HEAD POSE DETECTION

CS584 - MACHINE LEARNING

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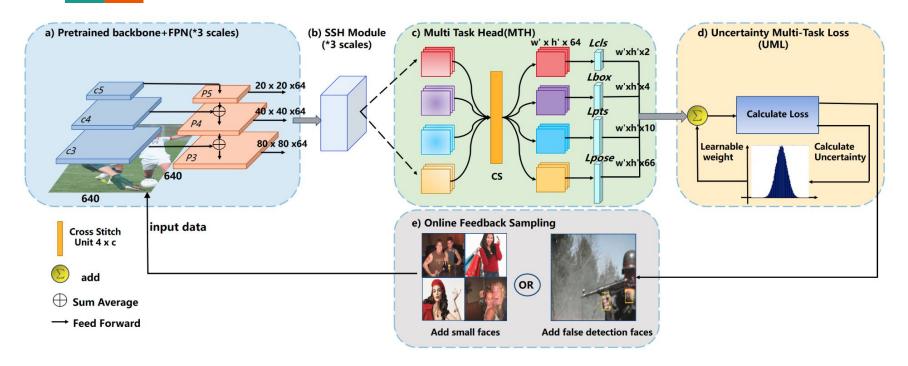
Introduction

- Motivation
- Use Cases HCI, Safety Standards etc
- Current State Head Pose Detection

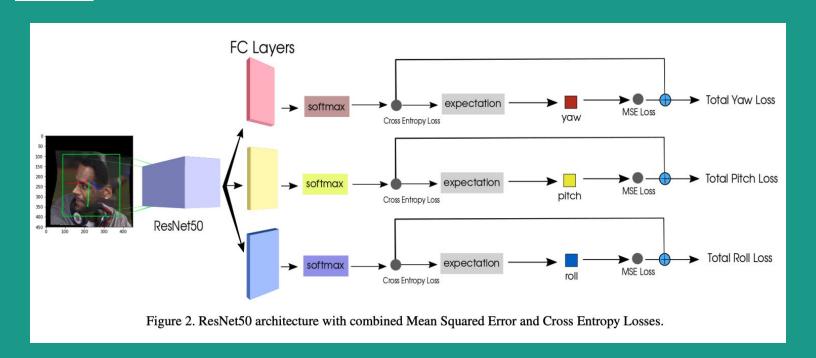
PAPERS

- MOS: A Low Latency and Lightweight Framework for Face Detection, Landmark Localization, and Head Pose Estimation
- Fine-Grained Head Pose Estimation Without Keypoints

MOS



HOPENET



Challenges with Head pose detection

- Amount of training data required
- Number of epochs
- Availability of computational resources

Proposed Solution

- Simple CNN for pose estimation.
- Use Pre-trained model like ResNet to perform Head pose detection
- Use similar network/Loss calculation as HOPENET[2] along with dlib to check the performance.

Dataset - 300W_LP (AFW)

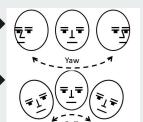
- Training 4434
- Validation 554
- **Testing** 555



Labels:

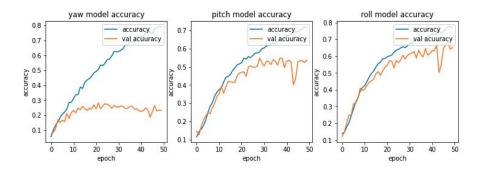


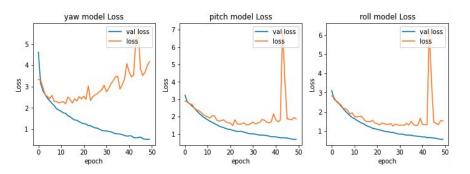
- Face x,y coordinates
- yaw left, right ⇒
- ←• pitch up, down
- roll rotate



Start

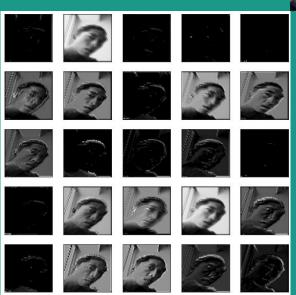
- Basic CNN layers
- Trained with direct images (without cropping)
- Classification
- Loss function categorical_crossentropy
- Observation Accuracy of *yaw* is under 0.3.





CNN Model





Layer (type)	Output Shape	Param #	Connected to
input_1 (InputLayer)	[(None, 240, 240, 3		[]
conv2d (Conv2D)	(None, 60, 60, 64)	23296	['input_1[0][0]']
max_pooling2d (MaxPooling2D)	(None, 29, 29, 64)	0	['conv2d[0][0]']
conv2d_1 (Conv2D)	(None, 29, 29, 192)	307392	['max_pooling2d[0][0]']
max_pooling2d_1 (MaxPooling2D)	(None, 14, 14, 192)	0	['conv2d_1[0][0]']
batch_normalization (BatchNormalization)	(None, 14, 14, 192)	768	['max_pooling2d_1[0][0]']
conv2d_2 (Conv2D)	(None, 14, 14, 384)	663936	['batch_normalization[0][0]']
conv2d_3 (Conv2D)	(None, 14, 14, 256)	884992	['conv2d_2[0][0]']
<pre>batch_normalization_1 (BatchNo rmalization)</pre>	(None, 14, 14, 256)	1024	['conv2d_3[0][0]']
conv2d_4 (Conv2D)	(None, 14, 14, 256)	590080	['batch_normalization_1[0][0]']
conv2d_5 (Conv2D)	(None, 14, 14, 512)	1180160	['conv2d_4[0][0]']
max_pooling2d_2 (MaxPooling2D)	(None, 6, 6, 512)	0	['conv2d_5[0][0]']
batch_normalization_2 (BatchNormalization)	(None, 6, 6, 512)	2048	['max_pooling2d_2[0][0]']
flatten (Flatten)	(None, 18432)	0	['batch_normalization_2[0][0]']
dropout (Dropout)	(None, 18432)	0	['flatten[0][0]']
dense (Dense)	(None, 4096)	75501568	['dropout[0][0]']
yaw (Dense)	(None, 66)	270402	['dense[0][0]']
pitch (Dense)	(None, 66)	270402	['dense[0][0]']
roll (Dense)	(None, 66)	270402	['dense[0][0]']

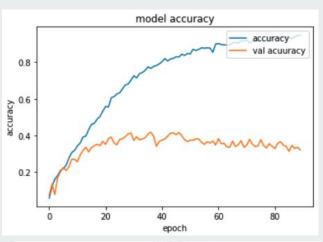
Total params: 79,966,470
Trainable params: 79,964,550
Non-trainable params: 1,920

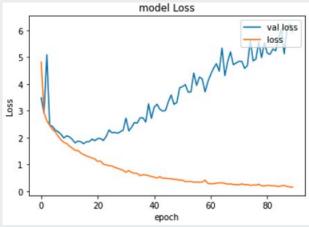
Model considering yaw only

- Basic CNN layers
- Trained with direct images (without cropping)
- Classification
- Loss function -

categorical_crossentropy

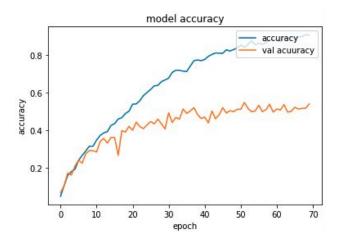
• Observation - Accuracy is sub 0.4.

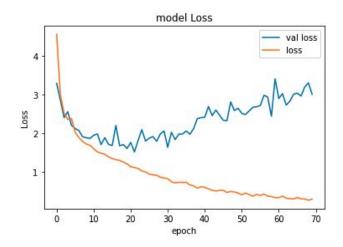




Model with cropped images

- Basic CNN layers
- Trained with cropped images and previous model (yaw only)
- Observation Accuracy increased to 0.54



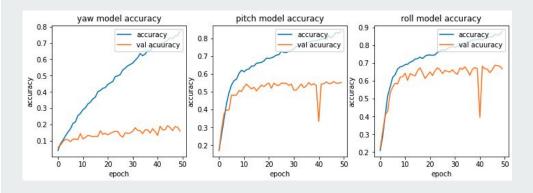


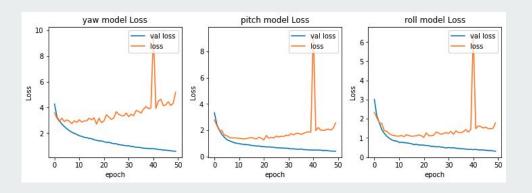
Combining y, p, r

- Basic CNN layers
- Trained with cropped images
- Loss function -

categorical_crossentropy

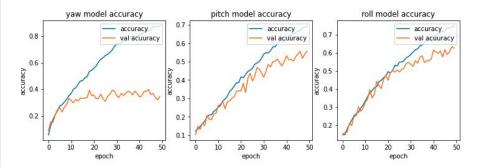
 Observation - no improvement in yaw accuracy

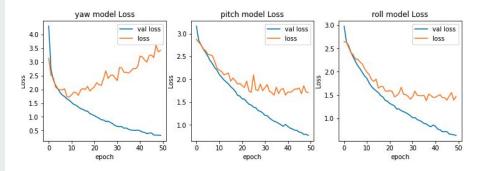




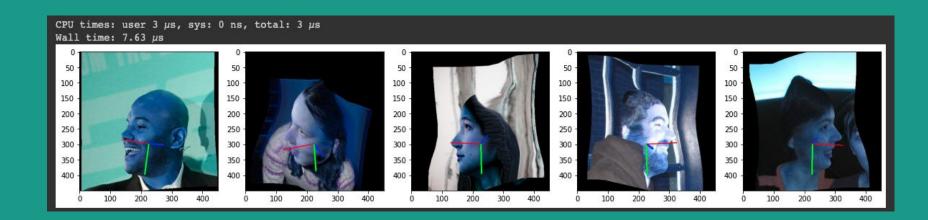
Final Basic Model

- Basic CNN layers
- Trained with cropped images
- For the previous model we updated loss_weights to (1, 0.3, 0.3)
- Observations: Comparatively the model performed better with yaw accuracy around 0.4





Final result of CNN

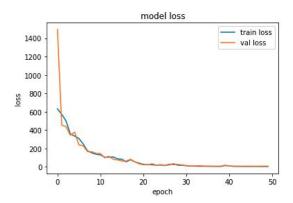


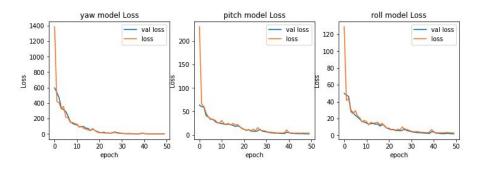
Updated Model

- Basic CNN layers
- Trained with cropped images
- Loss Function -

$$\mathcal{L} = H(y, \hat{y}) + \alpha \cdot MSE(y, \hat{y})$$

 Observations: Nothing much changed (comparing MAE)



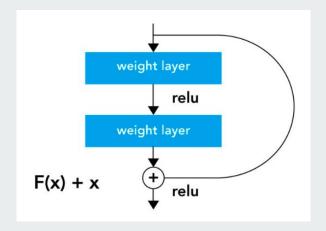


Added Residual net

Loss Function -

$$\mathcal{L} = H(y, \hat{y}) + \alpha \cdot MSE(y, \hat{y})$$

 Observation - No improvement in MAE



	3 Resnet blocks	Resnet-50
Yaw	49.08	48.71
Pitch	10.82	8.79
Roll	8.03	4.73
Average	22.648	20.74

Demo



THANK YOU