PROJECT REPORT: AI-POWERED RECOMMENDATION SYSTEM

Project Overview:

Built a scalable AI-powered recommendation system for TechCorp Inc. that serves over 2 million daily active users.

Technical Implementation:

- Backend: Python with FastAPI framework

- Database: PostgreSQL for user data, Redis for caching

- AI/ML: TensorFlow for deep learning models, OpenAI GPT-4 for natural language processina
- Infrastructure: AWS EC2, S3, Lambda, and RDS Vector Database: Pinecone for similarity search
- Message Queue: Apache Kafka for real-time data processing

Architecture:

- Data Collection Layer: Collects user behavior data and preferences
- Feature Engineering: Processes raw data into ML-ready features Model Training: Uses collaborative filtering and deep learning

4. Inference Engine: Real-time recommendation generation

5. API Layer: RESTful APIs for frontend integration

Caching Layer: Redis for fast response times

Key Features Implemented:

- Real-time personalized recommendations

- Content-based filtering using NLP

- Collaborative filtering for user similarity
 A/B testing framework for model evaluation
- RAG (Retrieval-Augmented Generation) for explanations Multi-armed bandit for exploration vs exploitation

Performance Metrics:

- Latency: Average response time of 50ms

- Throughput: Handles 10,000 requests per second - Accuracy: 35% improvement in user engagement
- Scalability: Successfully scaled from 10K to 2M+ users
 Cost Efficiency: Reduced infrastructure costs by 25%

Code Quality:

- 95% test coverage with unit and integration tests

- Comprehensive documentation and API specs

- CI/CD pipeline with automated testing and deployment

- Code review process with senior engineers

- Monitoring and alerting with Prometheus and Grafana

Challenges Overcome:

- 1. Cold Start Problem: Implemented hybrid approach combining content and collabo rative filtering
- 2. Scalability: Designed microservices architecture with horizontal scaling

3. Real-time Processing: Used Apache Kafka for streaming data processing

4. Model Drift: Implemented automated retraining pipeline

Data Privacy: Ensured GDPR compliance with data anonymization

Results and Impact:

- 35% increase in user engagement metrics

- 40% reduction in system latency

- 92% user satisfaction scoreSuccessfully handling 2M+ daily active users
- Became core product differentiator for the company

Future Enhancements:

- Integration with more data sources

Advanced deep learning modelsReal-time model updatesEnhanced explainability featuresMobile app optimization

Technologies Used: Python, FastAPI, TensorFlow, PostgreSQL, Redis, AWS, Docker, Kubernetes, Apache Kafka, Pinecone, OpenAI GPT-4, Prometheus, Grafana, GitHub Actions

Project Duration: 8 months (2023-2024)
Team Size: 5 engineers (led the technical implementation)