# Face Keypoints Detection (W207 Final Project)

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#### Face Keypoints Detection: task and dataset

**Task**: Detect coordinates of 15 face keypoints on each image

**Evaluation criteria**:

RMSE = 
$$\sqrt{\frac{1}{n} \sum_{i=1}^{n} (y_i - \hat{y}_i)^2}$$
,

Train Images (2K)



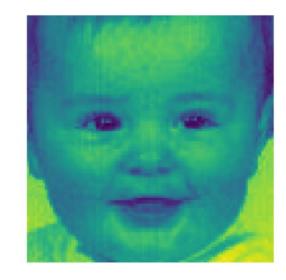
**Incomplete Train Images (5K)** 



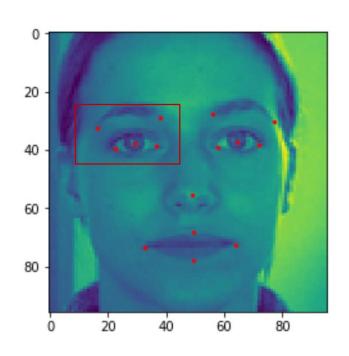
**Evaluation Images (5K)** 

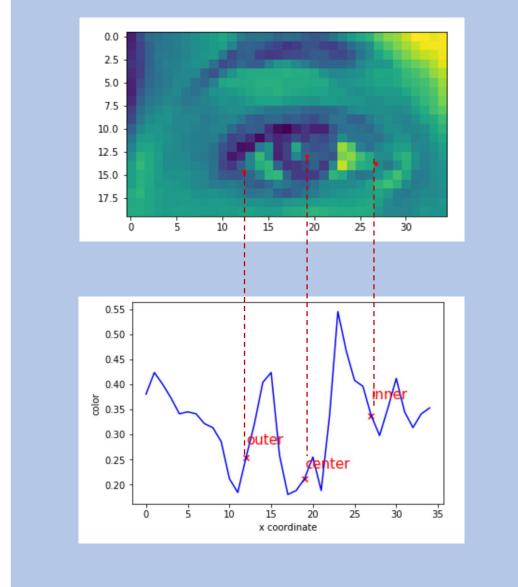


Test Images (1.7K)



#### Creating features for keypoint detection

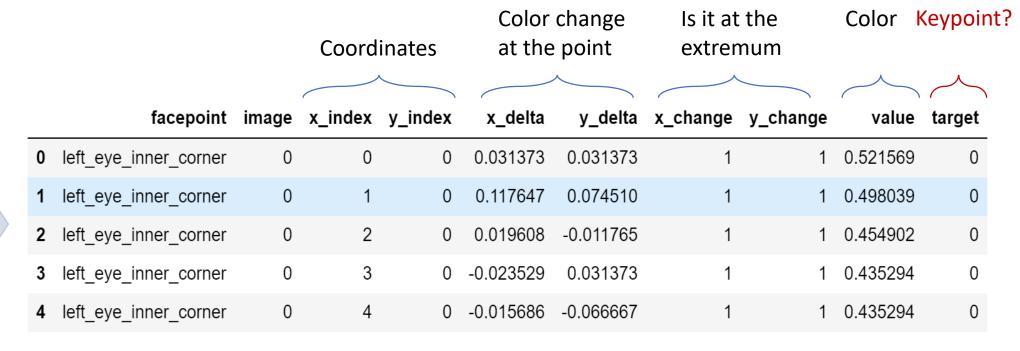




#### **Key features selected:**

- Coordinates of the key point
- Color of the point
- Change of the color compared to the neighboring keypoint
- Is it near the extremum where the color radically changes

### First try: Random Forest



Train dataset (9K 96X96 images)

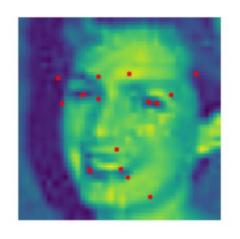
- 8 features + target
- 9K images\* 96\*96 pixels \*15 features = 15.7 M observations

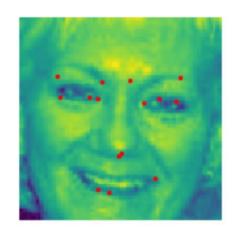
#### **Suggested Machine Learning approaches:**

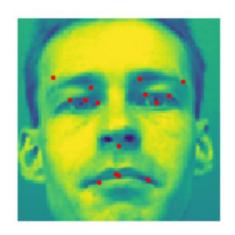
- Probability methods (Naïve Bayes, Bernoulli) to estimate the keypoint probability for each pixel
- Tree methods to find out feature combinations to detect the keypoints

### Random Forest method result: very approximate detection

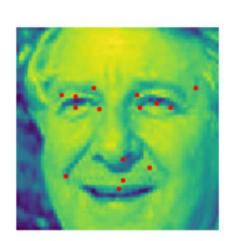






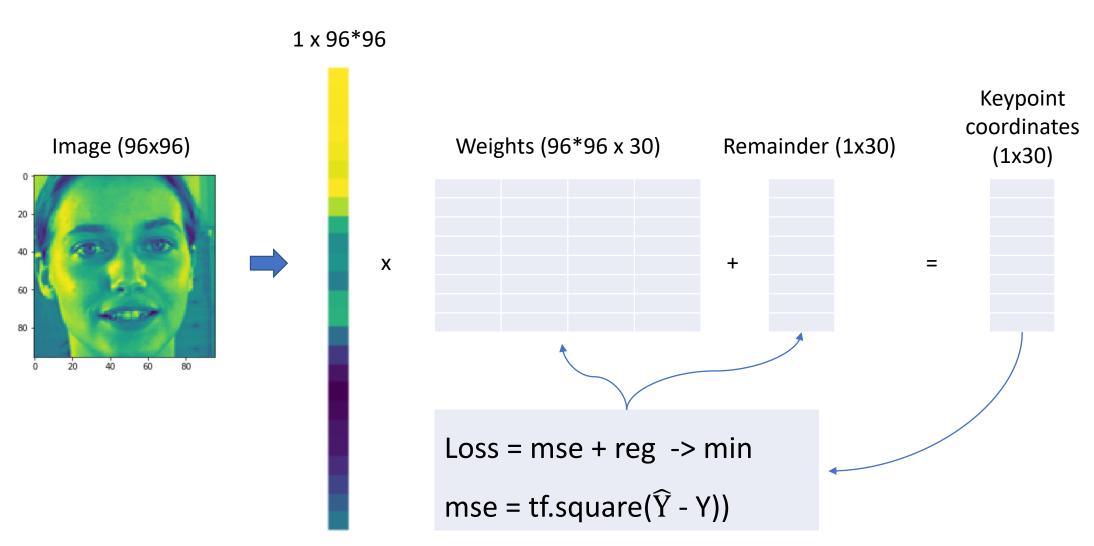






- RMSE = 3.57
- Kaggle public score = 4.98

### Second try: Neural net using Tensorflow

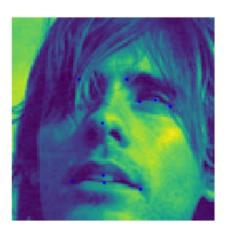


## Expanding the number of available observations by 23K images

• Flipping (+2k)



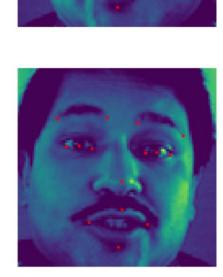
images



Complete incomplete

observations (+5k)

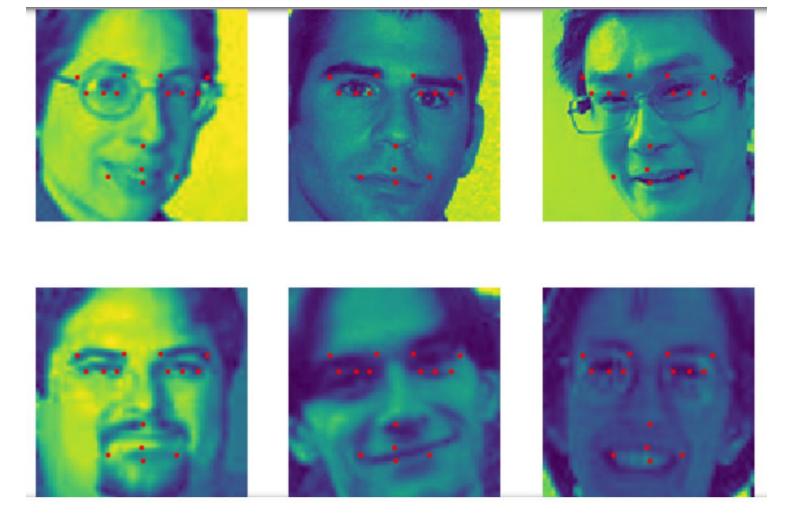
 Shift images in 4 directions (+16K)





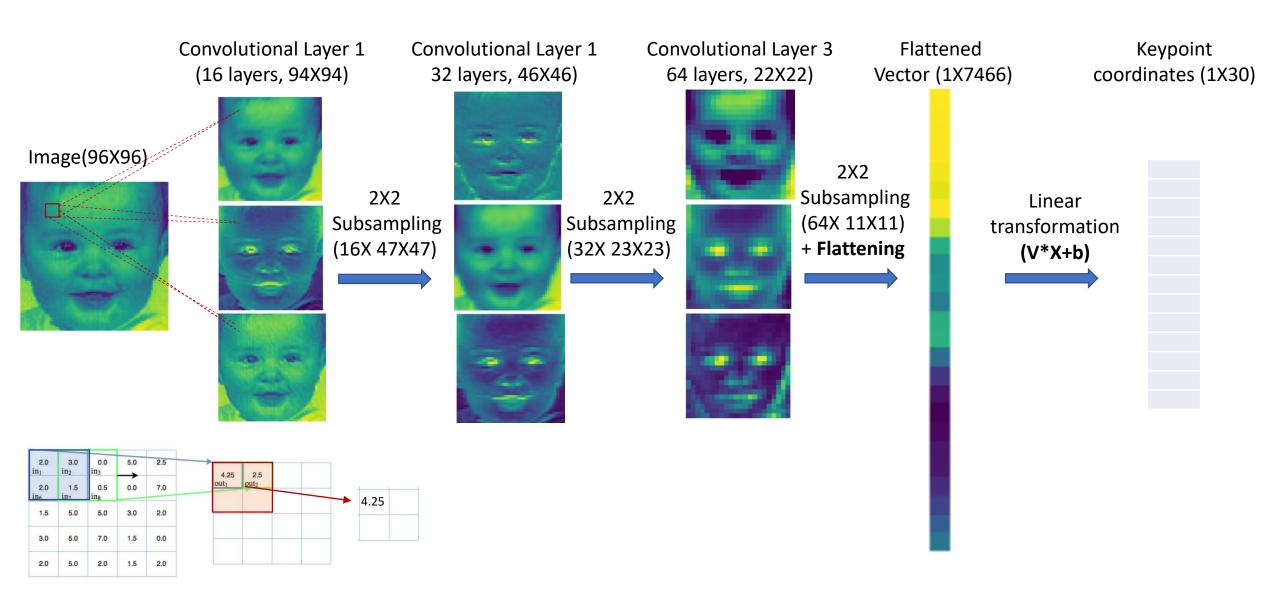


## 1-layer neural net result: better result. Poorer performance on inclined faces



- RMSE = 2.91
- Kaggle public score = 3.98

## Third try: deep learning through convolutional neural networks (CNN)

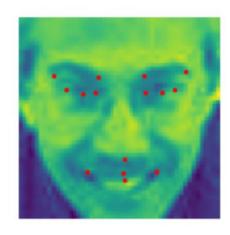


### Convolutional neural net result. Works with inclined faces











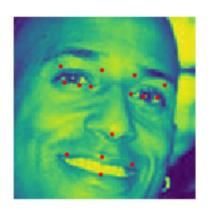


- RMSE = 2.28
- Kaggle public score = 2.45

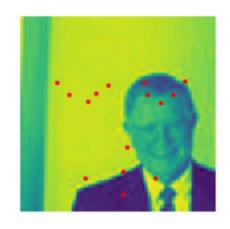
#### How I would continue to improve the results

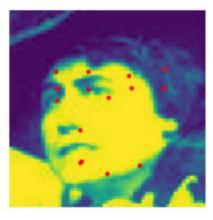






Expand and shift images
(to train the model on
 "non-full" and non centered faces)







 Squeeze and shift by diagonal (to train the model on "small" and non-centered faces)