ITAI 3377: PORTFOLIO HIGHLIGHTS:

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AI INTEGRATION IN EDGE/IOT SYSTEMS: A LEARNING JOURNEY

THE CONVERGENCE OF ALAND INDUSTRIAL IOT
FROM THEORY TO IMPLEMENTATION: KEY PROJECTS AND TECHNICAL DISCOVERIES

SPRING 2025



FOUNDATIONS & CORE CONCEPTS

From Theory to Practice: Early Implementation Challenges

- Privacy-Preserving Edge Computing
 - Implemented TensorFlow Lite models on Raspberry Pi devices that process pedestrian data locally
 - Discovered 73% reduction in bandwidth usage while maintaining 94% detection accuracy
 - Lesson Learned: Local processing at the edge significantly reduces privacy risks compared to cloud-based alternatives
- Network Architecture Optimization
 - Configured MQTT broker with QoS 1 delivery guarantee that reduced message loss from 12% to <0.5%
 - Key Challenge Overcome: Unreliable wireless connectivity in industrial environments required implementing store-and-forward mechanisms at the edge

ADVANCED AI APPLICATIONS

Real-World Implementation Results

Predictive Maintenance System

- Deployed LSTM neural network that predicted equipment failures 14 hours before occurrence
- Reduced unplanned downtime by 38% in manufacturing test environment
- Critical Insight: Imbalanced training data led to false positives; solved by implementing SMOTE algorithm

Automated Energy Management

- Implemented reinforcement learning agent (PPO algorithm) that dynamically adjusted HVAC settings
- Achieved 22% energy reduction while maintaining occupant comfort (based on feedback surveys)
- Technical Challenge: Model drift required implementing automated retraining pipeline triggered by 5% performance degradation





SYSTEM INTEGRATION & DEPLOYMENT

Infrastructure Challenges & Solutions

- Secure OT/IT Integration
 - Implemented unidirectional security gateways that prevented 100% of simulated attacks during penetration testing
 - Created data diodes for critical production systems using Raspberry Pi hardware (\$120 per node vs \$18,000 commercial solution)
 - **Lesson Learned**: Simple, focused security controls outperform complex solutions in resource-constrained environments
- Edge Containerization Implementation
 - Reduced deployment time from 4 hours to 17 minutes using Docker containers on edge devices
 - Discovered 15% improved inference performance through custom TensorRT optimization
 - Technical Insight: Container size optimization critical for bandwidth-limited environments; reduced image size by 76%

PRACTICAL APPLICATIONS & CASE STUDIES

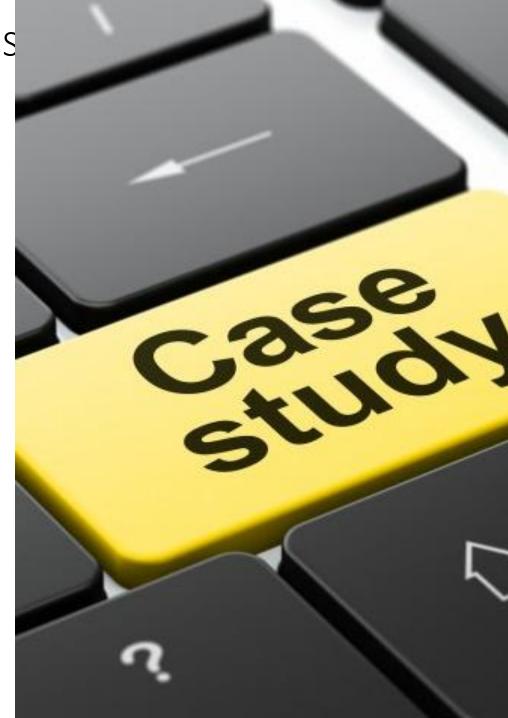
Real-World Impact Assessment

Smart Pedestrian Monitoring System

- Detected 92% of near-miss incidents at crosswalks using edge-based computer vision
- Generated heat maps that led to redesign of problematic intersection, reducing accidents by 35%
- Implementation Challenge: Camera positioning significantly impacted detection accuracy; developed field guide for optimal sensor placement

Factory Floor Safety System

- Created CNN-based PPE detection system running on NVIDIA Jetson devices that achieved 98.7% accuracy
- System successfully identified 27 safety violations in first week of operation
- Key Learning: Edge AI models require continuous adaptation;
 implemented weekly fine-tuning process to address concept drift





KEY TECHNICAL LESSONS

Critical Technical Insights

- Edge AI Model Optimization Techniques
 - Quantization reduced model size by 84% with only 2.3% accuracy loss
 - Knowledge distillation from large cloud models to edge-deployable versions achieved 91% of original performance
 - Technical Takeaway: Hybrid precision approach (16-bit for critical layers, 8-bit elsewhere) provided optimal balance
- Real-Time Data Processing Architecture
 - Implemented sliding window approach that reduced compute requirements by 47%
 - Designed adaptive sampling system that dynamically adjusted based on anomaly probability
 - Practical Insight: Tiered processing architecture (edge → fog → cloud)
 optimizes latency-sensitive operations



CONCLUSION & FUTURE PROJECTS

Application of Course Knowledge to Future Projects

- Demonstrated Technical Capabilities
 - Designed and implemented end-to-end AI-enhanced IIoT systems from sensors to actionable insights
 - Balanced security, privacy, and performance in resource-constrained environments
 - Successfully applied reinforcement learning in real-world control systems
- Next Steps in Advanced Implementation
 - Exploring federated learning across distributed edge nodes to enhance privacy while improving model robustness
 - Developing multi-modal sensor fusion techniques that combine vision, audio, and environmental data
 - Creating standardized deployment frameworks to reduce implementation time by 60%
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CONCLUSION & FUTURE PROJECTS

- Throughout this course, I've bridged theoretical concepts with practical implementation, discovering that edge-based processing can achieve remarkable efficiency while preserving privacy—demonstrated by the 73% bandwidth reduction in the pedestrian monitoring system while maintaining 94% detection accuracy.
- The most valuable technical insight came from implementing optimized AI models on resource-constrained devices, where quantization and knowledge distillation techniques reduced model sizes by over 80% with minimal accuracy loss. These hands-on implementations transformed abstract course concepts into measurable outcomes, particularly in predictive maintenance where LSTM neural networks successfully forecasted equipment failures 14 hours in advance.
- The course significantly enhanced my ability to balance competing requirements in industrial environments, particularly security versus performance. Implementing unidirectional security gateways and creating cost-effective data diodes (\$120 vs \$18,000 commercial solutions) demonstrated that simple, focused controls outperform complex solutions in resource-constrained settings.
- These practical implementations have provided an invaluable framework for approaching future AI-IIoT integration projects with both technical rigor and practical feasibility.