

1. Sample Estimation



Dataset: MORPH Real Age: 47 Esti. Age: 46.999



Dataset: MORPH Real Age: 22 Esti. Age: 22.216

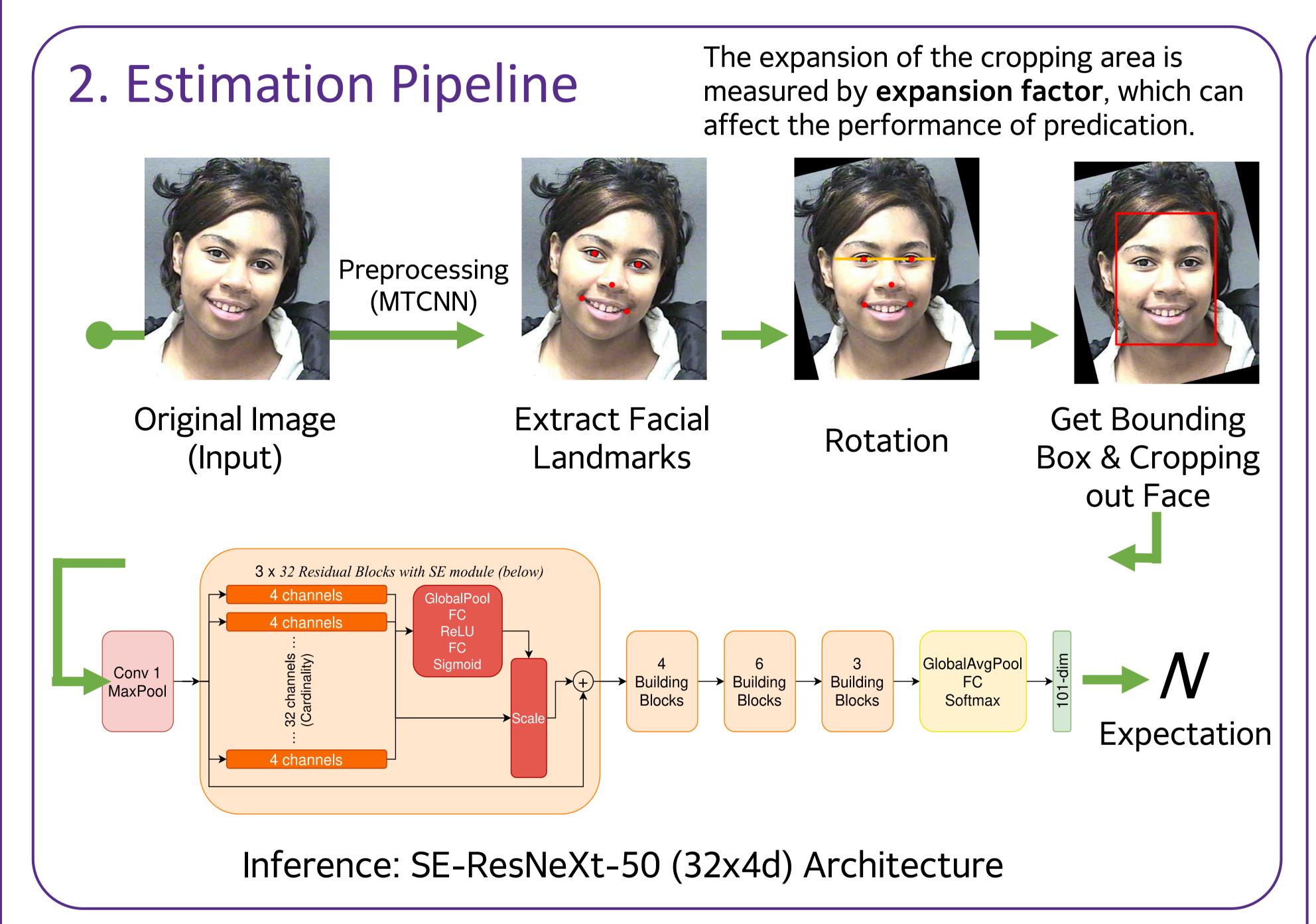


Dataset: MegaAge-Asian

Real Age: 28 Esti. Age: 28.000

DURF 2020 Yijian Liu Computer Science Supervised by Li Guo

Facial Age Estimation with Deep Neural Network

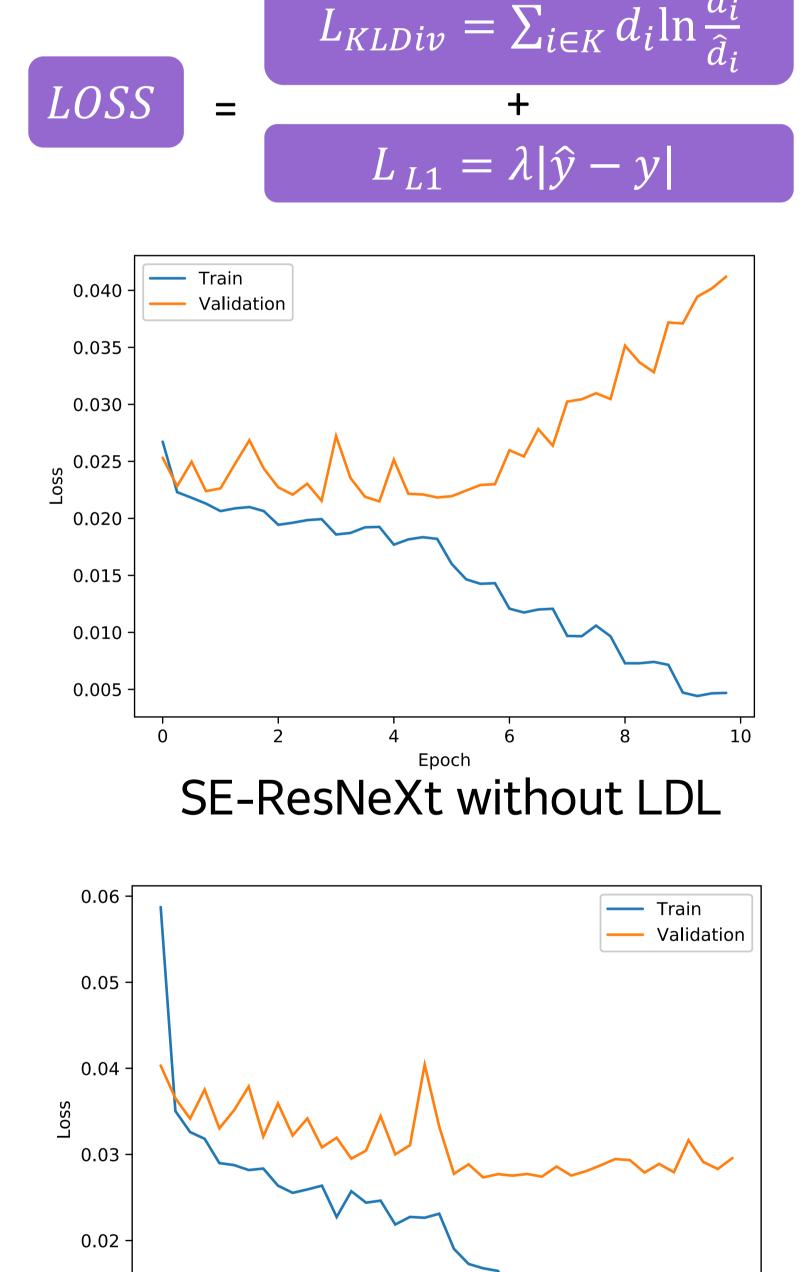


Label Distribution Learning (LDL) refers to

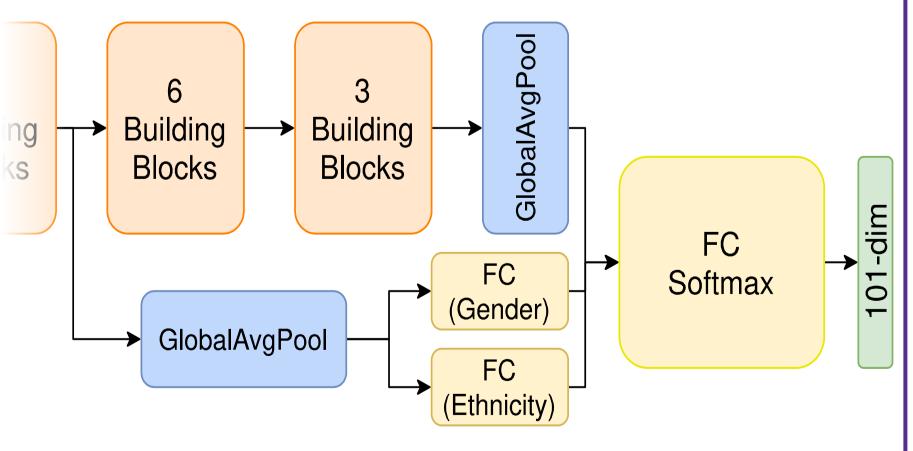
3. DLDL-v2

the method that using model to learn an age distribution from a facial image.

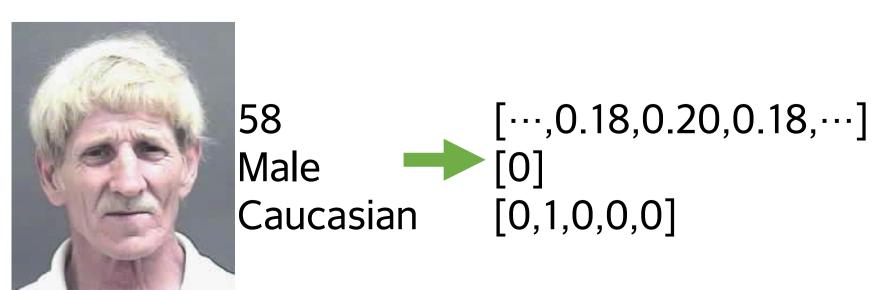
In this research, we follow the design of loss function in DLDL-v2, it consists two parts:



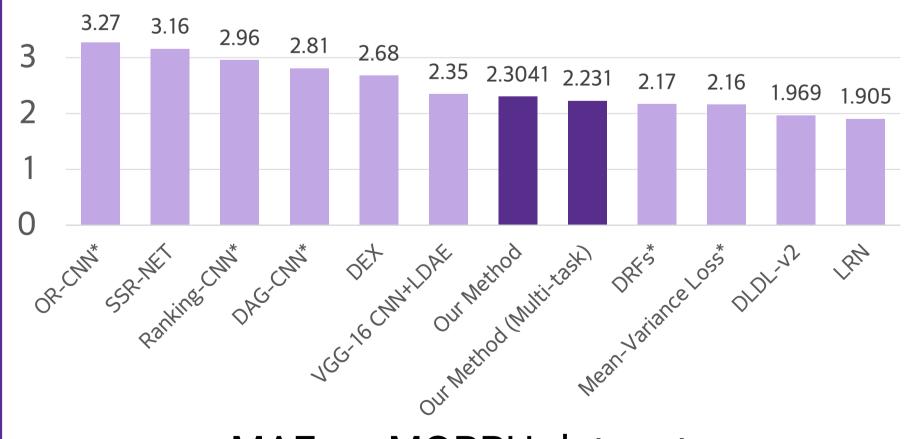
4. Multi-task Learning



The multi-task learning model add another two fully connected layer for gender and ethnicity predication to the original architecture. The results are fed into the final FC layer to produce the age estimation. This improves the accuracy of estimation on MORPH dataset.



5. Comparison



MAE on MORPH dataset (The lower the better) (*: The evaluation protocol is slightly different)



CA(3) on MegaAge-Asian dataset (The higher the better)

6. References

0.01

B.-B. Gao, H.-Y. Zhou, J. Wu, and X. Geng, "Age estimation using expectation of label distribution learning," in Proceedings of the 27th International Joint Conference on Artificial Intelligence (IJCAI 2018),

SE-ResNeXt with LDL

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- G. Antipov, M. Baccouche, S.-A. Berrani, and J.-L. Dugelay, "Effective training of convolutional neural networks for face-based gender and age prediction," *Pattern Recognition*, vol. 72, pp. 15 – 26, 2017. [Online].
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Please scan the QR-code for the paper and code

