

Abnormality Detection of Upper Extremity X-Rays

Gargee Jagtap, Ali Rivera, Anne Louise Seekford



Musculoskeletal Radiographs

“The AAMC (Association of American Medical Colleges) predicts there will be a shortage of 17,000 - 42,000 radiologists in the next decade”

Motivation

- Radiologist shortage
- 1.7 billion people worldwide are affected by musculoskeletal conditions
- X-Rays are the most common diagnostic technique
- Radiology imaging delays were an independent predictor of the length of a patient's hospital stay



Data



Image Info

patient No.
study No.
label
image No.



Abnormality Label

Each image is labeled as
either:

0: *Normal*
1: *Abnormal*



Body part

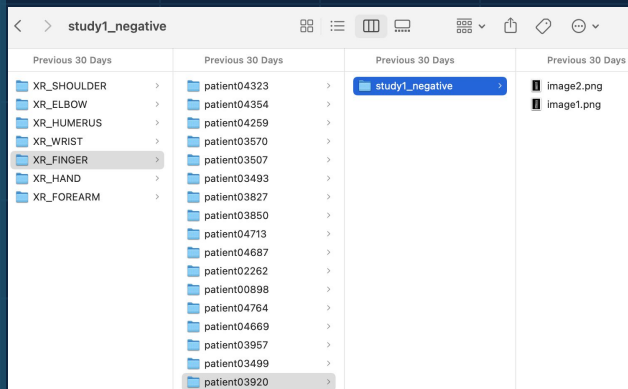
Elbow
Finger
Forearm
Hand
Humerus
Shoulder
Wrist

Data Preprocessing



Cleaning

- Nested folders
- Labels located in image file name



Transformations

- Resized: 256x256
- Random Horizontal Flip
- Normalized :
 $(\mu, \sigma) = (0.5, 0.5)$



Train

29,583 images (74%)

Validation

3,197 images (8%)

Test

7,395 images (18%)

path	label	body_part
patient11511_study1_positive_image2.png	1	hand
patient11511_study1_positive_image1.png	1	hand
patient11371_study1_negative_image3.png	0	hand
patient11371_study1_negative_image2.png	0	hand
patient11371_study1_negative_image1.png	0	hand
...
patient11359_study1_positive_image3.png	1	forearm
patient11359_study1_positive_image1.png	1	forearm
patient11392_study1_positive_image3.png	1	forearm
patient11392_study1_positive_image2.png	1	forearm
patient11392_study1_positive_image1.png	1	forearm

Model Pipeline



Stage 1

Body Part Classification



HUMERUS



HAND



SHOULDER



Stage 2

Normal/Abnormal Classification



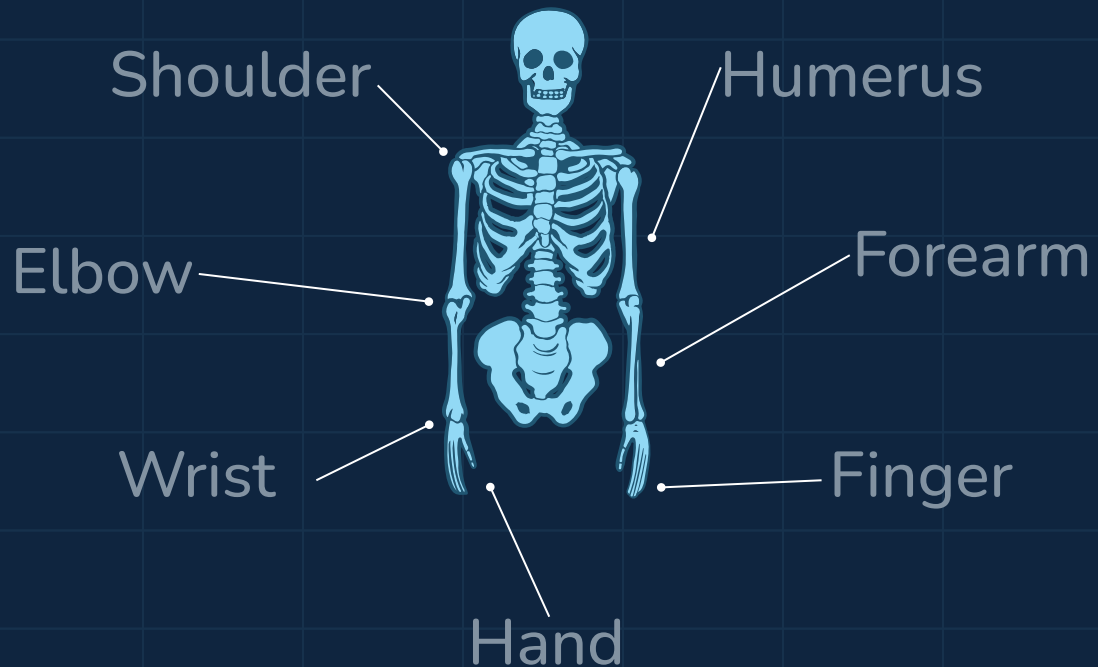
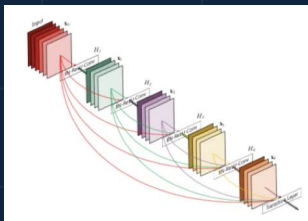
0: NORMAL



1: ABNORMAL

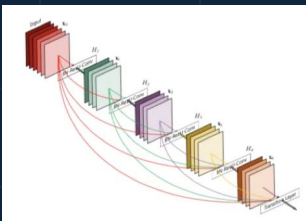
Stage 1

- Pretrained: DenseNet121
- Linear Layer input: 2048
- No. of predicted classes: 7
- Optimizer: Adam
- Loss Function: Cross Entropy
- Learning Rate: 0.001
- 15 Epochs

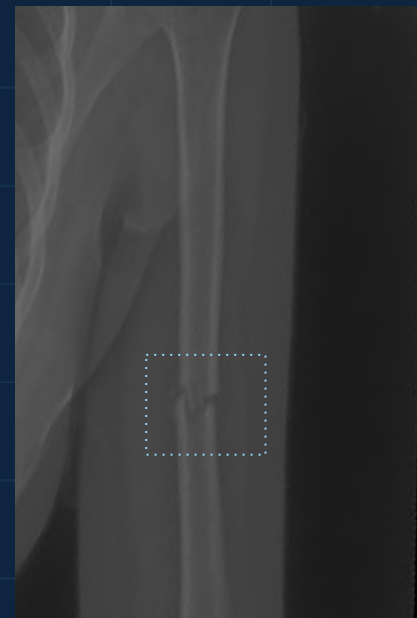


Stage 2

- Pretrained: DenseNet121
- Linear Layer input: 2048
- No. of predicted classes: 2
- Optimizer: Adam
- Loss Function: Cross Entropy
- Learning Rate: 0.001
- 15 Epochs



0: NORMAL



1: ABNORMAL

Previous Results

Agreement Measured using Cohen's Kappa Statistic

	Radiologist 1	Radiologist 2	Radiologist 3	Model
Elbow	0.850 (0.830, 0.871)	0.710 (0.674, 0.745)	0.719 (0.685, 0.752)	0.710 (0.674, 0.745)
Finger	0.304 (0.249, 0.358)	0.403 (0.339, 0.467)	0.410 (0.358, 0.463)	0.389 (0.332, 0.446)
Forearm	0.796 (0.772, 0.821)	0.802 (0.779, 0.825)	0.798 (0.774, 0.822)	0.737 (0.707, 0.766)
Hand	0.661 (0.623, 0.698)	0.927 (0.917, 0.937)	0.789 (0.762, 0.815)	0.851 (0.830, 0.871)
Humerus	0.867 (0.850, 0.883)	0.733 (0.703, 0.764)	0.933 (0.925, 0.942)	0.600 (0.558, 0.642)
Shoulder	0.864 (0.847, 0.881)	0.791 (0.765, 0.816)	0.864 (0.847, 0.881)	0.729 (0.697, 0.760)
Wrist	0.791 (0.766, 0.817)	0.931 (0.922, 0.940)	0.931 (0.922, 0.940)	0.931 (0.922, 0.940)
Overall	0.731 (0.726, 0.735)	0.763 (0.759, 0.767)	0.778 (0.774, 0.782)	0.705 (0.700, 0.710)

Results

Stage 1 Accuracy:

97.8%

Stage 2 Accuracy:

Body Part	Elbow	Finger	Forearm	Hand	Humerus	Shoulder	Wrist
Accuracy	81.5%	76.6%	72.1%	79.9%	73.4%	76.6%	83.1%

Conclusion

Compare Performances:

	Their Model:	Our Model:
Best Performance:	<i>Finger, Hand, Wrist</i>	<i>Wrist, Elbow</i>
Worst Performance:	<i>Forearm, Humerus, Elbow</i>	<i>Forearm, Humerus</i>

Future Steps:

- Get more data for body parts
- Further customize models for each bodypart

References

Alkoby. (2023, January). Alkoby/bone-fracture-detection: Bone Fracture Detection Using Deep Learning (RESNET50) - final project in the fourth year of the degree. GitHub. Retrieved April 17, 2023, from <https://github.com/Alkoby/Bone-Fracture-Detection>

Alzubaidi, M. S., Shah, U., Zubaydi, H. D., Dolaat, K., Abd-Alrazaq, A. A., Ahmed, A., & Househ, M. (2021). The Role of Neural Network for the Detection of Parkinson's Disease: A Scoping Review. *Healthcare*, 9(6). <https://doi.org/10.3390/healthcare9060740>

American Academy of Orthopedic Surgeons. (2017, June). *X-rays, CT scans, and MRI scans*. OrthoInfo. Retrieved March 15, 2023, from [https://orthoinfo.aaos.org/en/treatment/x-rays-ct-scans-and-mris/#:~:text=X%2Drays%20\(radiographs\)%20are.get%20an%20X%2Dray%20first](https://orthoinfo.aaos.org/en/treatment/x-rays-ct-scans-and-mris/#:~:text=X%2Drays%20(radiographs)%20are.get%20an%20X%2Dray%20first)

MetaAI. (n.d.). *Papers with Code - MURA dataset*. Papers With Code. Retrieved March 15, 2023, from <https://paperswithcode.com/dataset/mura>

Mina S. Makary, M. D., & Noah Takacs, B. S. (2022, January 20). *Are We Prepared for a Looming Radiologist Shortage?* Diagnostic Imaging. Retrieved March 15, 2023, from <https://www.diagnosticimaging.com/view/are-we-prepared-for-a-looming-radiologist-shortage->

Rajpurkar, P., Irvin, J., et al., (n.d.). *MURA: Large Dataset for Abnormality Detection in Musculoskeletal Radiographs*. Stanford ML Group. Retrieved March 15, 2023, from <https://stanfordmlgroup.github.io/competitions/mura/>

Tsang, S.-H. (2018, November 25). *Review: DenseNet - Dense Convolutional Network (Image Classification)*. Medium. Retrieved March 15, 2023, from <https://towardsdatascience.com/review-densenet-image-classification-b6631a8ef803>

Wang, X., Peng, Y., Lu, L., Lu, Z., Bagheri, M., & Summers, R. M. (2017, December 14). *Chestx-Ray8: Hospital-scale Chest X-ray Database and Benchmarks on Weakly-Supervised Classification and Localization of Common Thorax Diseases*. arXiv.org. Retrieved March 15, 2023, from <https://arxiv.org/abs/1705.02315>

World Health Organization. (2022, July 14). *Musculoskeletal health*. World Health Organization. Retrieved March 15, 2023, from <https://www.who.int/news-room/fact-sheets/detail/musculoskeletal-conditions#:~:text=Musculoskeletal%20conditions%20are%20typically%20characterized,form%20of%20non%2Dcancer%20pain>

Thanks!

Do you have any questions?

wra2iv@virginia.edu

wat6sv@virginia.edu

bng3be@virginia.edu

CREDITS: This presentation template was created by
Slidesgo, including icons by Flaticon, and
infographics & images by **Freepik**

