

Market Intelligence & Competitor Analysis Platform - Complete Implementation Roadmap

Project Overview

Project Name

Market Intelligence & Competitor Analysis Platform (MICAP)

Project Purpose

Build a scalable big data platform that performs large-scale sentiment analysis across social media to provide market intelligence and competitor analysis insights for Klewr Solutions' clients.

Dual Objectives

1. **Academic:** Fulfill Big Data Analysis course requirements with advanced sentiment analysis on 1.6M tweets
2. **Business:** Create a production-ready tool for Klewr Solutions to offer market intelligence services to clients

Technical Architecture

Core Technology Stack

Big Data Processing:

- Apache Spark 3.5.0 (PySpark)
- Hadoop HDFS 3.3.6
- Apache Kafka 3.6.0 (for streaming)
- MongoDB 7.0 (for results storage)

Backend:

- Python 3.11
- FastAPI 0.104.1
- Celery 5.3.4 (for async tasks)
- Redis 7.2 (for caching/queue)

Frontend:

- React 18.2.0
- TypeScript 5.3
- Material-UI 5.14
- Recharts 2.10 (for visualizations)
- D3.js 7.8 (for advanced visualizations)

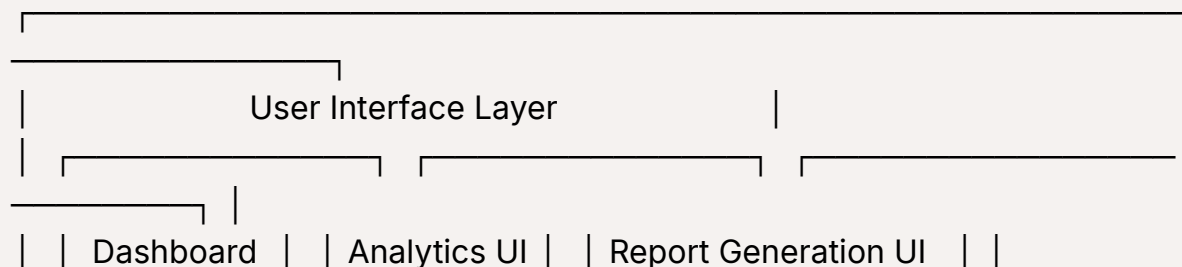
AI/ML:

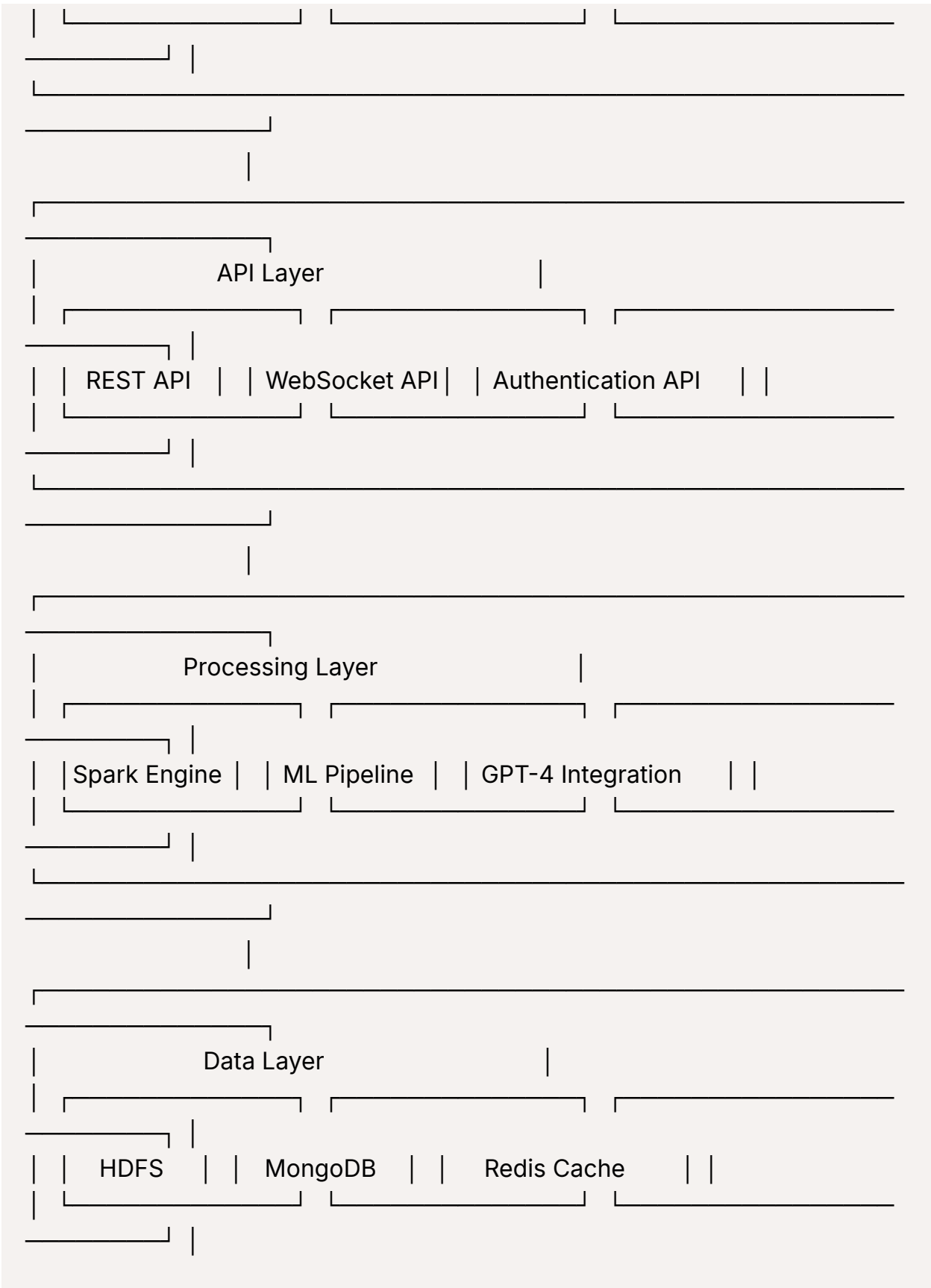
- OpenAI GPT-4 API
- Hugging Face Transformers 4.36
- scikit-learn 1.3.2
- NLTK 3.8.1
- spaCy 3.7.2

DevOps:

- Docker 24.0
- Docker Compose 2.23
- GitHub Actions (CI/CD)
- AWS/GCP (for deployment)

System Architecture Diagram





Phase 1: Foundation & Core Sentiment Analysis (Week 1-2)

1.1 Environment Setup

```
# Create project structure
mkdir micap
cd micap
mkdir -p {data,src/{spark,api,frontend,ml},tests,docs,notebooks,config,scripts}

# Initialize git repository
git init
echo "# Market Intelligence & Competitor Analysis Platform" > README.md

# Create Python virtual environment
python -m venv venv
source venv/bin/activate # On Windows: venv\Scripts\activate

# Create requirements files
touch requirements.txt requirements-dev.txt requirements-spark.txt
```

1.2 Data Pipeline Implementation

Task 1.2.1: Data Ingestion Module

Create `src/spark/data_ingestion.py` :

```
"""
Implement a data ingestion pipeline that:
1. Loads Sentiment140 dataset (1.6M tweets)
2. Validates data quality
```

3. Partitions data by date for efficient processing
4. Handles data schema evolution
5. Implements checkpointing for fault tolerance

Key functions to implement:

- load_sentiment140_data()
 - validate_data_quality()
 - partition_by_date()
 - save_to_hdfs()
- ```
"""
```

## Task 1.2.2: Data Preprocessing Module

Create `src/spark/preprocessing.py` :

```
"""
Implement preprocessing pipeline:
1. Text cleaning (URLs, mentions, special characters)
2. Emoji/emoticon handling and conversion
3. Language detection and filtering
4. Tokenization with multiple strategies
5. Stop word removal (configurable by domain)
6. Lemmatization and stemming options

Key functions:
- clean_text()
- handle_emojis()
- tokenize_text()
- remove_stopwords()
- lemmatize_text()
"""
```

## Task 1.2.3: Feature Engineering Module

Create `src/spark/feature_engineering.py` :

```
"""
```

Implement feature extraction:

1. TF-IDF vectorization (with configurable parameters)
2. N-gram extraction (unigrams, bigrams, trigrams)
3. Word embeddings (Word2Vec, FastText)
4. Sentiment lexicon features (VADER, TextBlob)
5. Temporal features (hour, day, week patterns)
6. Named entity recognition for brand/product mentions

Key functions:

- create\_tfidf\_features()
- extract\_ngrams()
- generate\_embeddings()
- extract\_lexicon\_features()
- extract\_temporal\_features()
- extract\_entities()

```
"""
```

## 1.3 Base Sentiment Analysis Models

### Task 1.3.1: Model Implementation

Create `src/ml/sentiment_models.py` :

```
"""
```

Implement distributed ML models:

1. Naive Bayes (baseline)
2. Logistic Regression with ElasticNet
3. Random Forest (distributed)
4. Gradient Boosting (XGBoost on Spark)
5. LSTM with distributed training
6. Transformer-based models (DistilBERT)

Include:

- Model training pipelines

- Hyperparameter tuning with Spark MLlib
  - Cross-validation implementation
  - Model serialization and versioning
- ```
"""
```

Phase 2: Competitor Analysis Features (Week 3-4)

2.1 Brand/Competitor Detection Module

Task 2.1.1: Entity Recognition System

Create `src/ml/entity_recognition.py` :

```
"""
Implement brand/product detection:
1. Custom NER model for brand recognition
2. Product mention extraction
3. Competitor mapping configuration
4. Fuzzy matching for brand variations
5. Context-aware disambiguation

Key components:
- BrandRecognizer class
- ProductExtractor class
- CompetitorMapper class
- EntityDisambiguator class
"""
```

Task 2.1.2: Competitor Sentiment Comparison

Create `src/spark/competitor_analysis.py` :

```
"""
Implement comparative analysis:
1. Sentiment aggregation by brand
2. Time-series sentiment comparison
```

3. Feature-level sentiment analysis
4. Market share of voice calculation
5. Sentiment momentum indicators

Functions:

- aggregate_brand_sentiment()
 - compare_competitor_sentiment()
 - analyze_feature_sentiment()
 - calculate_share_of_voice()
 - compute_sentiment_momentum()
- """

2.2 Market Intelligence Analytics

Task 2.2.1: Trend Detection

Create `src/ml/trend_detection.py` :

```
"""
Implement trend analysis:
1. Emerging topic detection using LDA
2. Sentiment trend forecasting
3. Anomaly detection in sentiment patterns
4. Viral content identification
5. Influencer impact analysis
```

Components:

- TopicModeler class
- TrendForecaster class
- AnomalyDetector class
- ViralityPredictor class

```
"""
```

Phase 3: Business Intelligence Layer (Week 5-6)

3.1 GPT-4 Integration for Insights

Task 3.1.1: AI Insights Generator

Create `src/ml/ai_insights.py` :

```
"""
Implement GPT-4 powered insights:
1. Automated insight generation from sentiment patterns
2. Competitive positioning recommendations
3. Market opportunity identification
4. Risk and threat analysis
5. Executive summary generation

Key classes:
- InsightGenerator
- RecommendationEngine
- OpportunityIdentifier
- ThreatAnalyzer
- SummaryGenerator

Include rate limiting, caching, and fallback mechanisms
"""
```

3.2 API Development

Task 3.2.1: FastAPI Backend

Create `src/api/main.py` :

```
"""
Implement REST API with endpoints:

/api/v1/brands
- GET: List monitored brands
- POST: Add new brand to monitor
- PUT: Update brand configuration
"""
```

- DELETE: Remove brand

/api/v1/analysis

- GET /sentiment/timeline: Time-series sentiment data
- GET /sentiment/comparison: Competitor comparison
- GET /topics/trending: Trending topics by brand
- GET /insights/summary: AI-generated insights

/api/v1/reports

- POST /generate: Generate custom report
- GET /download/{report_id}: Download report
- GET /schedule: View scheduled reports

/api/v1/alerts

- GET: List active alerts
- POST: Create sentiment alert
- PUT: Update alert thresholds

Include:

- Authentication middleware
- Rate limiting
- Request validation
- Error handling
- API documentation

"""

Task 3.2.2: Real-time Updates

Create `src/api/websocket.py` :

"""

Implement WebSocket for real-time updates:

1. Live sentiment score updates
2. Alert notifications
3. Processing status updates
4. New insight notifications

Features:

- Connection management
- Message queuing
- Broadcast capabilities
- Client state management

""

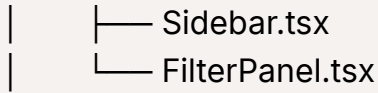
Phase 4: Frontend Development (Week 7-8)

4.1 React Dashboard

Task 4.1.1: Dashboard Components

Create component structure:

```
src/frontend/
├── components/
│   ├── Dashboard/
│   │   ├── SentimentOverview.tsx
│   │   ├── CompetitorComparison.tsx
│   │   ├── TrendingTopics.tsx
│   │   └── InsightsFeed.tsx
│   ├── Analytics/
│   │   ├── TimeSeriesChart.tsx
│   │   ├── SentimentHeatmap.tsx
│   │   ├── WordCloud.tsx
│   │   └── NetworkGraph.tsx
│   ├── Reports/
│   │   ├── ReportBuilder.tsx
│   │   ├── TemplateSelector.tsx
│   │   └── ExportOptions.tsx
│   └── Common/
│       └── Header.tsx
```



Task 4.1.2: State Management

Implement Redux/Context API for:

- Brand selection state
- Date range filters
- Sentiment data cache
- User preferences
- Alert configurations

4.2 Visualization Implementation

Task 4.2.1: Interactive Charts

Implement using Recharts and D3.js:

1. Multi-line sentiment timeline
2. Stacked area chart for market share
3. Sentiment gauge widgets
4. Correlation matrices
5. Geographic sentiment maps
6. Force-directed topic networks

Phase 5: Advanced Features & Integration (Week 9-10)

5.1 Streaming Analytics

Task 5.1.1: Kafka Integration

Create `src/spark/streaming_pipeline.py` :

```
"""
```

Implement real-time processing:

1. Kafka consumer for Twitter API/social media feeds
2. Micro-batch processing with Spark Streaming
3. Real-time sentiment scoring
4. Stream-to-batch integration
5. Late data handling

Components:

- StreamProcessor class
- SentimentScorer class
- AlertTrigger class
- StreamAggregator class

```
"""
```

5.2 Advanced Analytics

Task 5.2.1: Predictive Models

Create [src/ml/predictive_analytics.py](#) :

```
"""
```

Implement predictive capabilities:

1. Sentiment trend forecasting (ARIMA, Prophet)
2. Customer churn prediction based on sentiment
3. Campaign impact prediction
4. Crisis probability scoring
5. Market opportunity scoring

Models:

- TrendForecaster
- ChurnPredictor
- ImpactEstimator
- CrisisDetector

```
- OpportunityScorer
"""
```

Phase 6: Production Readiness (Week 11-12)

6.1 Testing Suite

Task 6.1.1: Comprehensive Tests

Create test structure:

```
tests/
├── unit/
│   ├── test_preprocessing.py
│   ├── test_sentiment_models.py
│   ├── test_api_endpoints.py
│   └── test_insights_generator.py
├── integration/
│   ├── test_spark_pipeline.py
│   ├── test_api_integration.py
│   └── test_frontend_api.py
└── performance/
    ├── test_scalability.py
    ├── test_response_times.py
    └── test_concurrent_users.py
```

6.2 Deployment Configuration

Task 6.2.1: Docker Setup

Create deployment files:

```
docker/
├── Dockerfile.spark
├── Dockerfile.api
└── Dockerfile.frontend
```

```
|— docker-compose.yml
|— docker-compose.prod.yml
|— .env.example
```

Task 6.2.2: CI/CD Pipeline

Create `.github/workflows/deploy.yml` :

```
# Implement GitHub Actions workflow for:
# 1. Automated testing
# 2. Docker image building
# 3. Deployment to cloud platform
# 4. Health checks
# 5. Rollback capabilities
```

Documentation Requirements

Technical Documentation

1. API documentation (OpenAPI/Swagger)
2. Code documentation (docstrings, comments)
3. Architecture diagrams
4. Database schemas
5. Deployment guide

Business Documentation

1. User manual for Klewr clients
2. Feature showcase presentation
3. ROI calculation templates
4. Case study templates
5. Marketing one-pager

Deliverables Checklist

For Academic Project:

- ☐ Working sentiment analysis pipeline
- ☐ Processing 1.6M tweets successfully
- ☐ Performance benchmarks and analysis
- ☐ Technical report with visualizations
- ☐ Presentation materials

For Klewr Solutions:

- ☐ Production-ready web application
- ☐ API with authentication
- ☐ Real-time analytics capabilities
- ☐ AI-powered insights generation
- ☐ Client onboarding materials
- ☐ Demo environment
- ☐ Source code with documentation
- ☐ Deployment packages

Data Sources Configuration

Primary Data:

- Sentiment140 dataset (academic requirement)

Additional Data Sources for Production:

```
{  
  "twitter": {  
    "api_version": "v2",  
    "endpoints": ["search/recent", "filtered_stream"],
```



```
"rate_limits": "300 requests/15min"
},
"reddit": {
  "subreddits": ["technology", "fintech", "IoT"],
  "api": "PRAW library"
},
"news": {
  "sources": ["NewsAPI", "Google News RSS"],
  "categories": ["business", "technology"]
}
}
```

Performance Targets

Processing Performance:

- Process 1.6M tweets in < 10 minutes
- Real-time sentiment scoring < 100ms latency
- API response time < 200ms for analytics queries
- Support 100 concurrent users

ML Model Performance:

- Sentiment accuracy > 85%
- Brand detection F1 score > 0.9
- Trend prediction RMSE < 0.15

Security Considerations

1. API authentication using JWT tokens
2. Rate limiting per client
3. Data encryption at rest and in transit
4. Secure storage of API keys

5. Input validation and sanitization
6. CORS configuration
7. SQL injection prevention
8. XSS protection

Monitoring & Maintenance

Monitoring Stack:

- Prometheus for metrics
- Grafana for visualization
- ELK stack for logs
- Sentry for error tracking

Maintenance Tasks:

1. Model retraining schedule
 2. Data cleanup policies
 3. Performance optimization
 4. Security updates
 5. Feature updates based on client feedback
-

Instructions for AI Implementation Assistant

When implementing this project:

1. **Start with Phase 1** - Get the core sentiment analysis working first
2. **Test each component** - Ensure each module works before moving to the next
3. **Use sample data** - Test with small datasets before processing full 1.6M tweets
4. **Document as you go** - Keep track of decisions and modifications
5. **Ask for clarification** - If any specification is unclear, ask for details

6. **Optimize iteratively** - Get it working first, then optimize
7. **Security first** - Implement security measures from the beginning
8. **Keep it modular** - Ensure components can be updated independently

This roadmap provides complete specifications for building a production-ready Market Intelligence & Competitor Analysis Platform that satisfies both academic requirements and creates a valuable tool for Klewr Solutions.