### Al Based Bone Age Prediction

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#### **Presentation Outline**

- Motivation
- Introduction
- Problem Statement & Objectives
- Scopes
- Applications

- Methodology
- Results
- Analysis/Discussion of Results
- Future Enhancements
- References

#### **Motivation**

- Costly Pre-existing software
- Variability in Interpretation
- Time-Consuming Process
- Limited Accessibility of specialized radiologists
- Risk of Human Error
- Lack of Standardization

### Objectives

• To implement an AI-based Bone age prediction model based on Left hand Xray Images.

### Why Left-Hand Radiographs only?

- Standardization
- Availability of reference data
- Convenience
- Contains Key Ossification centers

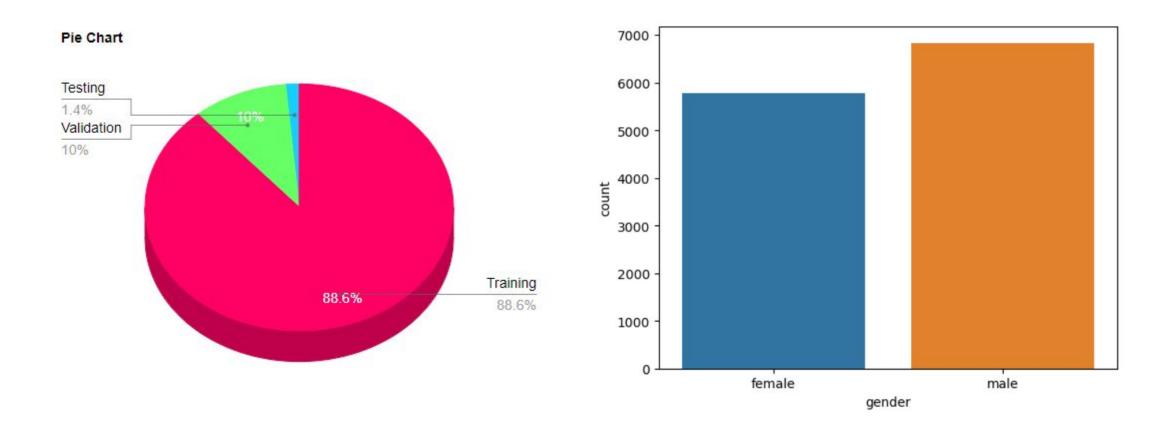
### Scopes

- Pediatric Growth Monitoring
- Orthopedics and Bone Health
- Puberty and Adolescence Assessment
- Skeletal Maturity test

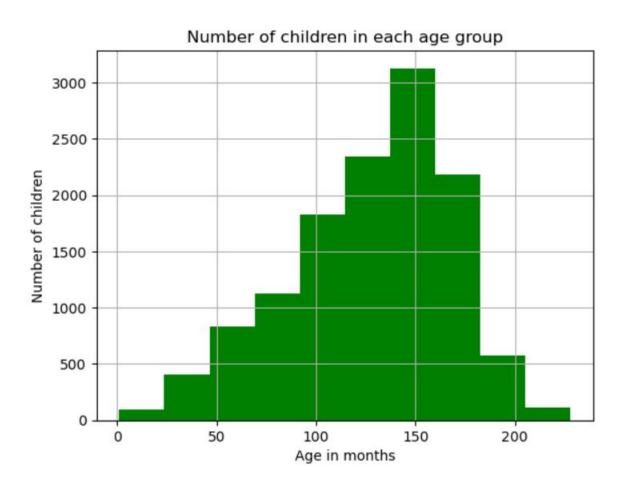
### **Applications**

- Educational tool enhancing understanding of skeletal development.
- Orthopedic surgery planning based on skeletal maturity.
- Clinical Decision support system.
- Determining the skeletal maturity of young athletes.
- Research and population health studies.

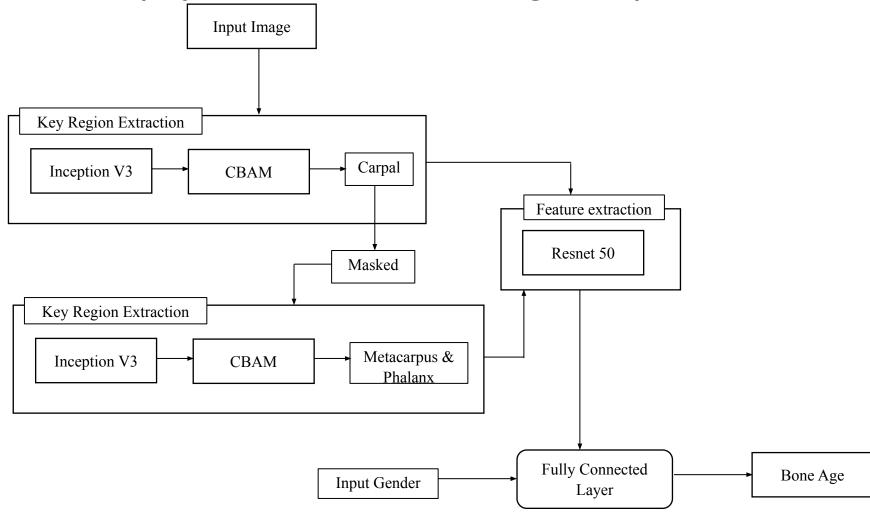
#### Dataset



#### Dataset



# Methodology-[1] (System Block Diagram)



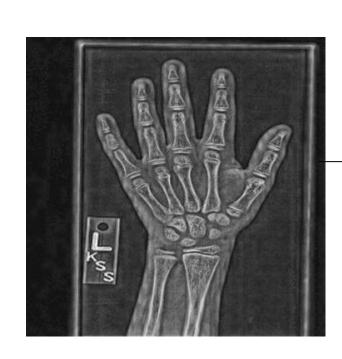
# Methodology-[2] Preprocessing

- Resize the image (Bicubic Interpolation)
- Enhance the image (CLAHE)
- Normalization

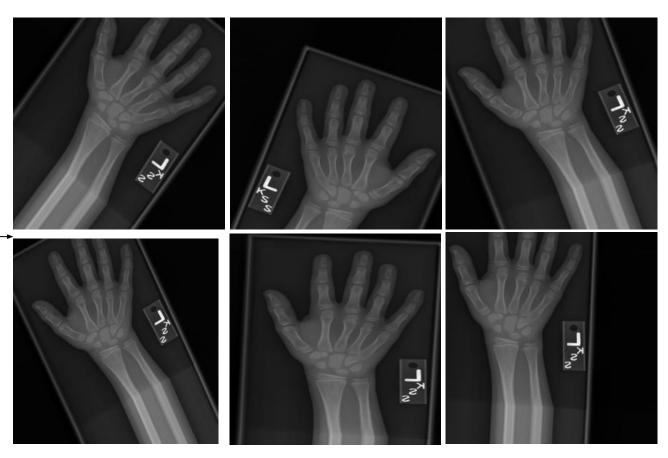
# Methodology-[3] Augmentation

- Rotation Range 40
- Width and Height Shift Range 0.2
- $\cdot$  Zoom -0.2
- Horizontal Flip True
- Fill mode Nearest

# Methodology-[4] Augmentation



Original

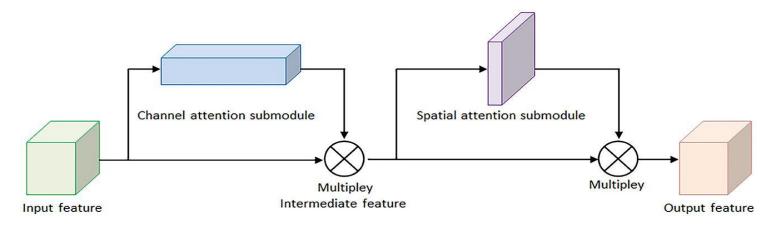


Augmented

# Methodology-[5] Inception V3

- Capture features at multiple scales in its inception modules.
- Use multiple filter sizes (1 X 1, 3 X 3, 5 X 5) and pooling operations in parallel.
- Introduces the factorized convolution.
- Incorporates auxiliary classifiers at intermediate layers.

# Methodology-[6] Convolutional Block Attention Module



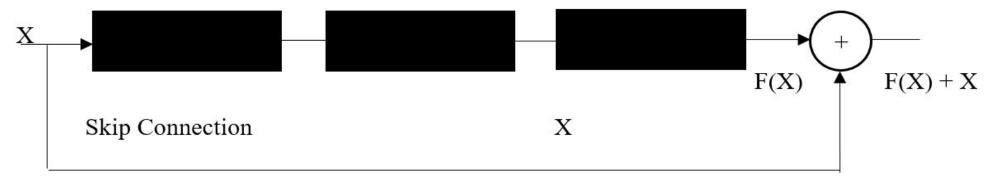
- CBAM with InceptionV3 adds an attention mechanism
- CAM improves feature representation by emphasizing important channels in feature map
- SAM enhances feature representation by assigning varying importance to different spatial regions within an image

# Methodology-[7] Masking

- Set threshold value (0.7)
- Identify Regions of interest
- Crops & Preserves regions

### Methodology-[8] ResNet50

- Convolutional operation
- Batch Normalization
- Relu Activation
- Residual Activation
- Pooling operation



### Result-[1]

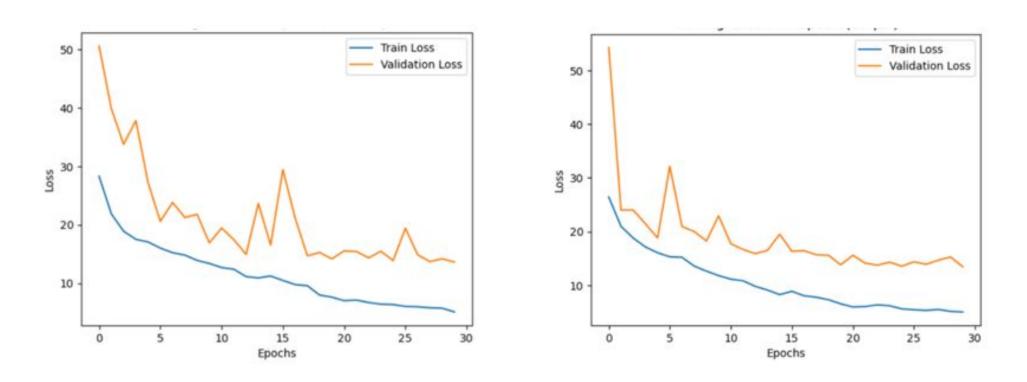


Fig: Training Loss For Metacarpal

Fig: Training Loss For carpal

### Result-[2]

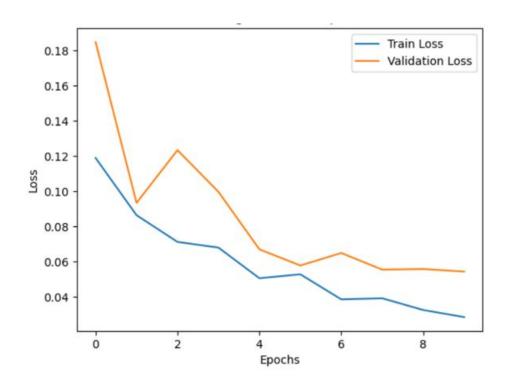
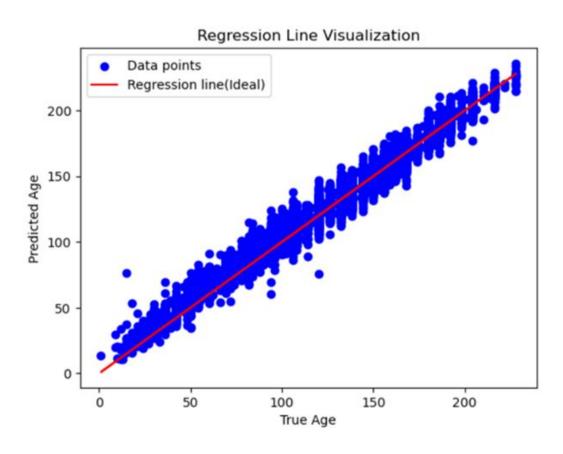


Fig: MAE vs Epochs (Combined Model)

### Result-[3]



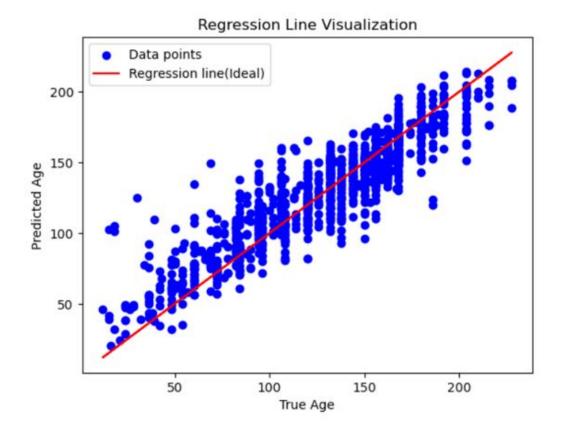
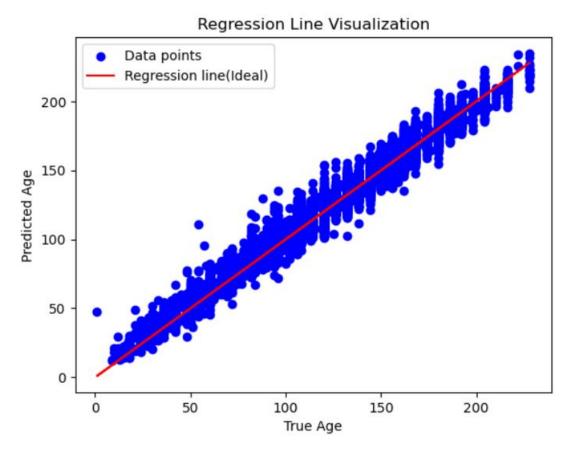


Fig: Predicted vs Actual bone age for Metacarpal(Training)

Fig: Predicted vs Actual bone age for Metacarpal(Validation)

### Result-[4]



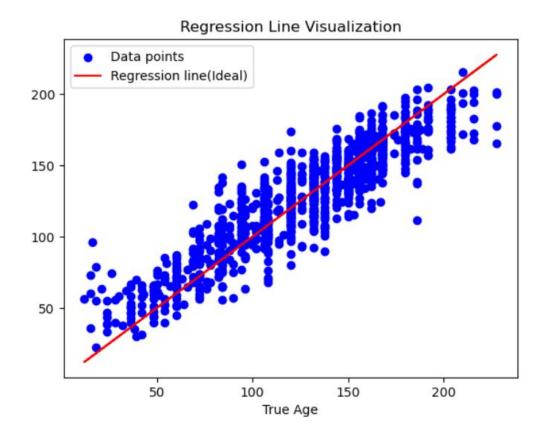
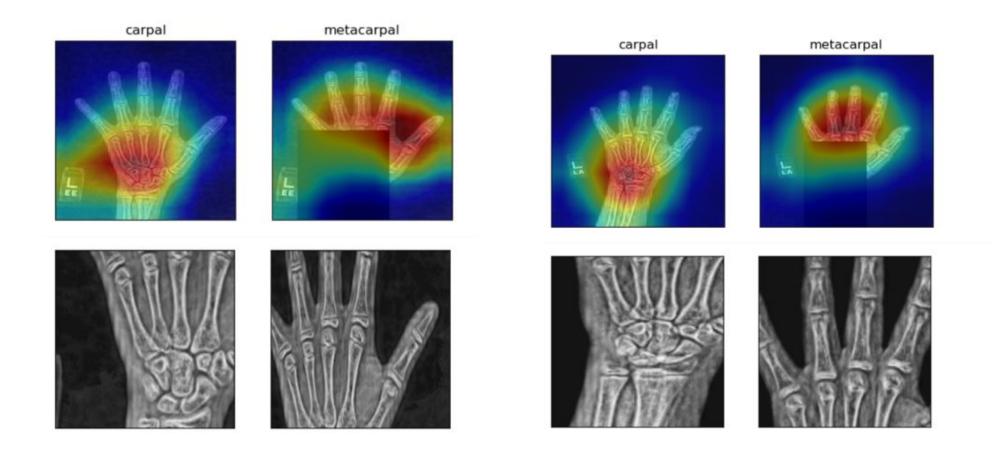


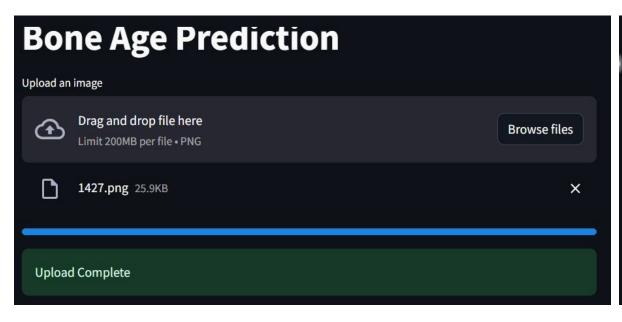
Fig: Predicted vs Actual bone age for carpal(Training)

Fig: Predicted vs Actual bone age for carpal(Validation)

# Results-[5]



### Results-[6]





## Results-[7]

•On the same testing dataset:

Our Model	Dr. Albert Model
R squared: 0.901167	R squared: 0.897288

### **Analysis-[1]**

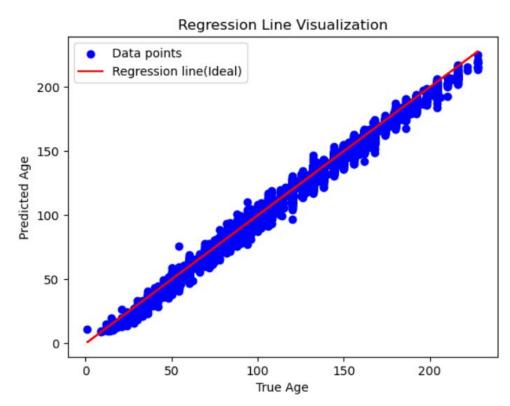


Fig: Regression Graph for Combined Training

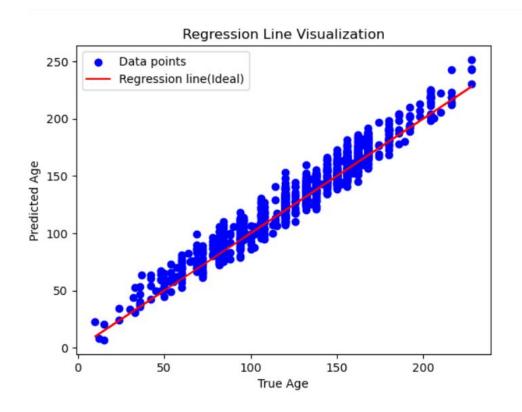


Fig: Regression Graph for Combined Validation

## Analysis-[2]

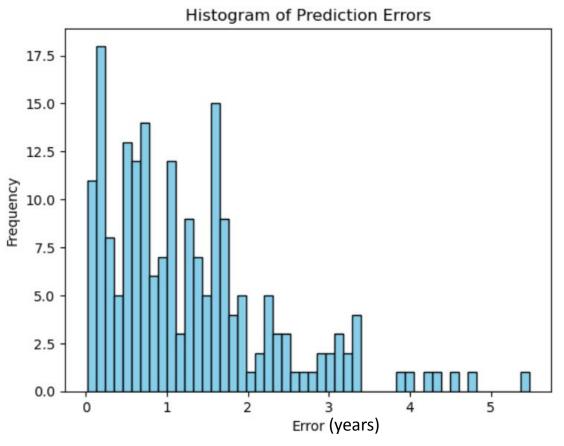


Fig: Histogram of Prediction Errors (Testing Dataset)

#### **Future Enhancements**

- Collection of Regional specific dataset for training.
- Newer Models like Vision Transformers can be implemented to further improve the model.
- Can be integrated in hospital service applications.

### References-[1]

- [1] H. H. Thodberg and S. Kreiborg, "The BoneXpert Method for Automated Determination of Skeletal Maturity," 2008.
- [2] W. Eric, K. Bin and W. Xin, "RESIDUAL ATTENTION BASED NETWORK FOR HAND BONE AGE ASSESSMENT," 2017.
- [3] V. Iglovikov, A. Rakhlin, K. A. and S. A., "Pediatric Bone Age Assessment Using Deep convolutional Newral Newworks," *International Workshop on Deep Learning in Medical Image Analysis*, 2017.
- [4] M. Escobar, F. Torresl and D. L., "Hand pose estimation for pediatric bone age assessment," *Internation Conference on Medical Image Computing And Computer-Assisted Intervention*, pp. 531-539, 2017.
- [5] W. Chong and W. Yang, "Attention-based multiple-instance learning for pediatric bone age assessment with efficient and interpretable," *Frontiers*, 2023.

### References-[2]

- [6] S. Ibrahim and H. A.Ben, "Ridge Regression Neural Network for Pediatric Bone Age Assessment," *Concordia Institure for Information Systems Engineering*, pp. 4-5, 2021.
- [7] P. Ewa, G. Arkadiusz and P. Sylwia, "Computer-Assisted Bone Age Assessment: Image Preprocessing and Epiphyseal/ Metaphyseal ROI Extraction," *IEEE Transactions on medical imaging,* pp. 3-6, 2001.
- [8] R. Aravinda, P. Sameena and A. Tanweer, "Pediatric Bone Age Assesment using Deep Learning Models," *Manipal Institute of Technology*, 2017.
- [9] L. Hyunkwang, T. Shahein, L. Jenny and Z. Maurice, "Fully Automated Deep Learning System for Bone Age Assessment," Massachusetts General Hospital and Harvard Medical School, 2017.
- [10] M. Satoh and K. Xin, "J-stage," 2015. [Online]. Available: https://www.jstage.jst.go.jp/article/cpe/24/4/24\_2015-0011/\_article#:~:text=The%20main%20bone%20age%20assessment,resonance %20(MR)%20imaging%20methods..
- [11] H. K. Pyeong, M. Y. Hee and R. K. Jeong, "Bone Age Assessmet Using Artificial Intelligence in Korean Pediatric Population," *Korean Journal*, 2023.
- [12] C. Szegedy, V. Vanhoucke, S. Ioffe, J. Shlens and Z. and Wojna, ""Rethinking the Inception Architecture for Computer Vision,"," in *CVPR*, Los Vegas, 2016.