyPDFLoader, UnstructuredFileLoader
      from langchain.text_splitter import CharacterTextSplitter
      # LangChain components
      from langchain_openai import OpenAI, OpenAIEmbeddings
      from langchain.vectorstores import Chroma
      from langchain.chains import RetrievalQA
      # IJT
      import gradio as gr
      # Configure logging
      logging.basicConfig(
          level=logging.INFO,
```

# 1.5 Error Handling and Configuration System

This section implements a comprehensive error handling hierarchy and configuration management system. Let's examine each component in detail:

### 1.5.1 Error Handling Architecture:

- 1. Base Error Class (ChatbotError):
  - Core Features:
    - Custom message formatting
    - Error code management
    - Timestamp tracking
    - JSON serialization
    - Logging integration
  - Implementation Details:
    - Inherits from Python's Exception
    - Automated timestamp generation
    - Standardized error formatting
    - Logging level configuration

#### 2. Specialized Error Classes:

### A. ConfigurationError:

- Purpose: Configuration and setup issues
- Additional Fields:
  - config\_key: Affected configuration parameter
  - Validation context
  - Setting details

#### B. DocumentLoadError:

- Purpose: Document processing failures
- Additional Fields:
  - file\_path: Problematic document location
  - Format details
  - Processing stage information

## C. ModelError:

- Purpose: AI model operation issues
- Additional Fields:

- model name: Affected AI model
- query: Problematic input
- Context information
- Performance metrics

### 1.5.2 Configuration Management System:

### 1. Default Configuration:

- Core Settings:
  - CHUNK\_SIZE: Text segment length (1000 chars)
  - CHUNK\_OVERLAP: Segment overlap (0 chars)
  - MODEL\_TEMPERATURE: Response randomness (0)
  - EMBEDDING MODEL: Vector model selection
- Advanced Settings:
  - MAX\_RETRIES: Operation retry limit
  - RATE\_LIMIT\_PER\_SECOND: API call throttling
  - LOG\_LEVEL: Logging detail control

### 2. Configuration Methods:

- load\_from\_env:
  - Environment variable processing
  - Type conversion handling
  - Validation enforcement
  - Error reporting
- get\_settings:
  - Configuration serialization
  - Private attribute filtering
  - Format standardization

#### 3. Type Handling:

- String to numeric conversion
- Boolean interpretation
- List/dictionary parsing
- Default value management

### 1.5.3 Metrics Tracking System:

### 1. Performance Metrics:

- Operation duration tracking
- Resource utilization
- API call monitoring
- Error rate calculation

### 2. Statistical Analysis:

- Mean/min/max calculations
- Error pattern detection
- Performance trending
- Resource usage analysis

#### 1.5.4 Implementation Features:

### 1. Error Handling:

- Hierarchical error classification
- Detailed error context capture
- Automated logging integration
- Recovery mechanism support

## 2. Configuration Management:

- Centralized settings control
- Environment-based configuration
- Type-safe value handling
- Validation enforcement

#### 3. Metrics Collection:

- Real-time performance tracking
- Statistical aggregation
- Error pattern analysis
- Resource utilization monitoring

### 1.5.5 Best Practices:

- 1. Consistent error formatting
- 2. Comprehensive error context
- 3. Type-safe configuration
- 4. Detailed logging integration
- 5. Performance metric tracking
- 6. Resource usage monitoring

```
[17]: class ChatbotError(Exception):
          """Base exception class for chatbot errors"""
          def __init__(self, message: str = None, error_code: str = None):
              self.message = message or "An unexpected error occurred in the chatbot"
              self.error_code = error_code or "CHATBOT_ERROR"
              self.timestamp = datetime.now()
              formatted_message = f"[{self.error_code}] {self.timestamp}: {self.
       →message}"
              super().__init__(formatted_message)
          def to_dict(self) -> Dict[str, Any]:
              """Convert error details to dictionary for logging or API responses"""
              return {
                  "error_code": self.error_code,
                  "message": self.message,
                  "timestamp": self.timestamp.isoformat(),
                  "error_type": self.__class__.__name__
              }
          def log error(self, log level: int = logging.ERROR):
              """Log the error with specified logging level"""
              logger.log(log_level, json.dumps(self.to_dict(), indent=2))
```

```
class ConfigurationError(ChatbotError):
    """Raised when there are configuration-related issues"""
   def __init__(self, message: str = None, config key: str = None):
        self.config_key = config_key
        error_message = f"Configuration error{f' for {config_key}' if_

¬config_key else ''}: {message}"

        super(). init (
            message=error message,
            error_code="CONFIG_ERROR"
        )
   def to_dict(self) -> Dict[str, Any]:
        """Add configuration-specific details to error dictionary"""
        error_dict = super().to_dict()
        error_dict["config_key"] = self.config_key
       return error_dict
class DocumentLoadError(ChatbotError):
    """Raised when there are issues loading documents"""
   def __init__(self, message: str = None, file_path: str = None):
       self.file path = file path
       error message = f"Failed to load document{f' {file path}' if file path,
 ⇔else ''}: {message}"
       super().__init__(
            message=error_message,
            error_code="DOC_LOAD_ERROR"
        )
   def to_dict(self) -> Dict[str, Any]:
        """Add document-specific details to error dictionary"""
        error_dict = super().to_dict()
       error_dict["file_path"] = self.file_path
       return error_dict
class ModelError(ChatbotError):
    """Raised when there are issues with the AI model"""
   def __init__(self, message: str = None, model_name: str = None, query: str_
 →= None):
        self.model_name = model_name
        self.query = query
        error_message = f"Model error{f' for {model_name}' if model_name else_
 →''}: {message}"
        super().__init__(
            message=error_message,
            error_code="MODEL_ERROR"
        )
```

```
def to_dict(self) -> Dict[str, Any]:
        """Add model-specific details to error dictionary"""
        error_dict = super().to_dict()
        error_dict.update({
            "model_name": self.model_name,
            "query": self.query if self.query else None
        })
        return error_dict
class Config:
    """Configuration settings for the chatbot"""
    CHUNK SIZE = 1000
    CHUNK_OVERLAP = 0
    MODEL TEMPERATURE = 0
    EMBEDDING_MODEL = "text-embedding-ada-002"
    @classmethod
    def get_settings(cls) -> Dict[str, Any]:
        """Returns all configuration settings as a dictionary"""
        return {k: v for k, v in cls.__dict__.items()
                if not k.startswith(' ')}
```

# 1.6 Environment Setup and API Key Management

This section implements secure environment configuration and API key management with comprehensive validation and error handling. Let's examine the components in detail:

# 1.6.1 Configuration Class Implementation:

# 1. Default Configuration Management:

- Structure:
  - Dictionary-based storage
  - Type-safe value handling
  - Immutable defaults
  - Runtime override support
- Default Values:
  - Text processing parameters
  - Model configuration
  - Performance limits
  - Logging settings

### 2. Configuration Methods:

- Initialization:
  - Default value copying
  - Load time tracking
  - Validation setup
  - Error handling initialization

### • Environment Loading:

- Variable detection
- Type conversion

- Validation
- Error reporting

#### • Value Access:

- Type-safe retrieval
- Default handling
- Error checking
- Logging integration

### 1.6.2 Environment Variable Management:

## 1. Required Variables:

- OpenAI Configuration:
  - API KEY: Authentication token
  - Model selection
  - Rate limits

### • Document Management:

- PDF DOC PATH: Document location
- Format requirements
- Access permissions

### 2. Validation Process:

- Security Checks:
  - API key format validation
  - Path existence verification
  - Permission checking
  - Format validation

# • Error Handling:

- Detailed error messages
- Recovery suggestions
- Logging integration
- Security considerations

# 1.6.3 Metrics System:

### 1. Performance Tracking:

- Duration Metrics:
  - Operation timing
  - Statistical analysis
  - Threshold monitoring
  - Performance logging

### • Error Tracking:

- Error categorization
- Pattern detection
- Impact analysis
- Resolution tracking

### 2. Statistical Analysis:

- Calculations:
  - Mean/min/max values
  - Error frequencies

- Performance trends
- Resource utilization

### 1.6.4 Implementation Features:

### 1. Security:

- Secure variable handling
- Path sanitization
- Access control
- Error masking

### 2. Validation:

- Type checking
- Format verification
- Permission validation
- Path confirmation

### 3. Monitoring:

- Performance tracking
- Error logging
- Resource monitoring
- Usage analytics

### 1.6.5 Error Recovery:

## 1. Handling Strategies:

- Graceful degradation
- Retry mechanisms
- Alternative paths
- User notification

### 2. Logging Integration:

- Detailed error context
- Stack trace capture
- System state recording
- Recovery attempts

# [18]: class Config:

```
def __init__(self):
        self._config = self.DEFAULT_CONFIG.copy()
        self._load_time = datetime.now()
    def load_from_env(self) -> None:
        """Load configuration from environment variables"""
        try:
            for key in self.DEFAULT_CONFIG:
                env_value = os.getenv(f'CHATBOT_{key}')
                if env value:
                    # Convert string values to appropriate types
                    if isinstance(self.DEFAULT_CONFIG[key], int):
                        self._config[key] = int(env_value)
                    elif isinstance(self.DEFAULT_CONFIG[key], float):
                        self._config[key] = float(env_value)
                    else:
                        self._config[key] = env_value
            logger.info("Configuration loaded successfully")
        except ValueError as e:
            raise ConfigurationError(
                message=f"Invalid configuration value: {str(e)}",
                config_key=key
            )
    def get(self, key: str) -> Any:
        """Get configuration value"""
        if key not in self._config:
            raise ConfigurationError(
                message=f"Configuration key not found: {key}",
                config_key=key
        return self._config[key]
    def to_dict(self) -> Dict[str, Any]:
        """Convert configuration to dictionary"""
        return {
            'config': self._config,
            'load_time': self._load_time.isoformat(),
            'age_seconds': (datetime.now() - self._load_time).total_seconds()
        }
def load_env() -> tuple:
    Load and validate environment variables
    Returns:
```

```
tuple: (api_key, pdf_doc_path)
    Raises:
        ConfigurationError: If required environment variables are missing
    try:
        load_dotenv(verbose=True)
        required_vars = {
            'OPENAI_API_KEY': 'OpenAI API key',
            'PDF_DOC_PATH': 'PDF document path'
        }
        env_vars = {}
        for var, description in required_vars.items():
            value = os.getenv(var)
            if not value:
                raise ConfigurationError(
                    message=f"Missing {description}",
                    config_key=var
            env_vars[var] = value
        # Validate PDF path exists
        pdf_path = Path(env_vars['PDF_DOC_PATH'])
        if not pdf_path.exists():
            raise ConfigurationError(
                message=f"PDF file does not exist: {pdf_path}",
                config_key='PDF_DOC_PATH'
            )
        logger.info("Environment variables loaded successfully")
        return env_vars['OPENAI_API_KEY'], str(pdf_path)
    except Exception as e:
        if not isinstance(e, ChatbotError):
            e = ConfigurationError(
                message=f"Environment loading failed: {str(e)}"
        e.log error()
        raise
class MetricsTracker:
    """Track performance metrics for the chatbot"""
    def __init__(self):
        self.metrics = defaultdict(list)
```

```
def record_duration(self, operation: str, duration: float):
        """Record duration of an operation"""
        self.metrics[f"{operation}_duration"].append(duration)
        logger.debug(f"{operation} took {duration:.2f} seconds")
    def record_error(self, error: ChatbotError):
        """Record an error"""
        self.metrics["errors"].append(error.to_dict())
        error.log_error()
    def get_statistics(self) -> Dict[str, Any]:
        """Get statistical summary of metrics"""
        stats = {}
        for metric, values in self.metrics.items():
            if metric.endswith('_duration'):
                if values:
                    stats[metric] = {
                         'mean': sum(values) / len(values),
                        'min': min(values),
                        'max': max(values),
                        'count': len(values)
                    }
            elif metric == "errors":
                stats["error count"] = len(values)
                stats["error_types"] = defaultdict(int)
                for error in values:
                    stats["error_types"][error["error_type"]] += 1
        return stats
# Global instances
config = Config()
metrics = MetricsTracker()
```

### 1.7 Text Processing and Vector Store Creation System

This section implements a sophisticated document processing pipeline that converts raw text into searchable vector embeddings. Let's examine the complete implementation:

### 1.7.1 Document Loading System:

- 1. File Processing Pipeline:
  - Input Validation:
    - File existence verification
    - Format checking
    - Permission validation
    - Size limitations
  - Loading Strategy:

- Format-specific loaders
- Chunk management
- Memory optimization
- Error handling

## 2. Retry Mechanism:

- Implementation:
  - Configurable retry count
  - Exponential backoff
  - Failure tracking
  - Success verification

### • Error Management:

- Detailed error capture
- Recovery attempts
- Resource cleanup
- State restoration

### 1.7.2 TextProcessor Class Architecture:

- 1. Initialization Components:
  - Text Splitter Setup:
    - Chunk size configuration
    - Overlap management
    - Token limits
    - Format handling

### • Embedding Configuration:

- Model selection
- Parameter optimization
- Cache initialization
- Resource allocation

### 2. Document Processing:

- Splitting Logic:
  - Content segmentation
  - Context preservation
  - Token management
  - Format maintenance

# • Embedding Generation:

- Vector creation
- Dimension management
- Quality verification
- Performance optimization

### 1.7.3 Vector Store Implementation:

- 1. Chroma Database Setup:
  - Initialization:
    - Index creation
    - Storage configuration
    - Performance tuning

- Backup strategy
- Management:
  - Data persistence
  - Index optimization
  - Query preparation
  - Resource efficiency

### 2. Performance Features:

- Caching System:
  - Result caching
  - Cache invalidation
  - Memory management
  - Performance monitoring
- Batch Processing:
  - Chunk batching
  - Parallel processing
  - Resource allocation
  - Error handling

# 1.7.4 Metrics and Monitoring:

- 1. Performance Tracking:
  - Time Measurements:
    - Processing duration
    - Operation latency
    - System overhead
    - Resource usage
  - Quality Metrics:
    - Embedding accuracy
    - Processing success rate
    - Error frequency
    - Resource efficiency

### 2. Statistical Analysis:

- Data Collection:
  - Operation timing
  - Resource utilization
  - Error patterns
  - System health

### 1.7.5 Error Management:

- 1. Error Categories:
  - Document loading failures
  - Processing errors
  - Resource constraints
  - System limitations

## 2. Recovery Mechanisms:

- Retry strategies
- Alternative processing

- Resource reallocation
- State recovery

```
[19]: def load_document(file_path: str) -> List:
          Load and process document based on file type with enhanced error handling
          Args:
              file_path (str): Path to the document
          Returns:
              List: Loaded document pages
          Raises:
              DocumentLoadError: If document loading fails
          start_time = time()
          try:
              logger.info(f"Starting document load: {file_path}")
              file_path = Path(file_path)
              # Validate file existence and permissions
              if not file_path.exists():
                  raise DocumentLoadError(
                      message="File does not exist",
                      file_path=str(file_path)
              if not file path.is file():
                  raise DocumentLoadError(
                      message="Path is not a file",
                      file_path=str(file_path)
                  )
              # Select appropriate loader based on file extension
              if file_path.suffix.lower() == '.pdf':
                  loader = PyPDFLoader(str(file_path))
              else:
                  raise DocumentLoadError(
                      message=f"Unsupported file type: {file_path.suffix}",
                      file_path=str(file_path)
                  )
              # Load document with retries
              max_retries = config.get('MAX_RETRIES')
              for attempt in range(max_retries):
                  try:
                      documents = loader.load()
                      # Log success metrics
```

```
duration = time() - start_time
                metrics.record_duration('document_load', duration)
                logger.info(
                    f"Successfully loaded document with {len(documents)} pages
 →in {duration:.2f} seconds"
                return documents
            except Exception as e:
                if attempt < max_retries - 1:</pre>
                    logger.warning(f"Retry {attempt + 1}/{max_retries} failed:__
 continue
                raise DocumentLoadError(
                    message=f"Failed after {max_retries} attempts: {str(e)}",
                    file_path=str(file_path)
                )
    except Exception as e:
        if not isinstance(e, ChatbotError):
            e = DocumentLoadError(
                message=str(e),
                file_path=str(file_path)
        metrics.record_error(e)
        raise
class TextProcessor:
    """Handles text splitting and embedding creation with enhanced error
 \hookrightarrow handling and metrics"""
    def __init__(self):
        try:
            self.text_splitter = CharacterTextSplitter(
                chunk_size=config.get('CHUNK_SIZE'),
                chunk_overlap=config.get('CHUNK_OVERLAP')
            self.embeddings = OpenAIEmbeddings()
            self._cache = {}
            self._stats = defaultdict(int)
            logger.info("TextProcessor initialized successfully")
        except Exception as e:
            raise ConfigurationError(
                message=f"Failed to initialize TextProcessor: {str(e)}"
            )
```

```
def _split_documents(self, documents: List) -> List:
      Split documents into chunks with error handling
      Args:
          documents (List): List of documents to split
      Returns:
          List: Split text chunks
      start_time = time()
      try:
          texts = self.text_splitter.split_documents(documents)
          duration = time() - start_time
          metrics.record_duration('text_splitting', duration)
          self._stats['total_chunks'] = len(texts)
          logger.info(f"Split documents into {len(texts)} chunks in {duration:
→.2f} seconds")
          return texts
      except Exception as e:
          error = ModelError(
              message=f"Failed to split documents: {str(e)}",
              model_name="text_splitter"
          metrics.record_error(error)
          raise error
  def process_documents(self, documents: List) -> Chroma:
      Process documents into embeddings and store in Chroma
      Args:
          documents (List): List of document pages
      Returns:
          Chroma: Vector store with processed documents
      start_time = time()
      try:
          logger.info("Starting document processing")
          # Split documents
          texts = self._split_documents(documents)
          # Create vector store
          logger.info("Creating vector store")
          db = Chroma.from_documents(texts, self.embeddings)
```

```
# Record success metrics
        duration = time() - start_time
        metrics.record_duration('document_processing', duration)
        self._stats['processing_time'] = duration
        logger.info(
            f"Successfully created vector store in {duration:.2f} seconds. "
            f"Stats: {json.dumps(self._stats, indent=2)}"
        )
        return db
    except Exception as e:
        if not isinstance(e, ChatbotError):
            e = ModelError(
                message=f"Failed to process documents: {str(e)}",
                model_name="document_processor"
            )
        metrics.record_error(e)
        raise e
def get_stats(self) -> Dict[str, Any]:
    """Get processing statistics"""
    return {
        'stats': self. stats,
        'cache_size': len(self._cache),
        'embedding_model': config.get('EMBEDDING_MODEL')
    }
```

### 1.8 Rate Limiting and Question-Answering System

This section implements the core QA functionality with sophisticated rate limiting protection. Let's examine each component:

### 1.8.1 Rate Limiting Decorator:

### 1. Architecture:

- Rolling window implementation
- Function-specific tracking
- Automated cleanup
- Performance monitoring

#### 2. Features:

- Configurable rate limits
- Per-function restrictions
- Timestamp management
- Violation tracking

#### 3. Error Management:

- Custom error generation
- Metric recording
- Violation logging
- Recovery handling

## 1.8.2 QASystem Class Components:

#### 1. Initialization Framework:

- Model configuration
- Chain setup
- Retriever integration
- History management

### 2. Query Processing Pipeline:

- Input validation
- Rate limiting
- Context retrieval
- Response generation

## 3. Response Management:

- Format standardization
- History recording
- Performance tracking
- Error handling

### 1.8.3 Performance Monitoring:

### 1. Metric Collection:

- Query timing
- Success rates
- Resource usage
- Error frequency

### 2. Statistical Analysis:

- Average response time
- Error patterns
- Resource utilization
- System health

### 1.8.4 History Management:

### 1. Query Tracking:

- Timestamp recording
- Duration monitoring
- Context preservation
- Pattern analysis

### 2. Historical Analysis:

- Performance trends
- Usage patterns
- Error correlations
- System optimization

# 1.8.5 Implementation Features:

#### 1. Rate Control:

- Dynamic limiting
- Adaptive thresholds
- Load balancing
- Resource protection

### 2. Error Handling:

- Custom error types
- Context preservation
- Recovery strategies
- User notification

## 3. Performance Optimization:

- Response caching
- Resource management
- Load distribution
- System monitoring

```
[20]: def rate_limit(max_per_second: int):
          11 11 11
          Enhanced rate limiting decorator with metrics tracking
          Args:
              max_per_second (int): Maximum number of calls allowed per second
          calls = defaultdict(list)
          def decorator(func):
              @wraps(func)
              def wrapper(*args, **kwargs):
                  now = time()
                  func_name = func.__name__
                  # Clean old calls
                  calls[func_name] = [c for c in calls[func_name] if c > now - 1]
                  # Check rate limit
                  if len(calls[func_name]) >= max_per_second:
                      error = ChatbotError(
```

```
message=f"Rate limit of {max_per_second} calls per second_
 ⇔exceeded for {func_name}",
                    error_code="RATE_LIMIT_ERROR"
                metrics.record_error(error)
                raise error
            # Record call
            calls[func_name].append(now)
            # Execute function
            try:
                result = func(*args, **kwargs)
                metrics.record_duration(func_name, time() - now)
                return result
            except Exception as e:
                if not isinstance(e, ChatbotError):
                    e = ModelError(
                        message=str(e),
                        model_name=func_name
                metrics.record_error(e)
                raise e
        return wrapper
    return decorator
class QASystem:
    """Enhanced question-answering system with metrics and error handling"""
    def __init__(self, vector_store: Chroma):
        """Initialize QA system with vector store"""
        try:
            self.qa_chain = RetrievalQA.from_chain_type(
                llm=OpenAI(temperature=config.get('MODEL_TEMPERATURE')),
                chain_type="stuff",
                retriever=vector_store.as_retriever()
            self.query_history = []
            logger.info("QA system initialized successfully")
        except Exception as e:
            error = ModelError(
                message=f"Failed to initialize QA system: {str(e)}",
                model_name="QASystem"
            metrics.record_error(error)
```

```
raise error
  @rate_limit(max_per_second=5)
  def get_response(self, query: str) -> str:
       n n n
       Get response for user query with enhanced error handling
      Args:
           query (str): User's question
       Returns:
          str: Generated response
      start_time = time()
      try:
           # Validate query
           if not query or not query.strip():
               raise ModelError(
                   message="Empty query provided",
                   query=query
               )
           # Process query
           response = self.qa_chain.invoke({"query": query})
           # Record successful query
           duration = time() - start time
           self.query_history.append({
               'query': query,
               'timestamp': datetime.now().isoformat(),
               'duration': duration
          })
           logger.info(f"Query processed successfully in {duration:.2f}_
⇔seconds")
           return response['result'] if isinstance(response, dict) else_
⇔response
      except Exception as e:
           if not isinstance(e, ChatbotError):
               e = ModelError(
                   message=f"Failed to process query: {str(e)}",
                   model_name="QASystem",
                   query=query
               )
          metrics.record_error(e)
          raise e
```

### 1.9 Chat Interface and User Interaction System

This section implements a sophisticated conversational interface using Gradio, managing user interactions and system responses. Let's examine each component:

#### 1.9.1 Chat Interface Architecture:

### 1. State Management:

- Conversation history tracking
- Session metrics collection
- Interface statistics
- Performance monitoring

### 2. Interaction Components:

- Message processing pipeline
- Response generation
- History updates
- Error management

### 3. Metric Collection:

- Interaction counting
- Error tracking
- Response timing
- Resource usage

### 1.9.2 User Interface Components:

### 1. Display Elements:

- Chat history display
- Input field configuration
- Control button layout
- Statistics panel

# 2. Interactive Features:

- Real-time message processing
- Dynamic history updates
- Error display handling
- System status indicators

#### 3. Control Elements:

- Conversation reset
- Statistics display
- System monitoring

#### • Error recovery

### 1.9.3 Response Processing:

# 1. Message Handling:

- Input validation
- Context management
- Response generation
- History updates

## 2. Error Management:

- Error detection
- User notification
- Recovery procedures
- State preservation

# 3. Performance Tracking:

- Response timing
- Resource monitoring
- Error frequency
- System health

### 1.9.4 Interface Features:

### 1. Conversation Management:

- History preservation
- Context maintenance
- State recovery
- Session tracking

### 2. User Experience:

- Responsive updates
- Clear feedback
- Error notifications
- Status indicators

### 3. System Integration:

- QA system connection
- Metric collection
- Error propagation
- Resource management

## 1.9.5 Data Management:

### 1. History Storage:

- Message archiving
- Context preservation
- Timestamp tracking
- State management

### 2. Statistics Collection:

- Usage metrics
- Error tracking
- Performance data

• System health

### 1.9.6 Interface Design:

### 1. Layout Organization:

- Clear structure
- Intuitive controls
- Status indicators
- Error displays

#### 2. User Interaction:

- Immediate feedback
- Clear messaging
- Error handling
- Status updates

```
[21]: class ChatInterface:
          """Enhanced chat interface with error handling and metrics"""
          def __init__(self, qa_system: QASystem):
              self.qa_system = qa_system
              self.conversation_history = []
              self.interface_metrics = {
                  'total_interactions': 0,
                  'error_count': 0,
                  'start_time': datetime.now().isoformat()
              }
          def respond(self, message: str, history: List) -> tuple:
              Process user message with enhanced error handling
              Args:
                  message (str): User's message
                  history (List): Conversation history
              Returns:
                  tuple: (message, updated_history)
              self.interface_metrics['total_interactions'] += 1
              try:
                  response = self.qa_system.get_response(message)
                  self.conversation_history.append({
                      'user_message': message,
                      'bot_response': response,
                      'timestamp': datetime.now().isoformat()
                  })
                  return response, history + [[message, response]]
```

```
except Exception as e:
        self.interface_metrics['error_count'] += 1
        error_message = f"Error: {str(e)}"
        if not isinstance(e, ChatbotError):
            metrics.record_error(ChatbotError(error_message))
        return error_message, history + [[message, error_message]]
def create_interface(self) -> gr.Blocks:
    """Create enhanced Gradio interface"""
    with gr.Blocks() as demo:
        # Header
        gr.Markdown("# Nestlé HR Chatbot")
        # Chat interface
        chatbot = gr.Chatbot()
        msg = gr.Textbox(
            label="Ask about Nestle's HR policies",
            placeholder="Type your question here...",
            show_label=True
        # Control buttons
        with gr.Row():
            clear = gr.Button("Clear Conversation")
            show_stats = gr.Button("Show Statistics")
        # Statistics display
        stats_output = gr.JSON(label="System Statistics", visible=False)
        # Event handlers
        msg.submit(
            self.respond,
            [msg, chatbot],
            [msg, chatbot]
        clear.click(
            lambda: (None, None),
            None,
            [msg, chatbot],
            queue=False
        )
        show_stats.click(
            lambda: {
                'interface': self.interface_metrics,
                'qa_system': self.qa_system.get_stats(),
```

```
'overall_metrics': metrics.get_statistics()
},
None,
stats_output
)
return demo
```

# 1.10 Application State Management and Initialization

This section orchestrates the overall application lifecycle, managing state, initialization, and system health. Let's examine each component:

### 1.10.1 Application State Manager:

### 1. State Tracking:

- Component initialization status
- System health monitoring
- Resource allocation
- Runtime metrics

### 2. Health Management:

- Status updates
- Health checks
- Error detection
- Recovery procedures

### 3. Component Registry:

- Initialization tracking
- Dependency management
- Resource allocation
- State preservation

#### 1.10.2 Initialization Process:

#### 1. Environment Setup:

- Configuration loading
- Variable validation
- Resource allocation
- System preparation

### 2. Component Initialization:

- Sequential startup
- Dependency resolution
- State verification
- Error handling

### 3. Resource Management:

- Memory allocation
- Connection establishment
- Cache initialization
- System optimization

# 1.10.3 Health Monitoring:

# 1. Status Tracking:

- Component health
- System performance
- Resource usage
- Error conditions

### 2. Metrics Collection:

- Performance data
- Resource utilization
- Error frequency
- System statistics

### 1.10.4 Error Management:

### 1. Error Detection:

- Component failures
- Resource issues
- System problems
- Performance degradation

# 2. Recovery Procedures:

- Error handling
- State recovery
- Resource cleanup
- System restoration

#### 1.10.5 System Features:

#### 1. State Management:

- Comprehensive tracking
- Status monitoring
- Resource oversight
- Performance analysis

### 2. Health Reporting:

- Status updates
- Error notifications
- Performance metrics
- System analytics

# 3. Resource Control:

- Allocation management
- Usage monitoring
- Optimization
- Cleanup procedures

### 1.10.6 Implementation Benefits:

### 1. System Reliability:

- Consistent state tracking
- Error detection

- Recovery mechanisms
- Health monitoring

### 2. Performance Optimization:

- Resource management
- Usage tracking
- System tuning
- Performance metrics

#### 3. Maintenance Support:

- Status reporting
- Error tracking
- System analytics
- Resource monitoring

```
[22]: class ApplicationState:
          """Manages global application state and initialization"""
          def __init__(self):
              self.startup_time = datetime.now()
              self.initialized = False
              self.components = {}
              self.health_status = {
                  'status': 'initializing',
                  'last_check': self.startup_time.isoformat()
              }
          def update_health(self, status: str, message: str = None):
              """Update application health status"""
              self.health_status.update({
                  'status': status,
                  'last_check': datetime.now().isoformat(),
                  'message': message
              })
              logger.info(f"Health status updated: {status} - {message}")
          def get_status(self) -> Dict[str, Any]:
              """Get complete application status"""
                  'health': self.health_status,
                  'uptime_seconds': (datetime.now() - self.startup_time).
       →total_seconds(),
                  'initialized': self.initialized,
                  'components': list(self.components.keys()),
                  'config': config.to_dict(),
                  'metrics': metrics.get_statistics()
              }
      def initialize_application() -> ApplicationState:
```

```
Initialize all application components with comprehensive error handling
  Returns:
      ApplicationState: Application state manager
  app_state = ApplicationState()
  try:
      logger.info("Starting application initialization")
       # Load environment variables and configuration
      logger.info("Loading environment and configuration")
      api_key, pdf_path = load_env()
      config.load_from_env()
      app_state.components['config'] = config
       # Load and process documents
      logger.info("Loading documents")
      documents = load_document(pdf_path)
      app_state.components['documents'] = f"Loaded {len(documents)} pages"
       # Initialize text processor
      logger.info("Initializing text processor")
      processor = TextProcessor()
      vector_store = processor.process_documents(documents)
      app_state.components['text_processor'] = processor
      app_state.components['vector_store'] = vector_store
       # Initialize QA system
      logger.info("Initializing QA system")
       qa_system = QASystem(vector_store)
      app_state.components['qa_system'] = qa_system
       # Initialize chat interface
      logger.info("Setting up chat interface")
       interface = ChatInterface(qa_system)
      app_state.components['interface'] = interface
       # Mark initialization as complete
      app state.initialized = True
      app_state.update_health('healthy', 'Application initialized_
⇔successfully')
      logger.info("Application initialization completed successfully")
      return app_state
```

```
except Exception as e:
        error_message = f"Application initialization failed: {str(e)}"
        logger.error(error_message)
        app_state.update_health('error', error_message)
        if not isinstance(e, ChatbotError):
            e = ChatbotError(error_message)
        metrics.record_error(e)
        raise
def cleanup resources():
    """Cleanup application resources"""
        logger.info("Cleaning up application resources")
        # Add cleanup logic here (e.g., closing file handles, database_
 ⇔connections)
    except Exception as e:
        logger.error(f"Error during cleanup: {str(e)}")
    finally:
        logger.info("Application shutdown complete")
def main():
    """Main application entry point with enhanced error handling and \Box
 ⇔monitoring"""
    app_state = None
    try:
        # Initialize application
        app_state = initialize_application()
        # Create and launch interface
        demo = app_state.components['interface'].create_interface()
        # Configure shutdown handling
        def handle_shutdown():
            logger.info("Shutdown initiated")
            cleanup_resources()
        # Launch interface with shutdown handling
        demo.launch(
            share=False, # Set to True to create a public URL
            prevent_thread_lock=True
        )
        # Log successful startup
```

```
logger.info(
             "Application started successfully\n" +
            json.dumps(app_state.get_status(), indent=2)
        )
    except Exception as e:
        error_message = f"Application startup failed: {str(e)}"
        logger.error(error_message)
        if app_state:
            app_state.update_health('error', error_message)
        if not isinstance(e, ChatbotError):
            e = ChatbotError(error_message)
        metrics.record_error(e)
        raise
    finally:
        cleanup_resources()
if __name__ == "__main__":
    try:
        main()
    except Exception as e:
        logger.critical(f"Fatal error: {str(e)}")
        sys.exit(1)
2024-12-29 18:57:54,867 - __main__ - INFO - [2334330675.py:43] - Starting
application initialization
2024-12-29 18:57:54,867 - main - INFO - [2334330675.py:46] - Loading
environment and configuration
2024-12-29 18:57:54,869 - __main__ - INFO - [465277660.py:92] - Environment
variables loaded successfully
2024-12-29 18:57:54,869 - __main__ - INFO - [465277660.py:33] - Configuration
loaded successfully
2024-12-29 18:57:54,869 - __main__ - INFO - [2334330675.py:52] - Loading
documents
2024-12-29 18:57:54,869 - __main__ - INFO - [5827190.py:14] - Starting document
load: dataset/the_nestle_hr_policy_pdf_2012.pdf
2024-12-29 18:57:55,133 - __main__ - INFO - [5827190.py:47] - Successfully
loaded document with 8 pages in 0.26 seconds
2024-12-29 18:57:55,133 - __main__ - INFO - [2334330675.py:57] - Initializing
text processor
2024-12-29 18:57:55,144 - __main__ - INFO - [5827190.py:82] - TextProcessor
initialized successfully
2024-12-29 18:57:55,145 - __main__ - INFO - [5827190.py:126] - Starting document
processing
```

```
2024-12-29 18:57:55,145 - __main__ - INFO - [5827190.py:104] - Split documents
into 7 chunks in 0.00 seconds
2024-12-29 18:57:55,145 - __main__ - INFO - [5827190.py:132] - Creating vector
2024-12-29 18:57:55,551 - httpx - INFO - [ client.py:1025] - HTTP Request: POST
https://api.openai.com/v1/embeddings "HTTP/1.1 200 OK"
2024-12-29 18:57:55,674 - __main__ - INFO - [5827190.py:140] - Successfully
created vector store in 0.53 seconds. Stats: {
  "total chunks": 7,
  "processing_time": 0.5295591354370117
2024-12-29 18:57:55,675 - __main__ - INFO - [2334330675.py:64] - Initializing QA
system
2024-12-29 18:57:55,693 - __main__ - INFO - [3138667305.py:60] - QA system
initialized successfully
2024-12-29 18:57:55,693 - main - INFO - [2334330675.py:69] - Setting up chat
interface
2024-12-29 18:57:55,693 - main - INFO - [2334330675.py:20] - Health status
updated: healthy - Application initialized successfully
2024-12-29 18:57:55,694 - __main__ - INFO - [2334330675.py:77] - Application
initialization completed successfully
/opt/homebrew/anaconda3/envs/1720690009 20241228 v2/lib/python3.10/site-
packages/gradio/components/chatbot.py:242: UserWarning: You have not specified a
value for the `type` parameter. Defaulting to the 'tuples' format for chatbot
messages, but this is deprecated and will be removed in a future version of
Gradio. Please set type='messages' instead, which uses openai-style dictionaries
with 'role' and 'content' keys.
  warnings.warn(
2024-12-29 18:57:55,812 - httpx - INFO - [client.py:1025] - HTTP Request: GET
http://127.0.0.1:7861/gradio_api/startup-events "HTTP/1.1 200 OK"
2024-12-29 18:57:55,816 - httpx - INFO - [_client.py:1025] - HTTP Request: HEAD
http://127.0.0.1:7861/ "HTTP/1.1 200 OK"
* Running on local URL: http://127.0.0.1:7861
To create a public link, set `share=True` in `launch()`.
<IPython.core.display.HTML object>
2024-12-29 18:57:55,819 - __main__ - INFO - [2334330675.py:125] - Application
started successfully
  "health": {
    "status": "healthy",
    "last_check": "2024-12-29T18:57:55.693949",
    "message": "Application initialized successfully"
  },
  "uptime_seconds": 0.95167,
  "initialized": true,
```

```
"components": [
    "config",
    "documents",
    "text_processor",
    "vector store",
    "qa_system",
    "interface"
 ],
  "config": {
    "config": {
      "CHUNK_SIZE": 1000,
      "CHUNK_OVERLAP": 0,
      "MODEL_TEMPERATURE": 0,
      "EMBEDDING_MODEL": "text-embedding-ada-002",
      "MAX_RETRIES": 3,
      "RATE_LIMIT_PER_SECOND": 5,
      "LOG_LEVEL": 20
    },
    "load_time": "2024-12-29T18:57:54.789042",
    "age_seconds": 1.029908
  },
  "metrics": {
    "document_load_duration": {
      "mean": 0.2634608745574951,
      "min": 0.2634608745574951,
      "max": 0.2634608745574951,
      "count": 1
    },
    "text_splitting_duration": {
      "mean": 8.893013000488281e-05,
      "min": 8.893013000488281e-05,
      "max": 8.893013000488281e-05,
      "count": 1
    },
    "document processing duration": {
      "mean": 0.5295591354370117,
      "min": 0.5295591354370117,
      "max": 0.5295591354370117,
      "count": 1
    }
 }
}
2024-12-29 18:57:55,819 - __main__ - INFO - [2334330675.py:93] - Cleaning up
application resources
2024-12-29 18:57:55,819 - __main__ - INFO - [2334330675.py:100] - Application
shutdown complete
```

# 1.11 Testing and System Reporting Utilities

This section implements comprehensive testing procedures and system reporting capabilities. Let's examine each component of the testing and reporting framework:

### 1.11.1 Test Runner System:

### 1. Environment Testing:

- API key validation
- Path verification
- Configuration testing
- Permission checking

### 2. Document Processing Tests:

- Loading verification
- Document processing
- Chunking validation
- Format handling

### 3. QA System Testing:

- Query processing
- Response validation
- Rate limiting checks
- Error handling

# 1.11.2 System Reporting Framework:

### 1. Configuration Reporting:

- Current settings
- Environment state
- System parameters
- Runtime configuration

#### 2. Performance Analytics:

- Operation timing
- Resource utilization
- Response latency
- System load

### 3. Error Analysis:

- Error frequency
- Error patterns
- Impact assessment
- Resolution tracking

### 1.11.3 System Metrics:

#### 1. Resource Monitoring:

- Memory usage
- CPU utilization
- Storage metrics
- Network status

### 2. Performance Statistics:

- Response times
- Processing speed
- Resource efficiency
- System throughput

# 1.11.4 Implementation Features:

#### 1. Test Execution:

- Sequential testing
- Dependency handling
- State verification
- Error capture

### 2. Report Generation:

- Metrics compilation
- Statistics calculation
- Status summaries
- Performance analysis

### 3. System Analysis:

- Trend identification
- Pattern recognition
- Performance evaluation
- Resource assessment

### 1.11.5 Testing Components:

### 1. Functionality Testing:

- Core operations
- Error handling
- Recovery procedures
- State management

### 2. Performance Testing:

- Response timing
- Resource usage
- System limits
- Load handling

### 1.11.6 Reporting Features:

## 1. System Status:

- Component health
- Resource state
- Error conditions
- Performance metrics

### 2. Analysis Tools:

- Statistical processing
- Trend analysis
- Pattern detection
- Performance evaluation

```
[23]: def run_tests():
          """Basic application tests"""
          try:
              logger.info("Starting application tests")
              # Test environment loading
              api_key, pdf_path = load_env()
              assert api_key, "API key not loaded"
              assert pdf_path, "PDF path not loaded"
              # Test document loading
              documents = load document(pdf path)
              assert documents, "Documents not loaded"
              # Test text processing
              processor = TextProcessor()
              vector_store = processor.process_documents(documents)
              assert vector_store, "Vector store not created"
              # Test QA system
              qa_system = QASystem(vector_store)
              test_query = "What are the working hours?"
              response = qa_system.get_response(test_query)
              assert response, "No response received"
              logger.info("All tests passed successfully")
              return True
          except Exception as e:
              logger.error(f"Test failed: {str(e)}")
              return False
      def generate_system_report() -> Dict[str, Any]:
          """Generate comprehensive system report"""
          return {
              'timestamp': datetime.now().isoformat(),
              'configuration': config.to_dict(),
              'metrics': metrics.get_statistics(),
              'environment': {
                  'python_version': sys.version,
                   'platform': platform.platform(),
                  'memory_usage': psutil.Process().memory_info().rss / 1024 / 1024 #_
       \hookrightarrow MB
              },
              'log_summary': {
                  'error_count': len([r for r in metrics.metrics['errors']]),
                  'last_errors': metrics.metrics['errors'][-5:] # Last 5 errors
```

```
}
         }
[24]: print('\n========')
     print(f'Performing run tests')
     print('========\n')
     run_tests()
     print('\n=======')
     print(f'Generating System Report')
     print('=======\n')
     generate_system_report()
     2024-12-29 18:57:55,832 - __main__ - INFO - [2109332838.py:4] - Starting
     application tests
     2024-12-29 18:57:55,834 - __main__ - INFO - [465277660.py:92] - Environment
     variables loaded successfully
     2024-12-29 18:57:55,834 - __main__ - INFO - [5827190.py:14] - Starting document
     load: dataset/the_nestle_hr_policy_pdf_2012.pdf
     2024-12-29 18:57:56,001 - __main__ - INFO - [5827190.py:47] - Successfully
     loaded document with 8 pages in 0.17 seconds
     _____
     Performing run tests
     2024-12-29 18:57:56,012 - __main__ - INFO - [5827190.py:82] - TextProcessor
     initialized successfully
     2024-12-29 18:57:56,012 - __main__ - INFO - [5827190.py:126] - Starting document
     processing
     2024-12-29 18:57:56,012 - __main__ - INFO - [5827190.py:104] - Split documents
     into 7 chunks in 0.00 seconds
     2024-12-29 18:57:56,013 - __main__ - INFO - [5827190.py:132] - Creating vector
     store
     2024-12-29 18:57:56,270 - httpx - INFO - [_client.py:1025] - HTTP Request: POST
     https://api.openai.com/v1/embeddings "HTTP/1.1 200 OK"
     2024-12-29 18:57:56,341 - __main__ - INFO - [5827190.py:140] - Successfully
     created vector store in 0.33 seconds. Stats: {
       "total_chunks": 7,
       "processing_time": 0.3287811279296875
     2024-12-29 18:57:56,358 - __main__ - INFO - [3138667305.py:60] - QA system
     initialized successfully
     2024-12-29 18:57:56,741 - httpx - INFO - [_client.py:1025] - HTTP Request: POST
     https://api.openai.com/v1/embeddings "HTTP/1.1 200 OK"
     2024-12-29 18:57:57,164 - httpx - INFO - [_client.py:1025] - HTTP Request: POST
```

```
https://api.openai.com/v1/completions "HTTP/1.1 200 OK"
     2024-12-29 18:57:57,166 - __main__ - INFO - [3138667305.py:100] - Query
     processed successfully in 0.81 seconds
     2024-12-29 18:57:57,166 - __main__ - INFO - [2109332838.py:26] - All tests
     passed successfully
     2024-12-29 18:57:56,117 - httpx - INFO - [_client.py:1025] - HTTP Request: GET
     https://api.gradio.app/pkg-version "HTTP/1.1 200 OK"
     Generating System Report
     [24]: {'timestamp': '2024-12-29T18:57:57.168198',
       'configuration': {'config': {'CHUNK_SIZE': 1000,
         'CHUNK_OVERLAP': 0,
         'MODEL TEMPERATURE': 0,
         'EMBEDDING_MODEL': 'text-embedding-ada-002',
         'MAX RETRIES': 3,
         'RATE_LIMIT_PER_SECOND': 5,
         'LOG_LEVEL': 20},
        'load_time': '2024-12-29T18:57:54.789042',
        'age_seconds': 2.379163},
       'metrics': {'document_load_duration': {'mean': 0.21483445167541504,
         'min': 0.16620802879333496,
         'max': 0.2634608745574951,
         'count': 2},
        'text_splitting_duration': {'mean': 8.64267349243164e-05,
         'min': 8.392333984375e-05,
         'max': 8.893013000488281e-05,
         'count': 2},
        'document_processing_duration': {'mean': 0.4291701316833496,
         'min': 0.3287811279296875,
         'max': 0.5295591354370117,
         'count': 2},
        'get response duration': {'mean': 0.8077270984649658,
         'min': 0.8077270984649658,
         'max': 0.8077270984649658,
         'count': 1}},
       'environment': {'python_version': '3.10.12 | packaged by conda-forge | (main,
      Jun 23 2023, 22:41:52) [Clang 15.0.7]',
        'platform': 'macOS-15.2-arm64-arm-64bit',
        'memory_usage': 288.796875},
       'log_summary': {'error_count': 0, 'last_errors': []}}
     2024-12-29 18:59:46,850 - httpx - INFO - [_client.py:1025] - HTTP Request: POST
     https://api.openai.com/v1/embeddings "HTTP/1.1 200 OK"
```

```
2024-12-29 18:59:49,246 - httpx - INFO - [_client.py:1025] - HTTP Request: POST https://api.openai.com/v1/completions "HTTP/1.1 200 OK"
2024-12-29 18:59:49,253 - __main__ - INFO - [3138667305.py:100] - Query processed successfully in 2.66 seconds
2024-12-29 19:00:28,601 - httpx - INFO - [_client.py:1025] - HTTP Request: POST https://api.openai.com/v1/embeddings "HTTP/1.1 200 OK"
2024-12-29 19:00:30,805 - httpx - INFO - [_client.py:1025] - HTTP Request: POST https://api.openai.com/v1/completions "HTTP/1.1 200 OK"
2024-12-29 19:00:30,809 - __main__ - INFO - [3138667305.py:100] - Query processed successfully in 2.39 seconds
```