

Performance Testing - Artificial Intelligence

Diabetic Retinopathy Detection System

TEST OVERVIEW

Project: Diabetic Retinopathy Detection System
Test Date: March 17-20, 2026
Environment: Development & Production
Objectives: Validate model accuracy ($\geq 85\%$), measure inference time, evaluate scalability

MODEL ARCHITECTURE

Base Model: Xception (Pre-trained on ImageNet)
Input Shape: (299, 299, 3)
Output Classes: 5 (No_DR, Mild, Moderate, Severe, Proliferate_DR)
Total Parameters: ~22.9 million
Trainable Parameters: ~8.5 million
Model Size: 88 MB

Custom Head:

GlobalAveragePooling2D → Dense(1024) → Dropout(0.5) →
Dense(512) → Dropout(0.4) → Dense(256) → Dropout(0.3) →
Dense(5, softmax)

TRAINING PERFORMANCE

Configuration: - Optimizer: Adam (lr=0.0001) - Loss: Categorical Crossentropy - Batch Size: 32 - Epochs: 50 (early stopping at 48) - Training Data: 2,930 images (80%) - Validation Data: 732 images (20%)

Results:

Metric	Value
Training Time	3.5 hours (GPU)
Training Accuracy	92.45%
Validation Accuracy	88.12%
Training Loss	0.2134
Validation Loss	0.3567

Metric	Value
Best Epoch	38

MODEL ACCURACY TESTING

Overall Performance

Metric	Value	Target	Status
Overall Accuracy	88.12%	≥85%	✅ PASS
Precision (Macro)	86.73%	≥80%	✅ PASS
Recall (Macro)	85.91%	≥80%	✅ PASS
F1-Score (Macro)	86.31%	≥80%	✅ PASS

Class-wise Performance

Class	Samples	Precision	Recall	F1-Score
No_DR	361	91.2%	93.1%	92.1%
Mild	74	82.5%	78.4%	80.4%
Moderate	200	88.7%	89.5%	89.1%
Severe	39	79.3%	74.4%	76.8%
Proliferate_DR	58	91.8%	94.8%	93.3%

Analysis: - ✅ All classes meet minimum F1-score (≥75%) - ✅ Best performance on No_DR and Proliferate_DR - ⚠️ Severe class has lowest performance (smallest dataset)

INFERENCE PERFORMANCE

Test Setup: 100 random samples, Intel i7, 16GB RAM, NVIDIA GTX 1660 Ti

Metric	Value	Target	Status
Average Inference Time	2.34s	≤5s	✅ PASS
Minimum Time	1.89s	-	✅
Maximum Time	3.12s	-	✅
95th Percentile	2.87s	-	✅

Breakdown: - Model Loading (first time): 4.2s - Image Preprocessing: 0.15s - Model Inference: 1.95s - Post-processing: 0.24s

LOAD TESTING

Tool: Apache JMeter | **Duration:** 10 minutes

Concurrent Users	Avg Response	Throughput	Error Rate	Status
1	2.5s	24/min	0%	✓ PASS
5	3.1s	96/min	0%	✓ PASS
10	4.2s	142/min	0%	✓ PASS
20	7.8s	153/min	2.1%	⚠ WARNING
50	15.3s	195/min	8.5%	✗ FAIL

Findings: - System handles up to 10 concurrent users efficiently - Performance degrades beyond 20 users - Bottleneck: Single Flask instance - **Recommendation:** Implement load balancing

RESOURCE UTILIZATION

CPU Usage

Operation	CPU Usage	Duration
Idle	2-5%	-
Model Loading	85-95%	4.2s
Preprocessing	15-25%	0.15s
Inference	75-90%	1.95s

Memory Usage

Component	Memory
Flask Application	150 MB
Loaded Model	320 MB
TensorFlow Runtime	450 MB
Image Processing	50 MB
Total	~1 GB

MODEL ROBUSTNESS TESTING

Test Condition	Accuracy	Impact
High Quality (Original)	88.12%	Baseline
Compressed (JPEG 80%)	86.45%	Minimal
Compressed (JPEG 50%)	82.31%	Noticeable
Low Resolution (512x512)	85.67%	Acceptable

Test Condition	Accuracy	Impact
Blurred (Gaussian $\sigma=2$)	79.23%	Significant
Brightness +20%	86.89%	Minimal
Brightness -20%	84.12%	Slight

Conclusion: Model reasonably robust to common image variations

EDGE CASES TESTING

Edge Case	Expected	Actual	Status
Very dark image	Low confidence	45%	✓ PASS
Very bright image	Low confidence	52%	✓ PASS
Non-retinal image	Low confidence	38%	✓ PASS
Corrupted file	Error	Error handled	✓ PASS
Oversized (>16MB)	Rejection	Rejected	✓ PASS
Wrong format (.txt)	Rejection	Rejected	✓ PASS

COMPARISON WITH BENCHMARKS

Metric	Our System	Industry Avg	Best in Class
Accuracy	88.12%	85-90%	95%+
Inference Time	2.34s	2-5s	<1s
Model Size	88 MB	50-200 MB	25 MB
Concurrent Users	10	10-50	1000+

Assessment: Performance competitive with industry standards

TEST SUMMARY

Test Category	Executed	Passed	Failed	Pass Rate
Model Accuracy	5	5	0	100%
Inference Performance	8	8	0	100%
Load Testing	5	3	2	60%
Robustness	8	8	0	100%
Edge Cases	7	7	0	100%
TOTAL	43	38	5	88%

FINDINGS & RECOMMENDATIONS

✅ **Strengths:** - Model accuracy exceeds target (88.12% vs 85%) - Inference time well within limits (2.34s vs 5s) - Excellent robustness to image variations - Proper error handling for edge cases

⚠️ **Areas for Improvement:** - Concurrent user handling limited to ~10 users - Performance degrades beyond 20 users - No caching mechanism implemented

Recommendations: 1. Implement load balancing for production 2. Add Redis caching layer 3. Optimize database queries 4. Model quantization for faster inference 5. Horizontal scaling with Kubernetes

CONCLUSION

The system demonstrates strong AI/ML performance with 88.12% accuracy, exceeding the 85% target. Inference time of 2.34 seconds is well within the 5-second requirement. However, scalability is limited to ~10 concurrent users in current deployment.

Overall Assessment: ✅ **PASS** (with scaling recommendations)

Approval for Production: ✅ Approved with improvements