

Evaluating CNN Depth for Multi-Class Bone Tumor Tissue Classification in Histopathology Images

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Abstract: This study explores how convolutional neural network (CNN) depth affects the classification of bone tumor tissues in histopathological images. We analyze lightweight custom models versus deeper pre-trained ones, using techniques like contrast enhancement and data augmentation. A shallow CNN achieved **87.44% accuracy**, outperforming deeper models, indicating that optimized lightweight architectures can deliver efficient and accurate multi-class classification.

METHODOLOGY: Dataset

Source: **TCIA Osteosarcoma Dataset[1]**,

Annotated by expert pathologists

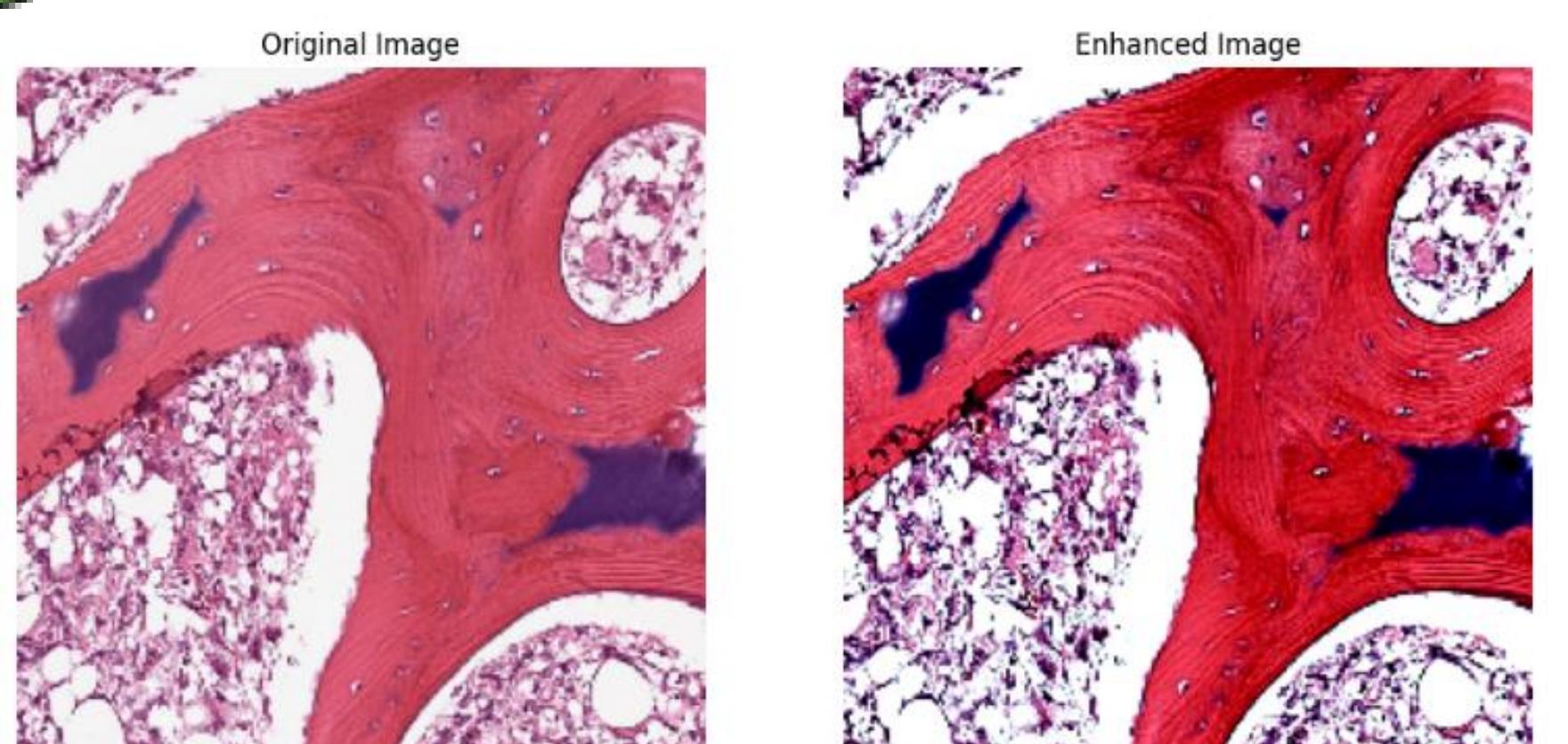
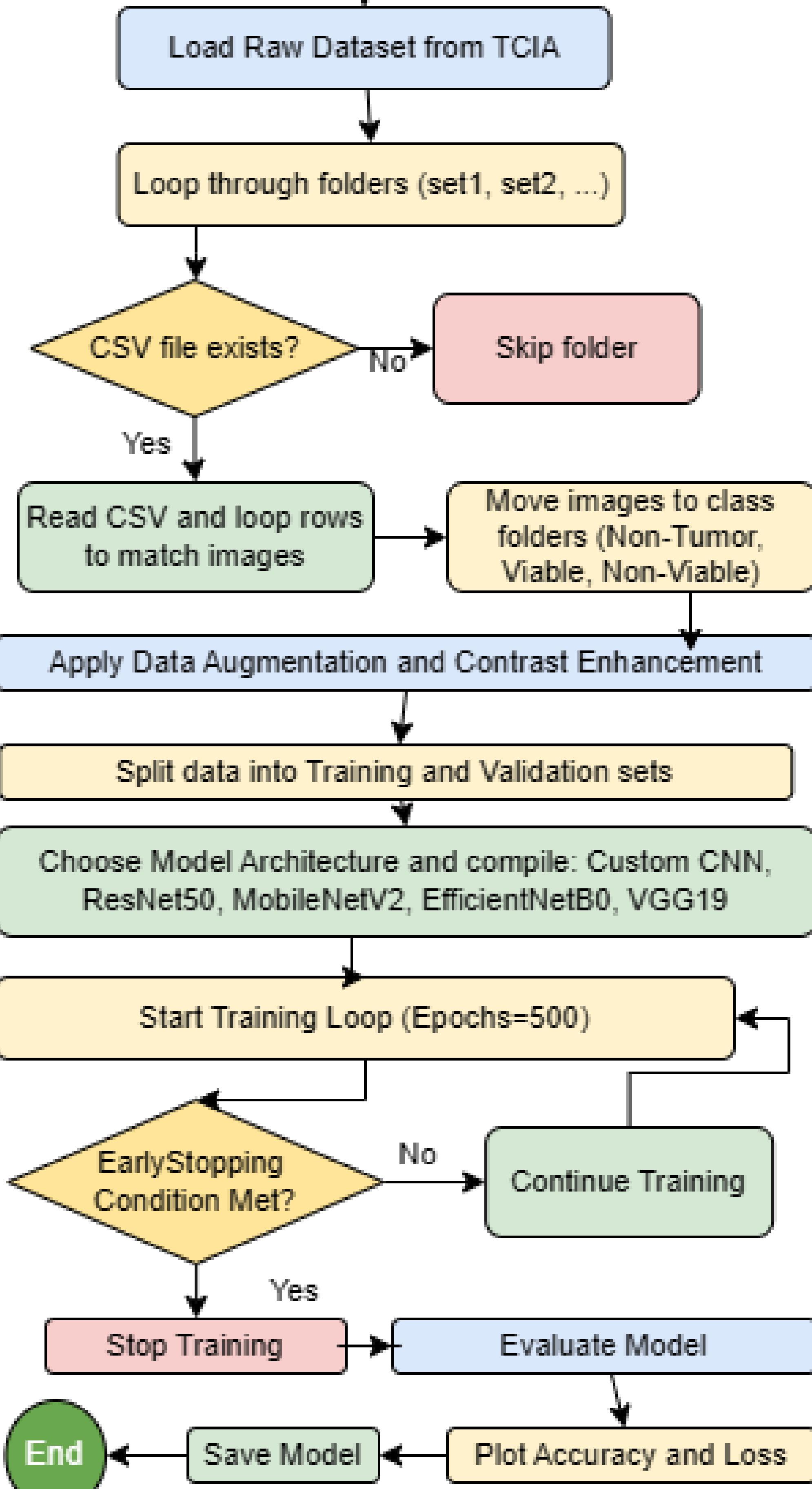
Images: **1,144 H&E-stained tiles** from 50 patients

Classes: Non-Tumor: 536 (47%)

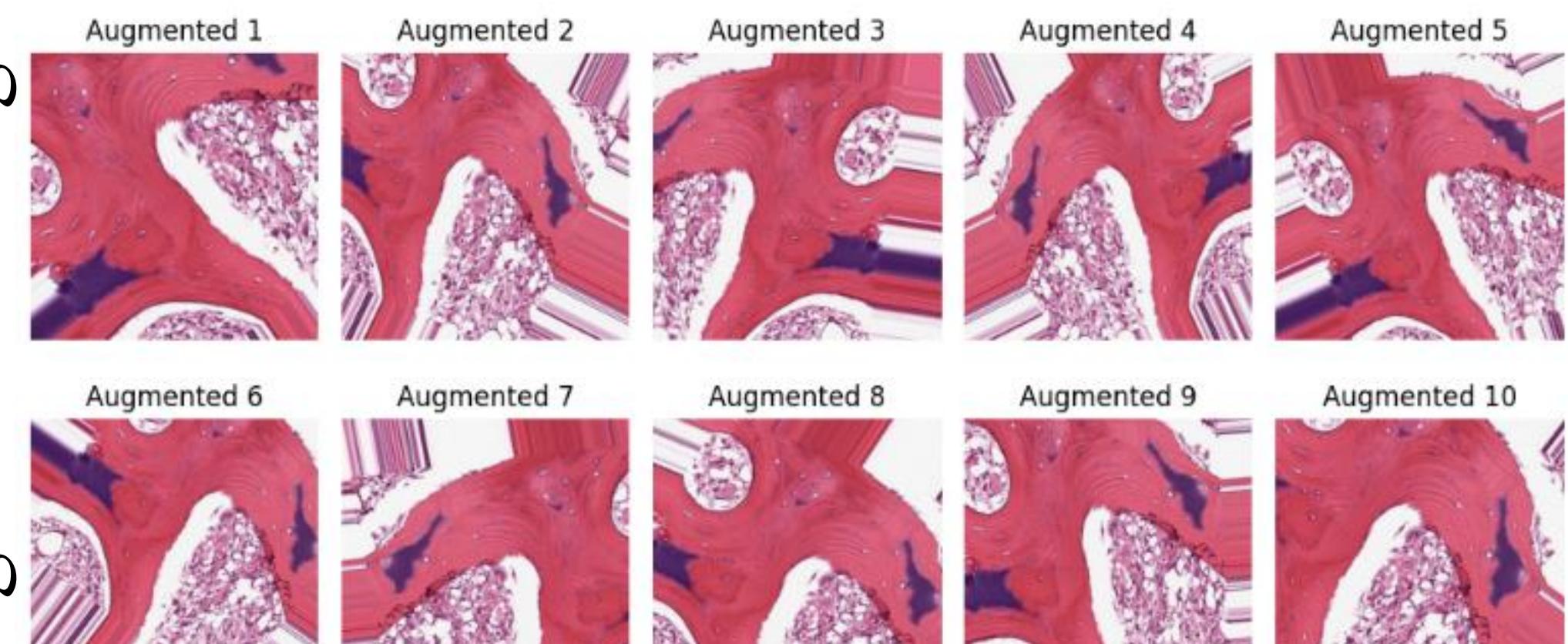
Viable Tumor: 345 (30%)

Necrotic Tumor: 263 (23%)

Start



Original and contrast image of "for Case-48-P5-C16-12239-11981.jpg"



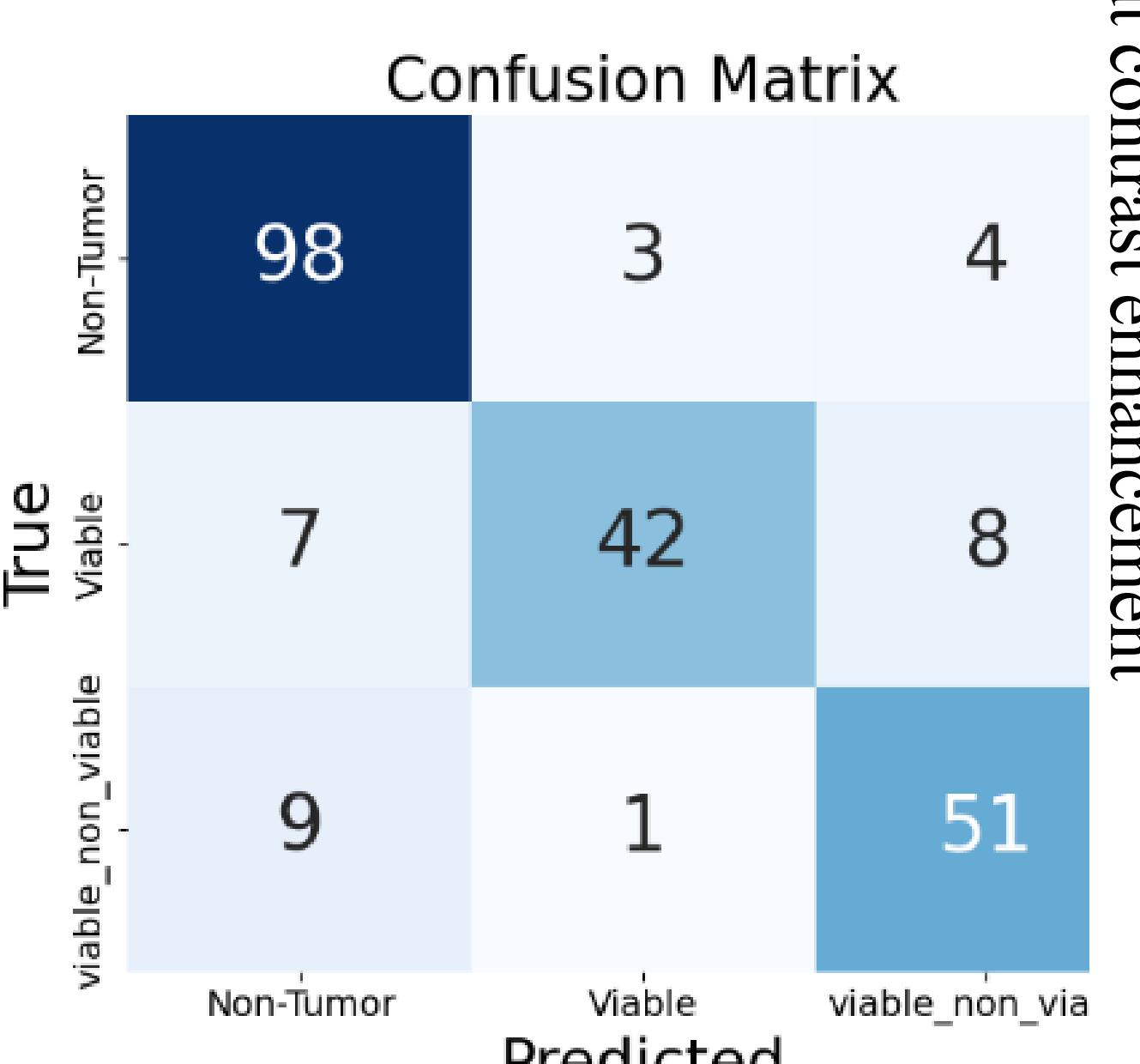
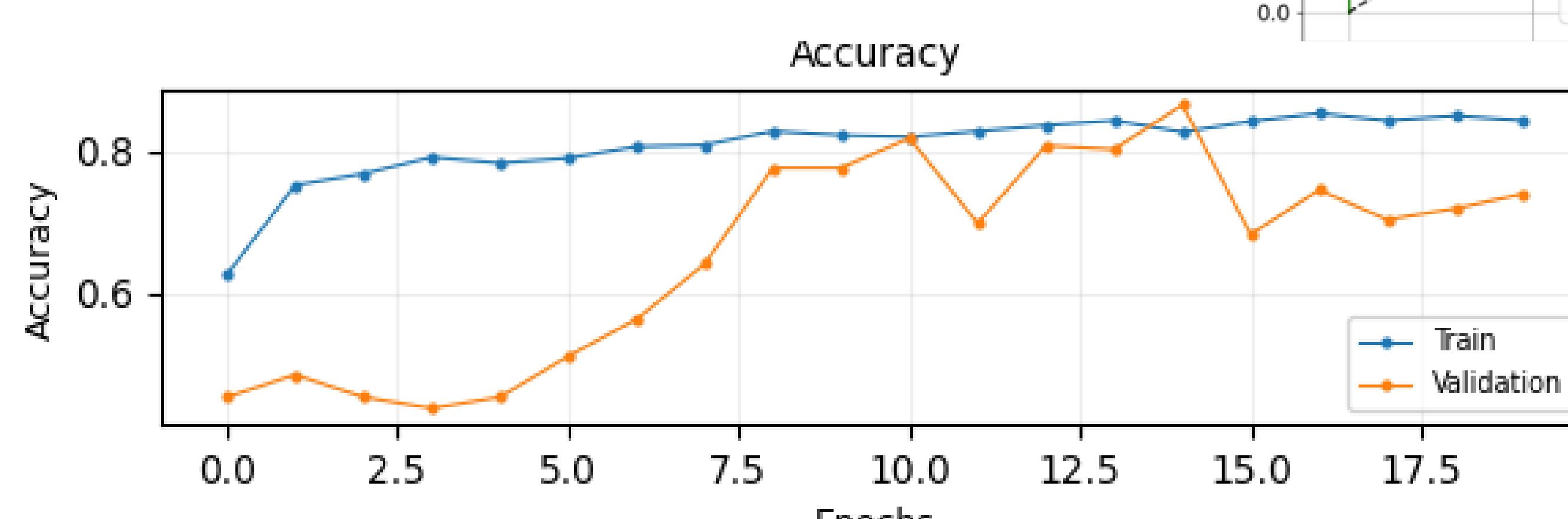
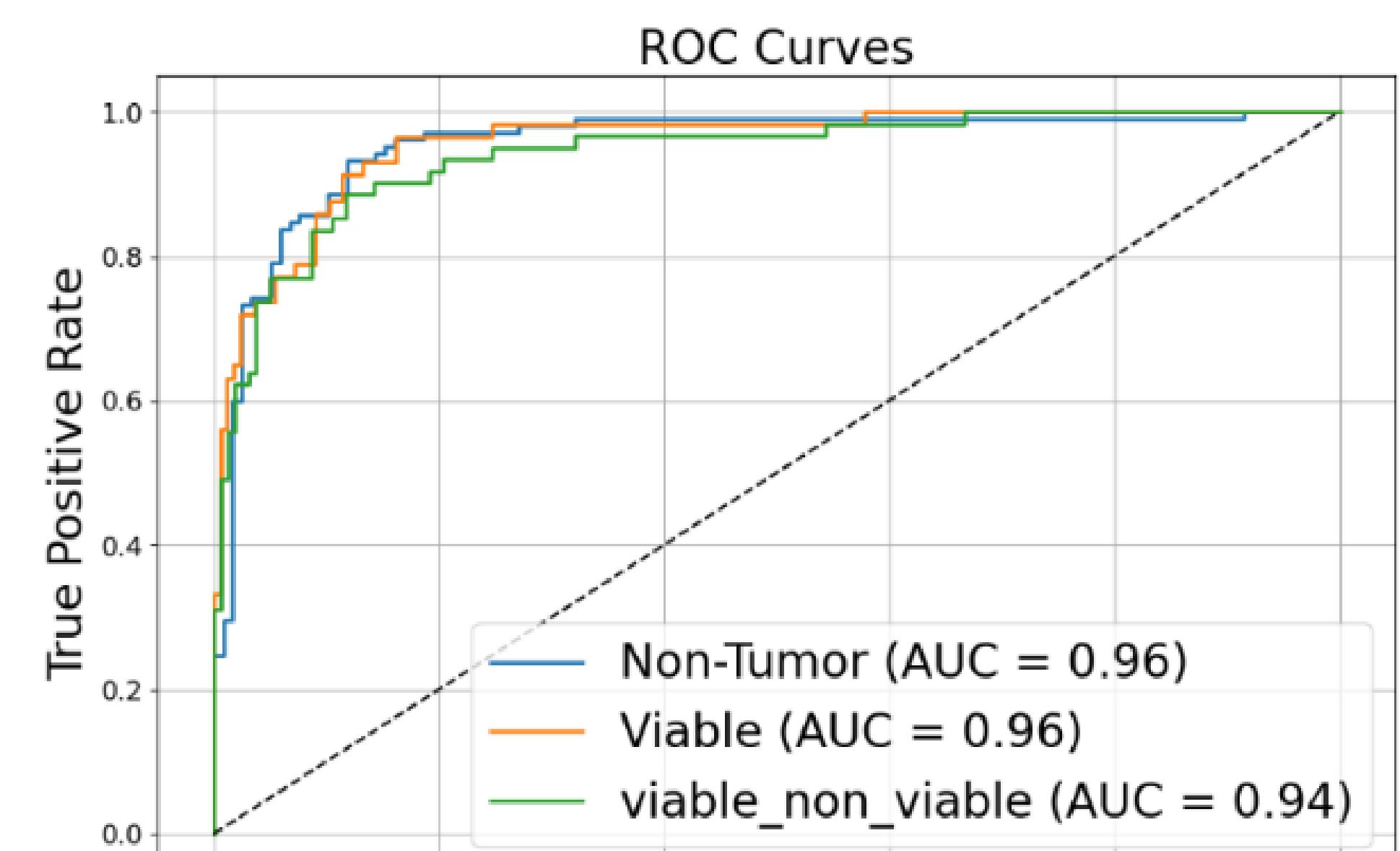
RESULTS AND DISCUSSION

Data Augmentation Strategies and Medical Justifications with performance evaluation

#	Rotation	Width Shift	Height Shift	Zoom	Flip	Validation Loss	Validation Accuracy	Medical Justification
1	10°	0.1	0.1	0.1	✗	1.4183	0.4709	Mild variation simulating scanner misalignment
2	20°	0.2	0.2	0.2	✓	0.6479	0.7534	Moderate variability; flip handles orientation issues
3	30°	0.3	0.3	0.3	✓	0.3594	0.8744	Strong regularization to reduce overfitting
4	0°	0.0	0.0	0.0	✗	0.6253	0.7758	Baseline with no augmentation
5	0°	0.0	0.3	0.1	✗	1.0264	0.4709	Mimics vertical slice shifts in imaging
6	5°	0.05	0.05	0.4	✗	Unable to process		Focused zoom simulates closeness to ROI (tumor core), useful in histology.
7	0°	0.0	0.0	0.0	✓	0.3960	0.8475	Flip only; safe for symmetric histology structures
8	40°	0.0	0.0	0.1	✗	0.3675	0.8565	Enhances rotational invariance for irregular shapes
9	8°	0.05	0.05	0.05	✓	0.4141	0.8475	Gentle, general-purpose augmentation
10	5°	0.4	0.4	0.0	✗	0.4374	0.8161	Strong translation to test spatial variability

Different Model Performance	Contrast Enhancement	Architecture	Early Stopped Epoch	Validation Loss	Validation Accuracy
	✗	Custom 2-Block CNN	20	0.3594	0.8744
	✓	Custom 2-Block CNN	19	0.4116	0.8475
	✗	Custom 3-Block CNN	19	1.0579	0.4709
	✓	Custom 3-Block CNN	13	1.0583	0.4709
	✗	ResNet50	16	0.5403	0.7982
	✓	ResNet50	18	0.5734	0.7631
	✗	MobileNetV2	16	0.5956	0.7758
	✓	MobileNetV2	10	0.7728	0.6637
	✗	EfficientNetB0	11	0.6180	0.6951
	✓	EfficientNetB0	10	0.7062	0.6323
	✗	VGG19	8	0.6304	0.7309
	✓	VGG19	24	1.0019	0.7354

This study shows that a lightweight CNN with targeted augmentation (30° rotation, 30% shift, zoom, and flipping) achieved 87.44% accuracy, demonstrating its effectiveness for assisting pathologists in osteosarcoma tissue classification.



References: [1]Clark, K., et al. (2013). The Cancer Imaging Archive (TCIA): maintaining and operating a public information repository. Journal of Digital Imaging, 26(6), 1045–1057.