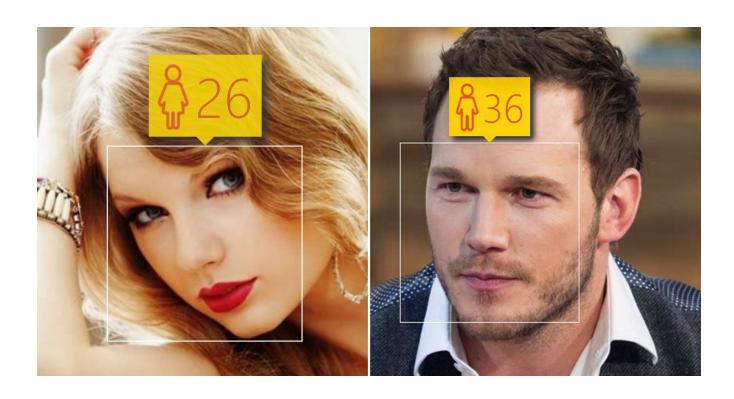
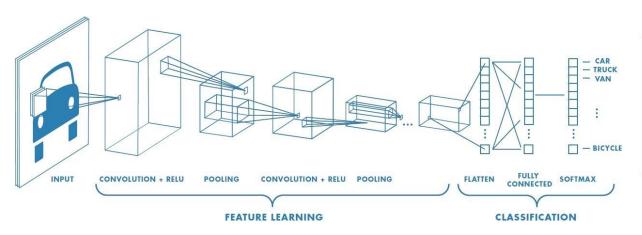
Building an Age Guesser using Transfer Learning

Arnav Garg, Shail Mirpuri, Darren Tsang

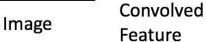
Task



Convolutional Neural Nets (CNNs)

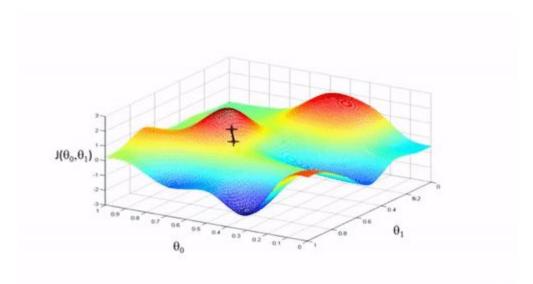


1,	1,0	1,	0	0
0,0	1,	1,0	1	0
0,1	0,×0	1,	1	1
0	0	1	1	0
0	1	1	0	0





Gradient Descent

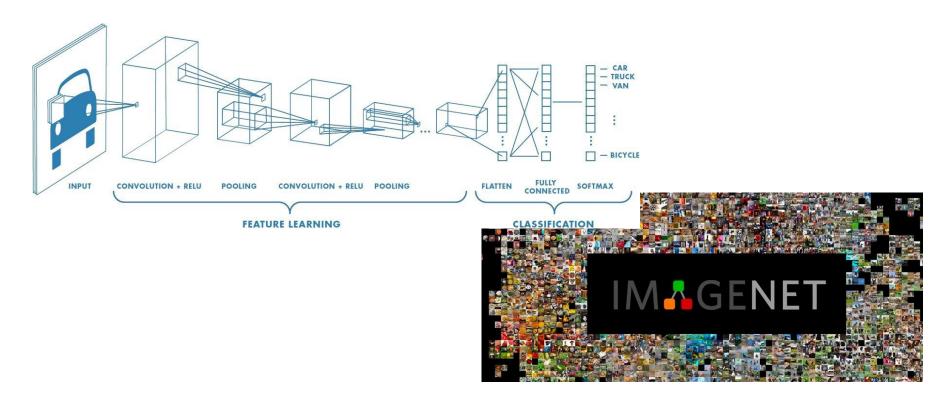


Repeat until convergence {

$$\theta_j \leftarrow \theta_j - \alpha \frac{\partial}{\partial \theta_j} J(\theta)$$

}

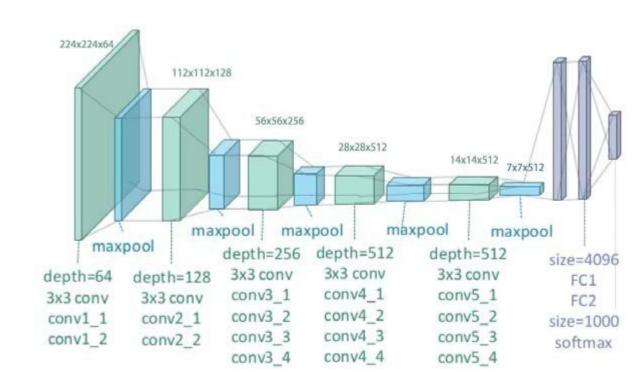
Transfer Learning



VGG-19

Consists of:

- 3 x 3 filters with a stride of 1 and zero-padding
- Max Pooling Layer with a stride of 2
- 19 layers



The IMDB-WIKI Dataset

- 500,000+ images of celebrities from IMDB and Wikipedia
 - Collected by researchers and publicly available online
- Discarded images where the age did not make sense
 - Negative ages
 - Ages over 100
- Created subsets from cleaned dataset for training, validation, and testing
 - Training used to train model
 - Validation used during training process to ensure overfitting does not happen
 - Testing used at the very end to evaluate how well our model does on new, unseen data
- https://data.vision.ee.ethz.ch/cvl/rrothe/imdb-wiki/

Hyperparameters

- Learning Rate: 0.01
- Batch Size: 10
- Trainable: Last Layer
- Optimizer: Adam
- Random Sample 50,000 images to train our dataset (about 10% of total dataset)

Good Predictions







Actual Age Predicted Age

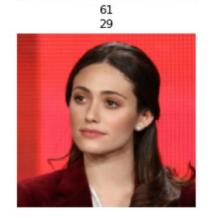


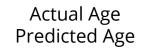




Bad Predictions

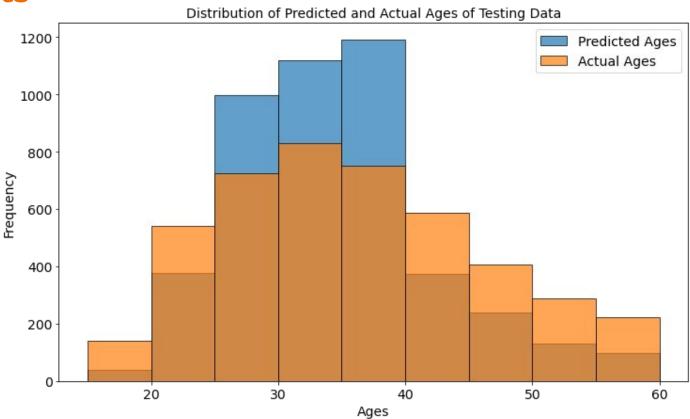




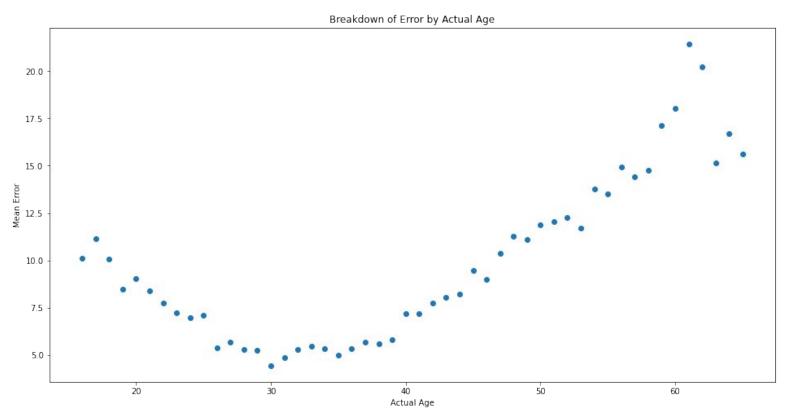


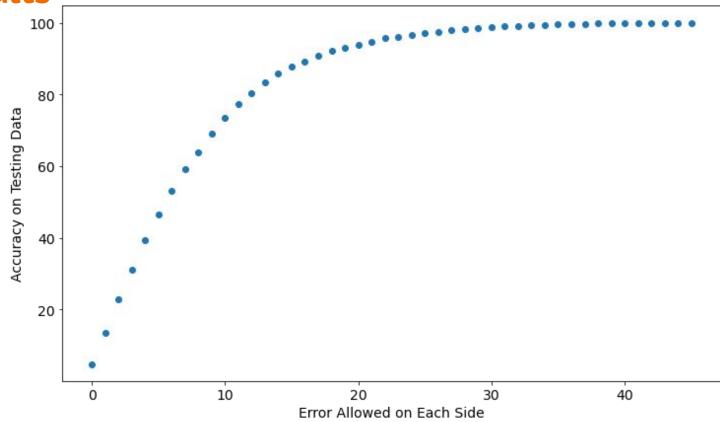






- **Mean** Absolute Error: **7.83** years
- Standard Deviation of Absolute Error: 6.84





Challenges

- Real age vs apparent age
- Massive dataset, low computing power
- Age skew in dataset



Future Work

- Make it work on uncropped images
- Multiple people in images
- Train for longer periods of time on full dataset
- Build user application to expand dataset
- Use loss function that better accounts for close guesses
- Look into newer, more advanced network architectures (ResNet, Inception)