

Efficient facial detection on embedded system with CNN for surveillance tasks

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BACKGROUND

Goal:

—> Integrate facial detection model for analysis of surveillance footage in real-time in an embedded environment

Motivation:

—> Traditional surveillance systems do not offer “smart” analysis of footage in real time.
—> Huge quantities of data are stored on disks far away despite not providing much use

Approach:

—> Given a video feed from a surveillance camera, a facial detection model can be run on the frames to determine if there are humans in the frames whom could be hailed as intruders

Advantage:

—> System is able to distinguish the interesting from non-interesting frames to minimize storage needs

PROBLEM STATEMENT

Design a facial detection model that can be an integrated with a real-time surveillance system that is running on a Raspberry Pi 3.

Input: Sequence of frames derived from splitting up a real-time camera feed where set of frames (S) is assumed to be in format $S = \{f_1, f_2, f_3 \dots f_n-1, f_n\}$ as time progresses forward

Output: Frames returned that have are found to have faces in them. Frames can be either stored or sent via email.

Performance goals:

FPS ≥ 2

Accuracy goals:

> 70%

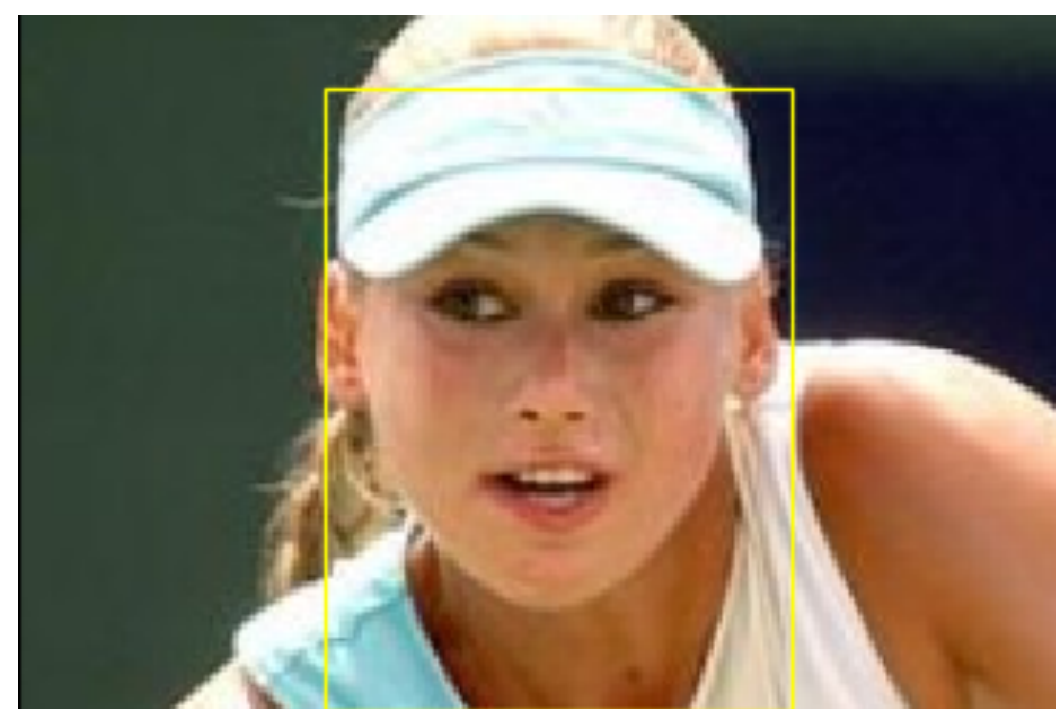


Figure 1; Facial detection model on an image from LFW dataset

DATASET

Dataset used : A Benchmark for Face Detection in Unconstrained Settings [FDDB]

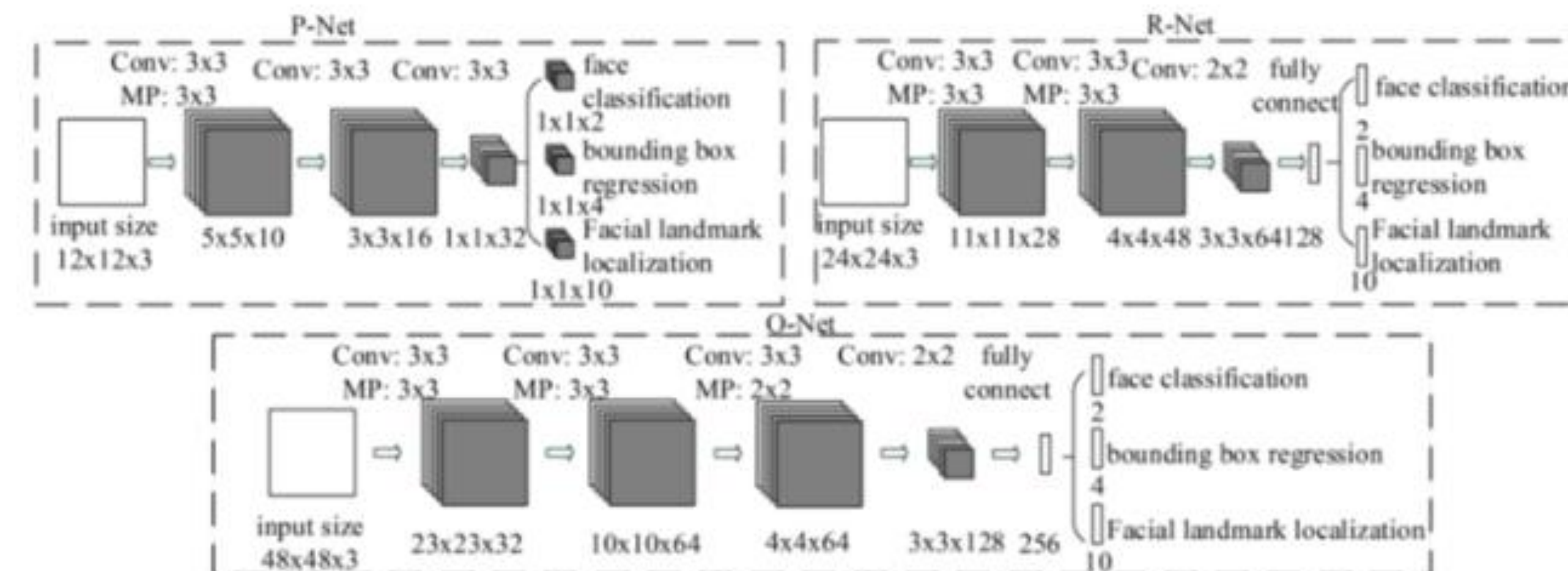
Breakdown :

- > 12,000 Training Images
- > 1,000 Validation Images
- > 1,000 Testing images



Figure 2; Images were fed to network in form of an image pyramid

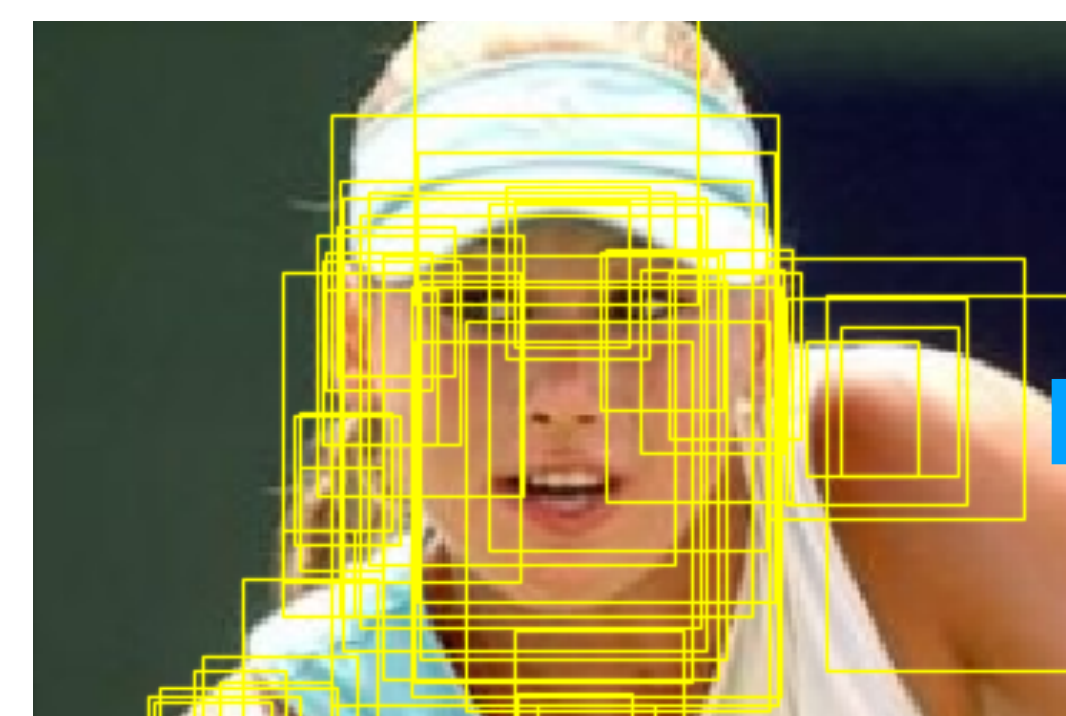
ARCHITECTURE



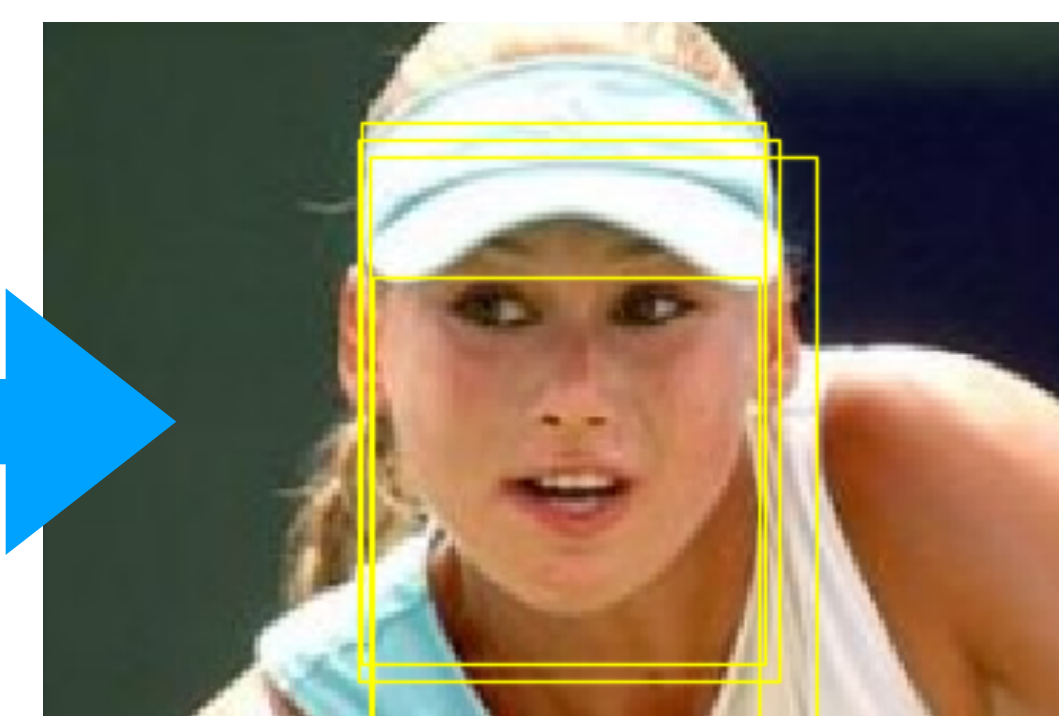
—> Proposal Network (P-NET) : Large number of bounding box proposals to the refinement network

—> Refinement Network (R-NET) : Refine the output from P-NET, drastically reduce number of bounding boxes

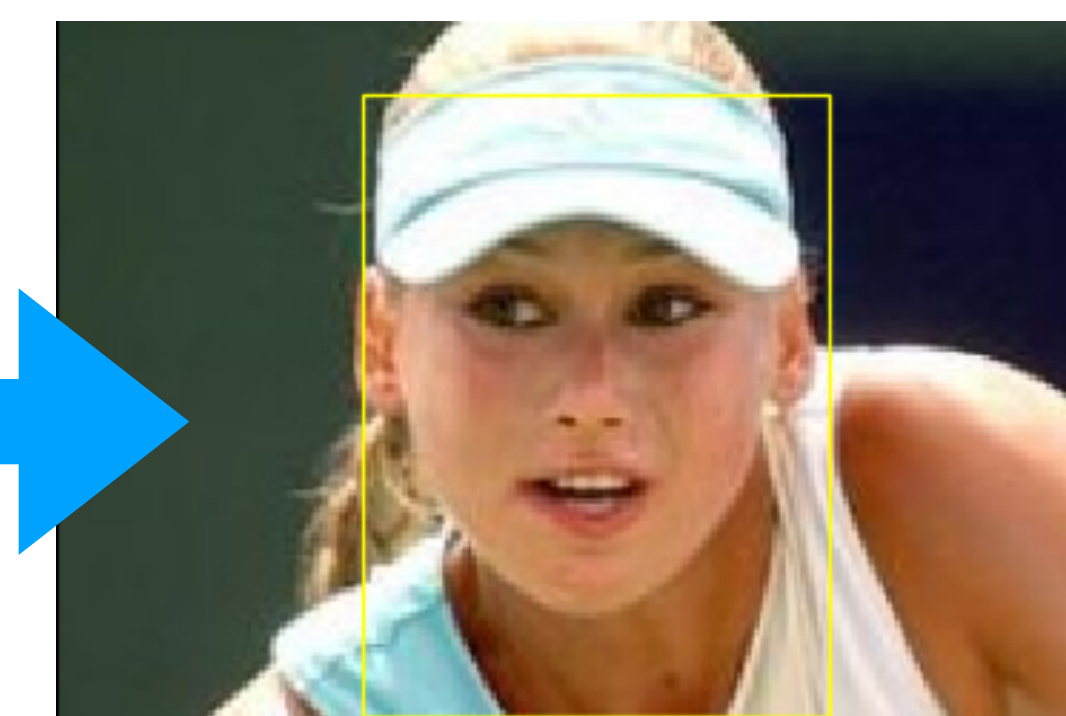
—> Output Network (O-NET) : Further refine the output from previous stage and return one bounding box per person with the highest confidence



P - NET RESULT ON AN IMAGE FROM LFW DATASET



R - NET RESULT ON AN IMAGE FROM LFW DATASET



O - NET RESULT ON AN IMAGE FROM LFW DATASET

EVALUATION

Statistics

Accuracy on finding correct number of faces = 84.7%

Accuracy IOU= 59.4%

FPS on laptop video feed: ~ 15 frames

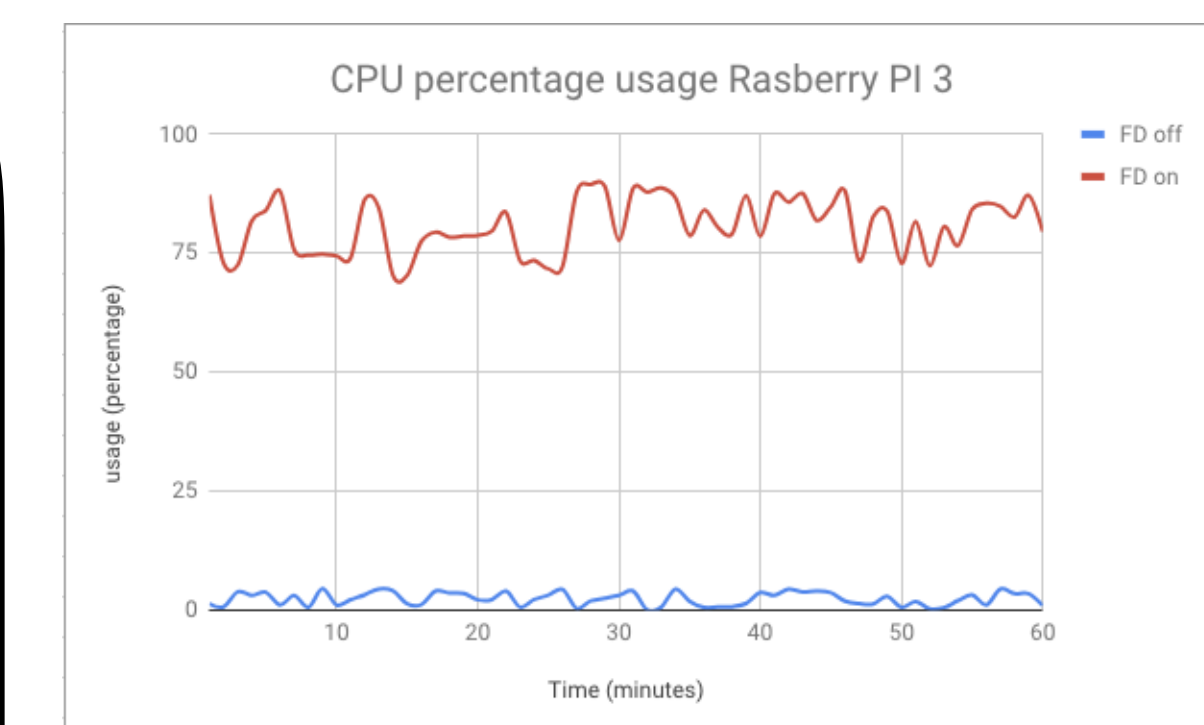
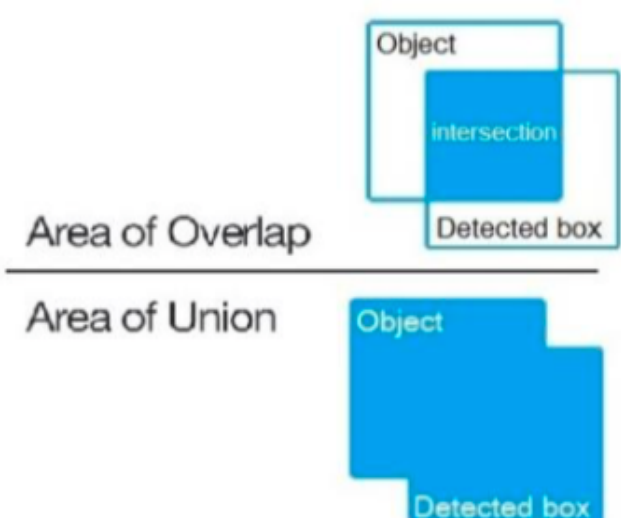
FPS on Rpi video feed: ~ 2 frames

Average Proposal Network cost = .157 seconds

Average Refinement Network cost = .096 seconds

Average Output Network cost = 0.017 seconds

** FPS and Performance statistics were taken on small images



INTG. W/ SURVEILLANCE AND FUTURE WORK

—> Surveillance system is integrated with face detection (Shown on the left)

Improvements:

—> Retrain the model and/or model changes with additions to perform facial recognition

—> Object tracking between frames instead of recalculating the model between each frame

—> Better visualization

—> Better performance by simplifying/optimizing the model