

# Dental AI Platform - Project Report

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**Project Title:** Multi-Model AI System for Dental X-Ray Analysis **Author:** Harsha **Version:** 2.4 **Date:** December 2025

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## Executive Summary

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This project developed a comprehensive dental AI platform that combines custom YOLOv8 object detection with multiple AI language models to provide accurate, visual analysis of dental X-rays. The system achieved 88% mAP accuracy in detecting dental pathologies and provides real-time analysis through an intuitive web interface.

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## 1. Project Duration

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**Timeline:** December 19 - December 22, 2025 **Total Duration:** Approximately 3-4 days of active development

### Development Phases:

- **Day 1:** Initial concept, tech stack selection, and vision model integration
- **Day 2:** YOLO integration, multi-model text analysis, and custom training
- **Day 3:** Detection refinement, dataset quality improvements, and UI/UX fixes
- **Day 4:** Final optimizations, documentation, and code cleanup

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## 2. Learning Outcomes

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### Technical Skills Acquired

#### 1. Computer Vision & Object Detection

- Implemented YOLOv8 custom training pipeline
- Achieved 88% mAP@50 accuracy on dental X-ray dataset

- Developed class-specific confidence thresholds and spatial filtering
- Learned image preprocessing and augmentation techniques

## 2. Multi-Model AI Integration

- Integrated three AI models in parallel (GPT-4o-mini, Llama 3.3 70B, Qwen 3 32B)
- Implemented asynchronous API calls for optimal performance
- Built conversation context management system
- Designed message routing logic for different input types

## 3. Web Application Development

- Built production-ready Gradio interface with multiple tabs
- Implemented state management for conversation history
- Created interactive annotation playground
- Developed PDF report generation system

## 4. Image Processing

- Implemented CLAHE (Contrast Limited Adaptive Histogram Equalization) enhancement
- Developed proper grayscale-to-RGB conversion preserving full dynamic range
- Created bounding box annotation system with color coding
- Built image quality validation and filtering system

## 5. Software Engineering Practices

- Modular architecture with separation of concerns
- Error handling and resilience patterns
- Performance optimization (caching, async processing)
- Comprehensive documentation and code organization

## Domain Knowledge Gained

- **Dental X-Ray Analysis:** Understanding of wisdom tooth impaction, dental pathologies, and X-ray interpretation
- **Medical AI Ethics:** Awareness of limitations and disclaimer requirements for medical AI systems
- **Dataset Management:** Experience with HuggingFace datasets, local dataset handling, and quality validation

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### 3. Challenges Encountered

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#### Challenge 1: Vision Model Hallucination

**Problem:** Initial implementation using GPT-4 Vision and Gemini Vision models produced inaccurate bounding box coordinates, often hallucinating detection locations.

**Solution:**

- Switched to YOLOv8 custom-trained model for accurate object detection
- Separated detection (YOLO) from analysis (text models)
- Implemented spatial filtering to reduce false positives

**Impact:** Detection accuracy improved from unreliable to 88% mAP@50.

#### Challenge 2: Image Corruption Issues

**Problem:** Images retrieved from HuggingFace dataset appeared corrupted (static noise, all black/white) in the Gradio interface.

**Root Causes:**

- Complex validation logic was corrupting images during processing
- Gradio PIL object serialization issues
- Dataset contained binary mask images mixed with real X-rays

**Solution:**

- Simplified image loading pipeline
- Switched to filepath-based display instead of PIL objects
- Implemented binary/mask image filtering (skips images with <10 unique values)
- Added proper grayscale-to-RGB conversion using NumPy stacking

**Impact:** All images now display correctly with proper grayscale preservation.

#### Challenge 3: Dataset Quality Issues

**Problem:** HuggingFace dataset contained many binary segmentation masks (only 2 unique pixel values) instead of real X-rays.

**Solution:**

- Implemented automatic filtering to skip binary images

- Added quality validation (unique values  $\geq 10$ , mean brightness 5-250)
- Created retry mechanism (tries up to 50 indices to find valid X-rays)
- Built quality scanning tool to analyze dataset composition

**Impact:** System now only displays valid X-rays, improving user experience.

## Challenge 4: Context Leakage in AI Models

**Problem:** AI models were inferring YOLO detection results from previous conversations even when no current image was present.

**Solution:**

- Modified context building to only include YOLO results when explicitly available
- Added explicit checks to prevent YOLO context inference
- Updated system prompts with critical instructions
- Implemented stricter None checks

**Impact:** Models now correctly state when X-ray analysis is needed.

## Challenge 5: Performance Optimization

**Problem:** Sequential API calls were slow, and image loading was inefficient.

**Solution:**

- Implemented parallel async API calls for all three models
- Added LRU cache for dataset images (10-image cache)
- Optimized image processing pipeline
- Reduced detection inference time to  $< 100\text{ms}$

**Impact:** Overall response time improved significantly with parallel processing.

## Challenge 6: YOLO Model Training and Hyperparameter Tuning

**Problem:** Initial YOLO model had high false positive rates, detecting impacted teeth in incorrect locations (middle of jaw instead of edges).

**Root Causes:**

- Pre-trained model was trained on generic objects, not dental X-rays
- Default confidence thresholds were too low

- No spatial filtering for anatomical constraints

**Solution:**

- Trained custom YOLO model on 1,075 dental X-rays from Roboflow dataset
- Implemented class-specific confidence thresholds (0.25 for Impacted, 0.30 for others)
- Added spatial filtering to restrict impacted tooth detections to jaw edges ( $x < 0.30$  or  $x > 0.70$ )
- Fine-tuned IoU threshold (0.4) for better NMS (Non-Maximum Suppression)
- Reduced training time to 9 minutes using transfer learning

**Impact:** Detection accuracy improved from ~70% to 88% mAP@50, false positives reduced significantly.

## Challenge 7: State Management Complexity

**Problem:** Managing conversation state, annotated images, and model selection across multiple tabs and interactions became complex and error-prone.

**Issues Encountered:**

- Annotated images disappearing during follow-up questions
- Conversation context not persisting correctly
- Model selection state not updating all displayed responses
- State synchronization between UI components

**Solution:**

- Implemented separate state variables: `conversation_state` (full context) and `stored_annotated_images` (persistent images)
- Created `update_model_selection()` function to update all historical responses when model changes
- Added explicit state management for image persistence across conversation turns
- Implemented proper state initialization and cleanup

**Impact:** Stable conversation flow with persistent images and consistent model switching.

## Challenge 8: Gradio UI/UX Issues

**Problem:** Multiple UI/UX issues affected user experience, including image dragging in Firefox, modal display problems, and inconsistent button behaviors.

**Specific Issues:**

- Images in Annotation Playground were draggable in Firefox, interfering with click detection
- Fullscreen image modal not working correctly
- Button styling inconsistencies across different browsers
- Image gallery not displaying properly on mobile devices

**Solution:**

- Added CSS and JavaScript to prevent image dragging in Firefox
- Implemented proper modal overlay with ESC key support
- Created consistent button styling with hover effects
- Added responsive design considerations for mobile
- Fixed gallery display with proper image sizing constraints

**Impact:** Improved user experience across all browsers and devices.

## Challenge 9: Dependency Management and Version Conflicts

**Problem:** Encountered version conflicts between different libraries, particularly with Gradio, PyTorch, and CUDA dependencies.

**Issues Encountered:**

- Gradio 6.x API changes breaking existing code
- PyTorch CUDA version mismatches
- httpx version conflicts with OpenAI API client
- PIL/Pillow version compatibility issues

**Solution:**

- Pinned specific versions in `requirements.txt` (e.g., `httpx<0.28` , `pillow==10.4.0` )
- Updated code to use Gradio 6.x API ( `type="filepath"` instead of `type="pil"` )
- Created virtual environment isolation
- Documented exact dependency versions for reproducibility

**Impact:** Stable, reproducible environment across different systems.

## Challenge 10: Error Handling and Edge Cases

**Problem:** System crashed or produced incorrect outputs when handling edge cases like empty images, API failures, malformed responses, or missing model files.

**Edge Cases Encountered:**

- API rate limiting causing failures
- Missing YOLO model file on first run
- Empty or corrupted image uploads
- Network timeouts during API calls
- Invalid bounding box coordinates

**Solution:**

- Implemented comprehensive try-catch blocks with detailed error messages
- Added fallback to base YOLOv8n model if custom model not found
- Created validation checks for image inputs
- Implemented retry logic for API calls
- Added graceful degradation (show error messages instead of crashing)
- Created user-friendly error messages with troubleshooting tips

**Impact:** Robust system that handles errors gracefully without crashing.

**Challenge 12: Model Selection and Response Switching**

**Problem:** When users switched between AI models (GPT-4o-mini, Llama, Qwen), only new responses showed the selected model, while historical responses remained unchanged, causing confusion.

**Solution:**

- Implemented `update_model_selection()` function that updates ALL historical responses
- Created mapping between conversation state entries and history display entries
- Preserved images and other metadata while updating response content
- Added visual feedback showing currently selected model

**Impact:** Consistent model selection across entire conversation history.

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## 4. Use of AI Tools

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### AI Tools Used for Development

## Primary Development Assistant: Claude

- Extensively used throughout the project for:
- Code generation and implementation
- Architecture design and refactoring
- Debugging and troubleshooting
- Documentation writing
- Code review and optimization suggestions

## Key Interactions:

### 1. Project Planning & Architecture (Day 1)

- "How to structure the flow for a dental AI platform project"
- "Which tech stack is preferred given time constraints and deliverables"
- "Why is Gradio better than other frameworks for this use case?"
- Result: Comprehensive project structure and technology recommendations

### 2. Initial Implementation (Day 1)

- "Build a Gradio app with 2 tabs for dental AI platform"
- Detailed specifications for wisdom tooth detection and multi-model chatbot
- Result: Complete initial application structure with GPT-4 Vision and Gemini Vision integration

### 3. YOLO Integration (Day 2)

- "How to integrate YOLOv8 for accurate bounding box detection"
- "Implement custom YOLO training pipeline for dental X-rays"
- "Add class-specific confidence thresholds and spatial filtering"
- Result: Custom YOLO model integration replacing vision models

### 4. Multi-Model AI Setup (Day 2)

- "Implement parallel async API calls for GPT-4o-mini, Llama 3.3 70B, and Qwen 3 32B"
- "Create conversation context management system"
- "Handle model response formatting and display"
- Result: Robust multi-model system with parallel execution

### 5. Image Processing Debugging (Day 3)

- "Images from HuggingFace are corrupted - how to fix?"



- "Images are still corrupted (static/black) - debug step by step"
- "Implement proper grayscale-to-RGB conversion preserving full range"
- Result: Fixed image corruption through filepath-based display and proper conversion

## 6. **Dataset Quality Issues** (Day 3)

- "HuggingFace dataset contains binary images - filter them out"
- "Add automatic quality validation and retry mechanism"
- "Create dataset quality scanning tool"
- Result: Automatic filtering system skipping binary/mask images

## 7. **State Management** (Day 2-3)

- "Fix annotated images disappearing during follow-up questions"
- "Implement persistent image storage across conversation turns"
- "Update all historical responses when model selection changes"
- Result: Robust state management with persistent images

## 8. **Error Handling** (Throughout)

- "Add comprehensive error handling for API failures"
- "Implement fallback mechanisms for missing model files"
- "Create user-friendly error messages with troubleshooting tips"
- Result: Robust error handling throughout the application

## 9. **UI/UX Improvements** (Day 3)

- "Fix image dragging issue in Firefox for Annotation Playground"
- "Implement fullscreen image modal with ESC key support"
- "Add consistent button styling and hover effects"
- Result: Improved cross-browser compatibility and user experience

## 10. **PDF Report Generation** (Day 2-3)

- "Create professional PDF report generation system"
- "Include YOLO detections, AI analysis, and recommendations"
- "Format clinical reports with proper styling"
- Result: Complete PDF report generation feature

## 11. **Documentation** (Day 4)

- "Create consolidated documentation covering all aspects"
- "Write detailed architecture explanation with code examples"
- "Update all documentation to reflect current implementation"
- Result: Comprehensive documentation (1,200+ lines)

#### 12. **Code Optimization** (Throughout)

- "Optimize async API calls for better performance"
- "Implement LRU cache for dataset images"
- "Reduce YOLO inference time"
- Result: Significant performance improvements

#### 13. **Context Management Fix** (Day 2-3)

- "Fix AI models inferring YOLO results from previous conversations"
- "Prevent context leakage when no image is present"
- "Update system prompts with critical instructions"
- Result: Proper context isolation preventing false inferences

#### 14. **Local Dataset Support** (Day 3)

- "Add automatic detection of local YOLO format datasets"
- "Implement fallback to HuggingFace if local dataset not found"
- "Create seamless switching between data sources"
- Result: Flexible dataset loading supporting both local and remote sources

#### 15. **Code Cleanup** (Day 4)

- "Remove unnecessary test files and debug scripts"
- "Clean up Python cache files and temporary images"
- "Organize project structure"
- Result: Clean, production-ready codebase

**Estimated Usage:** 50+ prompts/interactions over the project duration

#### **Interaction Patterns:**

- **Problem-Solving:** Many interactions followed a pattern of describing a problem, getting AI suggestions, implementing, testing, and iterating
- **Code Generation:** AI generated initial code structures, which were then refined through multiple iterations

- **Debugging:** AI assisted in debugging complex issues through step-by-step analysis
- **Architecture Design:** AI provided architectural recommendations that were validated and adapted
- **Documentation:** AI helped structure and write comprehensive documentation

### Example Interaction Flow:

#### Example 1: Image Corruption Debugging

User: "Images from HuggingFace are corrupted - how to fix?"

AI: [Suggests validation logic and image processing fixes]

User: "This is worse. Literally every image is static noise"

AI: [Suggests simplifying approach, removing complex validation]

User: "Images are still corrupted. Let's debug step by step"

AI: [Creates test scripts, identifies Gradio PIL serialization issue]

User: [Implements filepath-based display]

Result: Problem solved through iterative debugging

#### Example 2: YOLO Integration

User: "Vision models hallucinate coordinates. Need accurate detection"

AI: [Suggests YOLOv8 integration]

User: "How to train custom YOLO model for dental X-rays?"

AI: [Provides training pipeline with Roboflow dataset]

User: "Model has high false positives in wrong locations"

AI: [Suggests spatial filtering and class-specific thresholds]

Result: 88% mAP@50 accuracy achieved

#### Example 3: Multi-Model Setup

User: "Need to query 3 AI models in parallel"

AI: [Provides async implementation with asyncio]

User: "How to format and display responses from multiple models?"

AI: [Suggests response formatting and model selection UI]

User: "Historical responses don't update when switching models"

AI: [Provides update\_model\_selection() function]

Result: Seamless multi-model experience

### AI Assistance Effectiveness:

- **High Impact:** Architecture decisions, debugging complex issues, code structure
- **Medium Impact:** Implementation details, documentation writing, optimization
- **Low Impact:** Simple syntax questions, basic code snippets (could be done manually)

**Time Saved:** Estimated 60-70% reduction in development time through AI assistance, particularly in:

- Initial code scaffolding
- Debugging complex issues
- Writing comprehensive documentation
- Architecture design decisions

## AI Models Integrated in the Application

### 1. GPT-4o-mini (OpenAI)

- Purpose: Text-based clinical analysis
- Usage: Parallel execution with other models
- Performance: Fast, accurate responses

### 2. Llama 3.3 70B (via Groq)

- Purpose: Alternative perspective on analysis
- Usage: Parallel execution for consensus
- Performance: High-quality medical reasoning

### 3. Qwen 3 32B (via Groq)

- Purpose: Third model for multi-model consensus
- Usage: Parallel execution
- Performance: Good balance of speed and accuracy

## AI Tool Impact on Project

### Positive Impacts:

- **Accelerated Development:** AI assistance significantly reduced development time

- **Code Quality:** AI suggestions improved code organization and best practices
- **Problem Solving:** AI helped debug complex issues (image corruption, context leakage)
- **Documentation:** AI assisted in creating comprehensive documentation

#### Challenges with AI Tools:

- **Hallucination:** Initial vision models produced inaccurate coordinates (led to YOLO solution)
- **Context Management:** Required careful prompt engineering to prevent context leakage
- **Iterative Refinement:** Multiple iterations needed for complex fixes (image corruption)

#### Lessons Learned:

- AI tools are powerful accelerators but require human oversight
- Domain-specific problems (medical imaging) benefit from specialized models (YOLO)
- Combining AI assistance with custom-trained models yields best results
- Critical to validate AI-generated code and solutions

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## 5. Project Achievements

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### Technical Achievements

✓ **Custom YOLO Model:** Trained on 1,075 dental X-rays achieving 88% mAP@50 ✓ **Multi-Model System:** Successfully integrated 3 AI models with parallel execution ✓ **Real-Time Processing:** <100ms detection + parallel AI inference ✓ **Robust Image Handling:** Proper grayscale preservation and quality filtering ✓ **Production-Ready UI:** Full-featured Gradio interface with multiple tabs ✓ **PDF Report Generation:** Professional clinical report generation ✓ **Comprehensive Documentation:** 1,200+ lines of consolidated documentation

### Key Metrics

- **Detection Accuracy:** 88% mAP@50
- **Inference Speed:** ~5ms per image (YOLO)
- **Training Time:** 9 minutes (GPU)
- **Response Time:** <3 seconds (including all 3 AI models)
- **Codebase Size:** ~3,500 lines of Python code
- **Documentation:** 1,200+ lines

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## 6. Future Improvements

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### 1. Model Enhancements

- Expand YOLO training to more dental pathologies
- Fine-tune confidence thresholds based on clinical feedback
- Add support for panoramic X-rays

### 2. User Experience

- Add user authentication and history saving
- Implement batch processing for multiple X-rays
- Add comparison view for before/after treatment

### 3. Performance

- Implement model quantization for faster inference
- Add GPU acceleration for YOLO detection
- Optimize image caching strategy

### 4. Clinical Integration

- Add DICOM file support
- Integrate with dental practice management systems
- Add HIPAA compliance features

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## 7. Conclusion

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This project successfully developed a production-ready dental AI platform that combines the accuracy of custom-trained YOLOv8 models with the reasoning capabilities of multiple AI language models. The system demonstrates the power of hybrid AI approaches, where specialized computer vision models handle detection while general-purpose language models provide clinical analysis.

The project provided valuable learning experiences in computer vision, multi-model AI integration, web development, and software engineering. The extensive use of AI development tools accelerated the development process while teaching important lessons about their capabilities and limitations.

**Final Status:** Production-ready system with comprehensive documentation and robust error

handling.

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