



Bone Age Detection from X-ray

Group 01

Chirag Soni (1401CS13)

Vignesh Edithal (1401CS14)

Prashant (1401CS35)

Introduction

1. Radiologists manually find skeletal age with help of atlas.
2. Compare skeletal age with actual age to ensure normal growth in children.
3. RSNA held competition to automate the process.



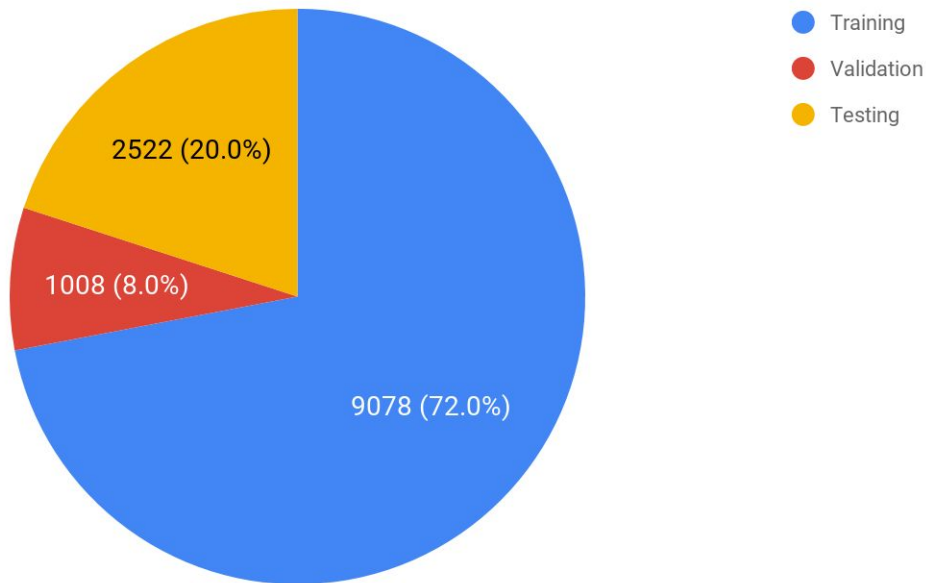
Dataset



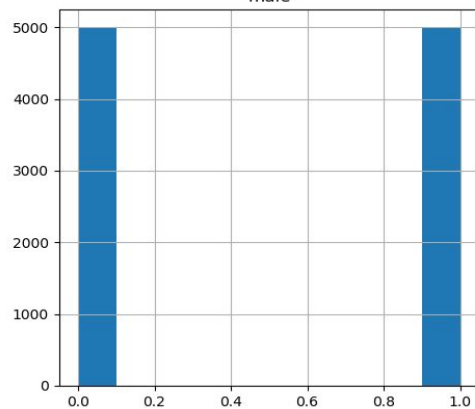
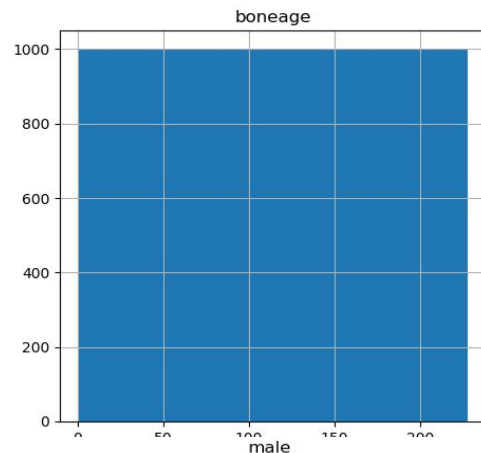
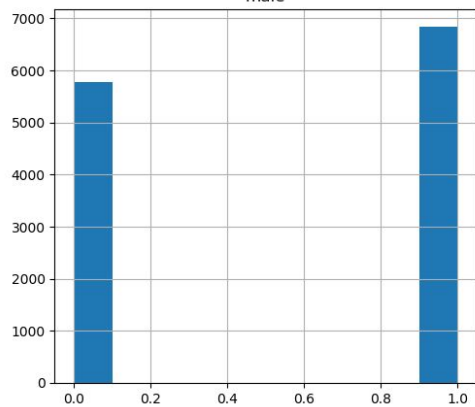
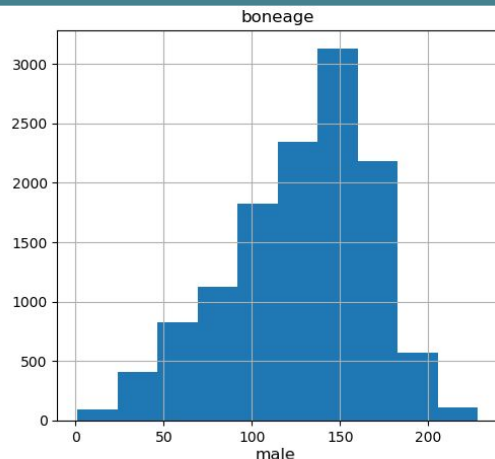
1.png

id	boneage	male
1	180	False
2	12	True
...

Total Images = 12611



Balancing Dataset



Preprocessing

1. Age normalization $(x - \mu)/\sigma$
 - Reduces effect of outlier data points
 2. Reduce Image Resolution $2000 \times 1500 \rightarrow 224 \times 224$
 - Reduce redundant pixels while retaining information.
 3. Rotation 5°
 4. Horizontal/Vertical Translation 15%
 5. Zoom 25%
 6. Horizontal Flip
- } For better generalization

Architecture

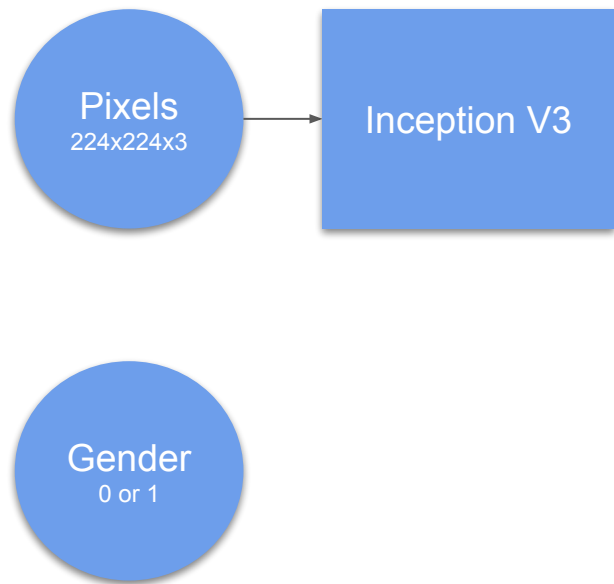


A diagram illustrating the input features for a model. It consists of two blue circles arranged vertically. The top circle is labeled 'Pixels' with '224x224x3' below it. The bottom circle is labeled 'Gender' with '0 or 1' below it.

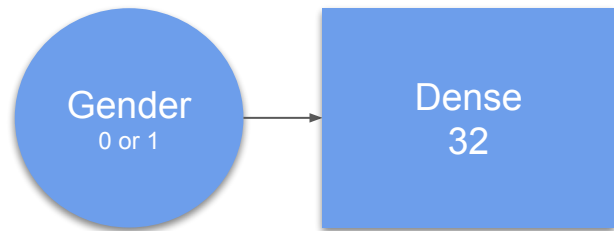
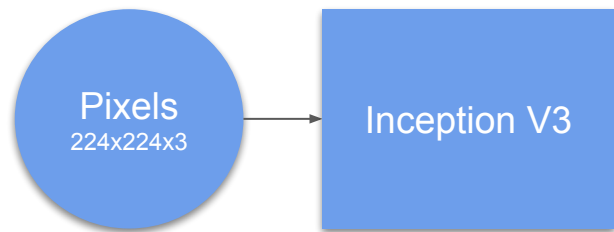
Pixels
224x224x3

Gender
0 or 1

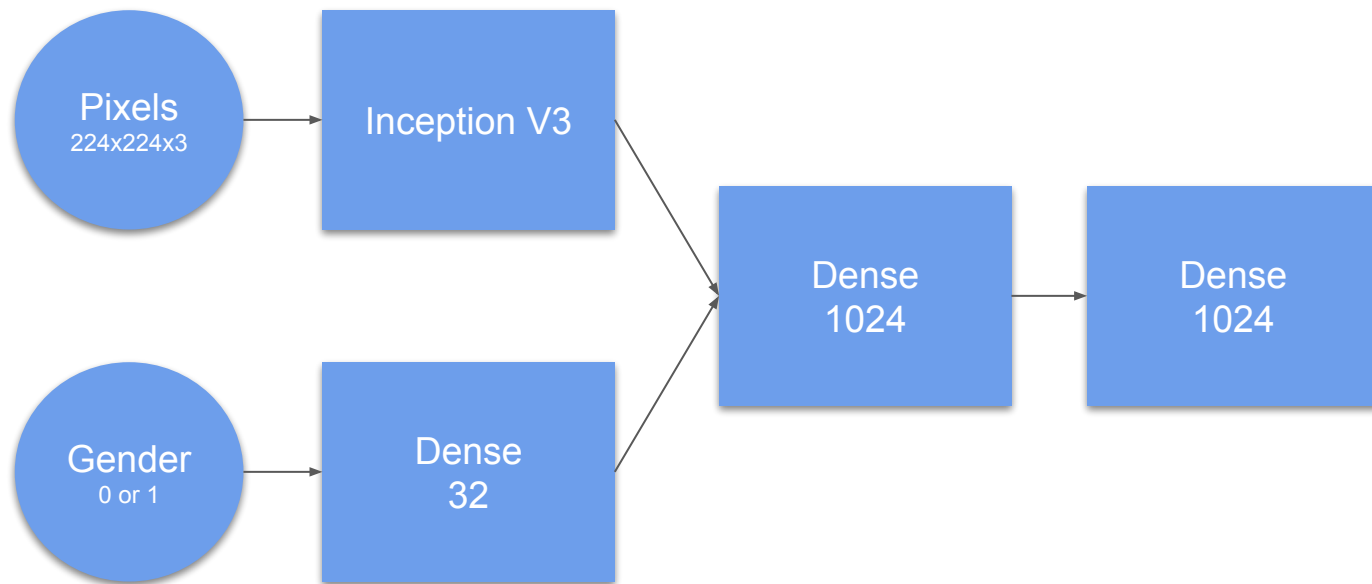
Architecture



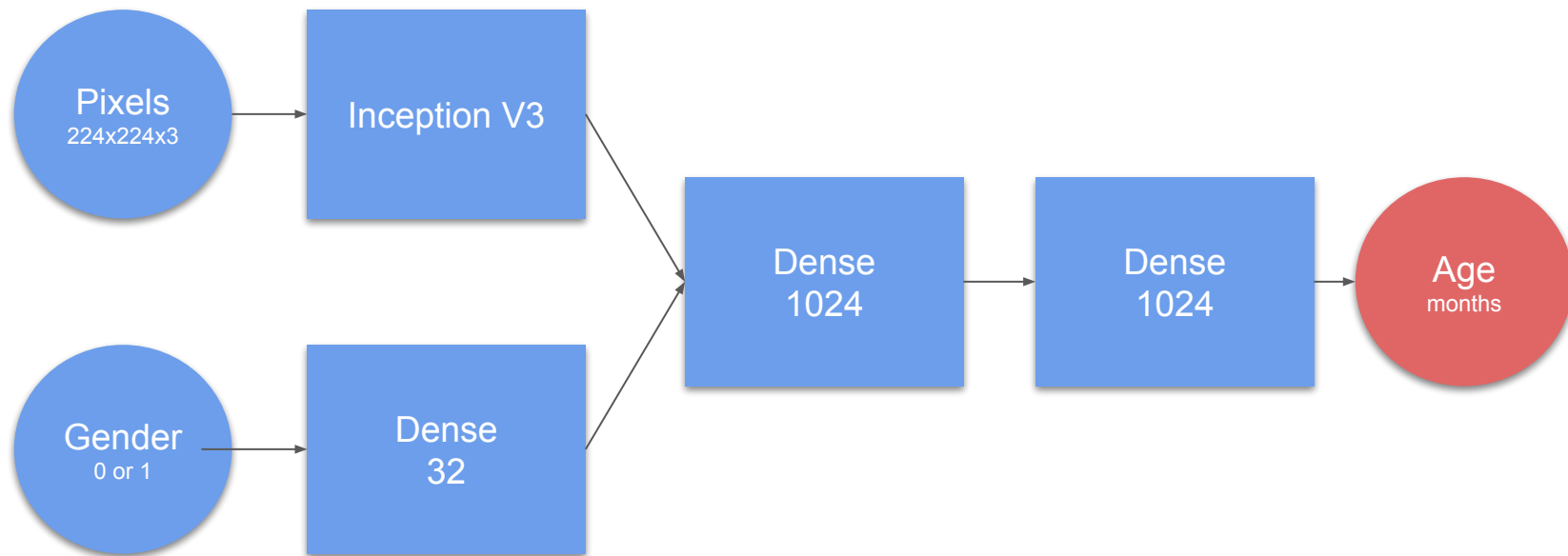
Architecture



Architecture



Architecture



Hyperparameters

Optimized over these parameters

- Pooling : Average pooling / Max pooling
- Dropout : 0.2 / 0.5
- Activation function : ReLU / tanh
- Gender : Use or not
- Image Size : 224x224, 500x500
- Pretrained models : Inception V3, VGG16, MobileNet

Training

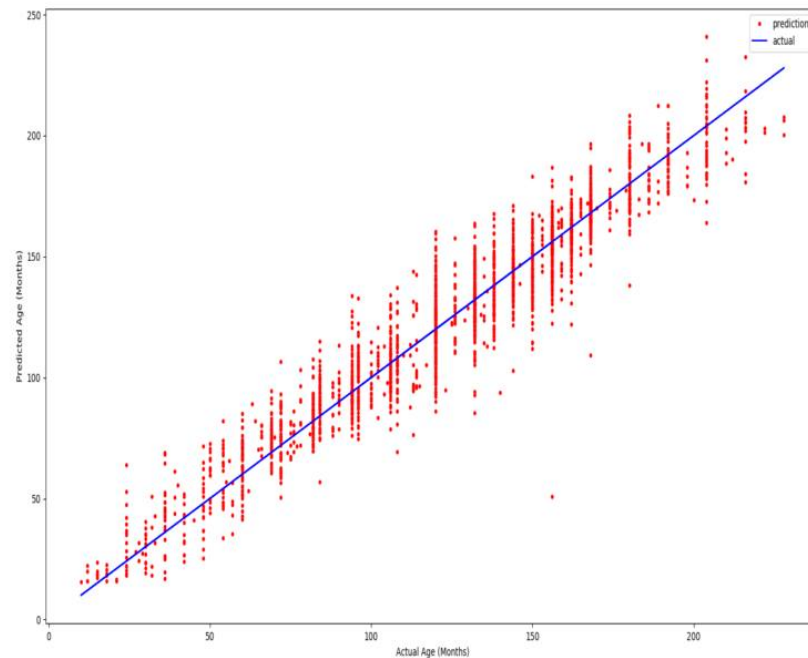
- Batchsize of 10
- Trained over 50 epochs (early stoppage with patience 10)
- Adam optimizer
- Mean Square Error as loss function
- Reduce Learning Rate on Plateau

Results

For the best model,

Test Mean Absolute Error = 8.615 months

- train_loss = 0.0809
- train_mae_months = 9.0737
- valid_loss = 0.0709
- val_mae_months = 8.1923



Code is uploaded on Github

<https://github.com/chiragiitp/Bone-Age-Detection-From-X-Ray>

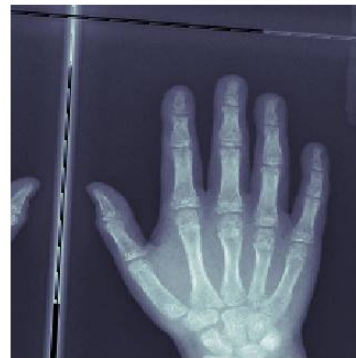
Age: 10.0
Predicted Age: 15.6
Gender: Female



Age: 82.0
Predicted Age: 81.3
Gender: Female



Age: 106.0
Predicted Age: 110.1
Gender: Female



Age: 126.0
Predicted Age: 129.8
Gender: Female



Age: 144.0
Predicted Age: 121.9
Gender: Male



Age: 156.0
Predicted Age: 146.3
Gender: Male



Age: 168.0
Predicted Age: 170.0
Gender: Male



Age: 228.0
Predicted Age: 200.3
Gender: Male



Conclusion

1. Gender information is vital for the network.
2. Mean Average Deviation (MAD) between different human observers 0.61 years. MAD with the winning model is 0.36 years.
3. Deep CNN can surpass human level efficiency.

Should we stop training Radiologists now?

References

- Kaggle [!\[\]\(cd3e54d951a9fb854f48e4697cf550f9_img.jpg\)](#)
- RSNA challenge [!\[\]\(cc729e263f29c0a76fbdc4cfe67fceb0_img.jpg\)](#)
- Performance of a Deep-Learning Neural Network Model in Assessing Skeletal Maturity on Pediatric Hand Radiographs [!\[\]\(90d36d418f8f7ab67431ba2525e00a5e_img.jpg\)](#)
- Lee, H., Tajmir, S., Lee, J. et al. J Digit Imaging (2017) 30: 427. Fully Automated Deep Learning System for Bone Age Assessment [!\[\]\(f70e40faeec369ff477dbaef549ee05b_img.jpg\)](#)

Thank you

