

Determination of vital parameters using photoplethysmography

6th semester summer training project



Objectives of the Training

- To create backend models for integration into their MyFitPrint Application.
- To create a contact plethysmograph model for determination of heart rate.
- To create a contact and contactless model for determination of SpO2.
- To create a deep learning based model for determination of blood pressure without ECG dependency.

The background features several decorative elements in shades of blue. On the left, there is a large, vertical, textured brushstroke that tapers towards the bottom, with numerous small, dark blue splatters scattered around it. At the top right, there are several parallel, slightly curved lines that resemble a fine brushstroke or a series of thin, overlapping strokes. In the bottom right corner, there are two overlapping, soft-edged, light blue shapes that look like watercolor washes or ink blots.

Techniques and Technology used

The Technique

Plethysmography

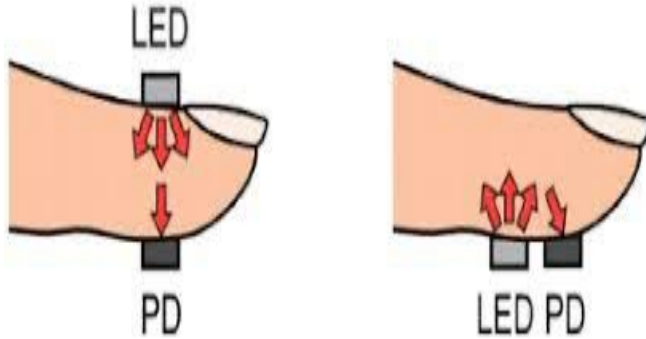
Plethysmography is the volumetric measurement of an organ, resulting from fluctuations in the amount of blood or air it contains. The change in blood volume is synchronous to the heart beat, so it can be used to detect heart rate.

Photoplethysmography

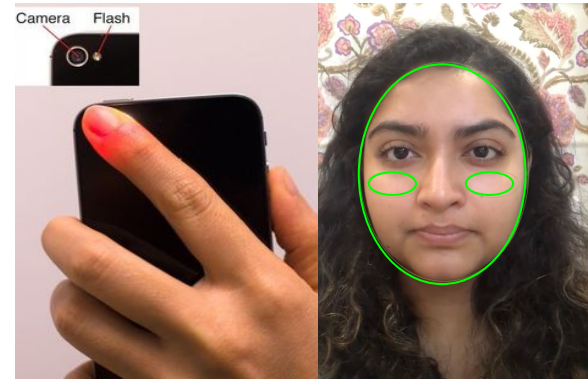
Photoplethysmography is just a means of plethysmography that uses optical techniques. There are two basic types of photoplethysmography: transmittance and reflectance. Another classification may be contact and contactless photoplethysmography.

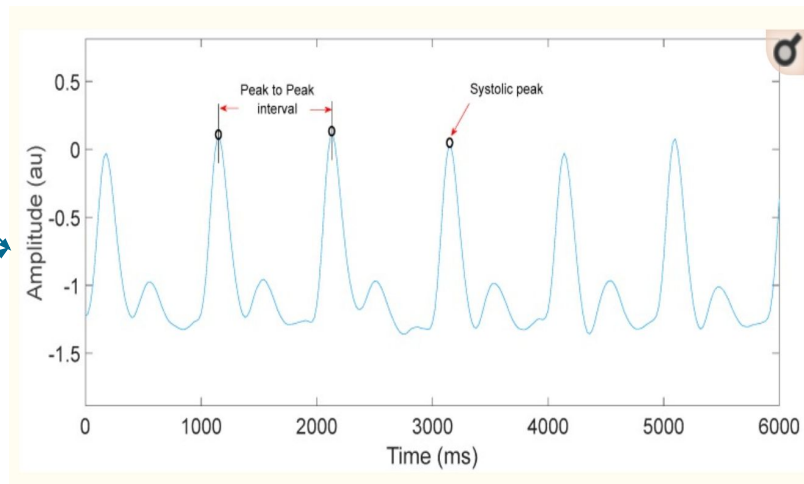
The Photoplethysmography Distinctions

Transmittance (left) and
Reflectance (right)



Contact (left) and
Contactless (right)





Heart Rate

SpO2

Blood Pressure

Technology Stack



Python

NumPy, SciPy, Dlib,
OpenCV, Matplotlib,
etc.



MATLAB

For preliminary
signal processing



Deep Learning

U-Nets and Residual
Neural Networks for
BP.



The Project

Three vital parameters: Heart Rate, SpO2, Blood Pressure

Model 1: Heart Rate Measurement

Obtaining the PPG

The photoplethysmograph is taken from a video of the fingertip between length 10-30 seconds

Filtering the signal



Band pass filtering + squaring + determination of areas of interest

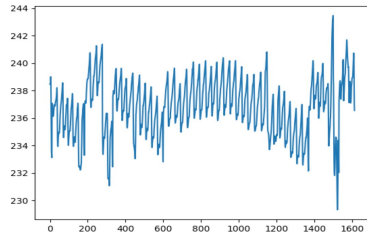
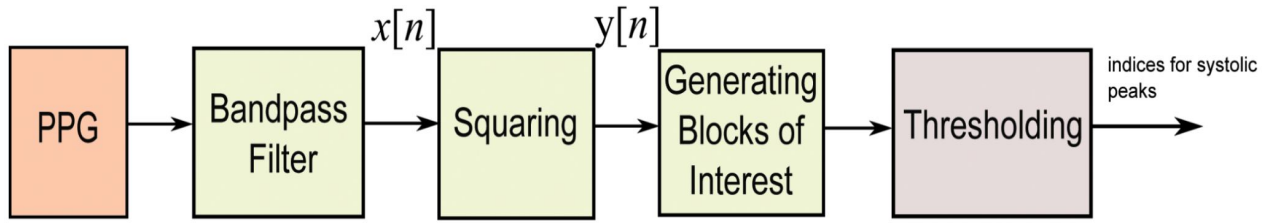
Detection of Parameter

Peaks are detected in the area of interest and the overall estimation of the heart rate is made

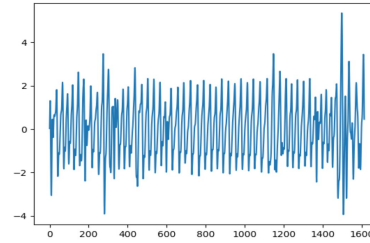


Salient Features & Advantages

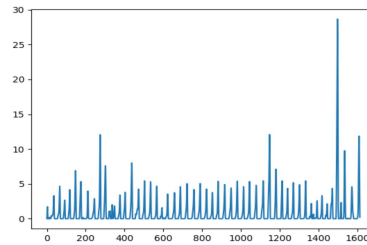
- Origin Independent: The video can be from the face or the fingertip, the computation methodology from the second stage onwards, i.e, bandpass filtering remains the same.
 - Event-related moving averages with dynamic threshold.
 - Signal adaptive and noise robust.
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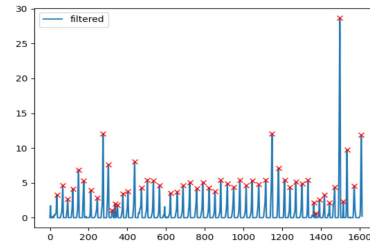
Raw PPG signal



Band pass filtering



Squaring



Thresholding and detection

| Video Name | Actual BPM | Predicted BPM | Error | Error Percentage | Mean Error |
|-------------------------|------------|---------------|-------|------------------|------------|
| 102ajay | 102 | 97.88 | 4.12 | 4.04 | 6.80 |
| 104ajay | 104 | 91.44 | 12.56 | 12.07 | |
| 106ajay | 106 | 92.12 | 13.88 | 13.09 | |
| 20210616finger_80bpm | 80 | 74.71 | 5.29 | 6.61 | |
| 20210616finger_85bpm(1) | 85 | 75.18 | 9.82 | 11.55 | |
| 20210616finger_85bpm | 85 | 78.96 | 6.04 | 7.11 | |
| 64hr_12rr_98spo2_ft | 64 | 58.78 | 5.22 | 8.16 | |
| 65karan | 65 | 67.23 | 2.23 | 3.43 | |
| 66karan | 66 | 63.28 | 2.72 | 4.12 | |
| 70karan | 70 | 66.30 | 3.70 | 5.28 | |
| 71karan | 71 | 66.58 | 4.42 | 6.23 | |
| 93ajay | 93 | 87.77 | 5.23 | 5.62 | |
| 98spo2_70bpm | 70 | 63.87 | 6.13 | 8.76 | |
| HC_10 | 85 | 76.30 | 8.70 | 10.24 | |
| HC_3 | 83 | 78.68 | 4.32 | 5.21 | |
| HC_4 | 80 | 72.71 | 7.29 | 9.12 | |
| HC_5 | 75 | 74.84 | 0.16 | 0.22 | |
| HC_6 | 80 | 77.17 | 2.83 | 3.54 | |
| HC_7 | 70 | 63.87 | 6.13 | 8.76 | |
| HC_Dro_2 | 96 | 86.54 | 9.46 | 9.86 | |
| HC_nit_1 | 85 | 81.97 | 3.03 | 3.57 | |
| HC_shr_1 | 107 | 56.19 | 50.81 | 47.49 | |
| HC_Son_2 | 68 | 60.48 | 7.52 | 11.06 | |
| hr100 | 100 | 87.77 | 12.23 | 12.23 | |
| hr65+ | 65 | 66.58 | 1.58 | 2.42 | |
| hr65_2 | 65 | 67.23 | 2.23 | 3.43 | |
| hr65_3 | 65 | 63.28 | 1.72 | 2.65 | |
| hr70 | 70 | 66.30 | 3.70 | 5.28 | |
| KL_rj_1 | 54 | 60.52 | 6.52 | 12.08 | |
| KL_sha_1 | 64 | 63.51 | 0.49 | 0.77 | |
| video2 | 72 | 75.73 | 3.73 | 5.18 | |
| video3 | 80 | 71.77 | 8.23 | 10.29 | |
| video4 | 77 | 73.82 | 3.18 | 4.13 | |
| video5 | 80 | 77.07 | 2.93 | 3.66 | |
| video6 | 85 | 75.18 | 9.82 | 11.55 | |

Model 2: SpO2 Measurement

Obtaining the PPG

The photoplethysmograph is taken from a video of the fingertip or face greater than 15 seconds.

Ac and Dc Component extraction

Done by taking up mean and standard deviation values of the red and blue components per frame.

Detection of Parameter

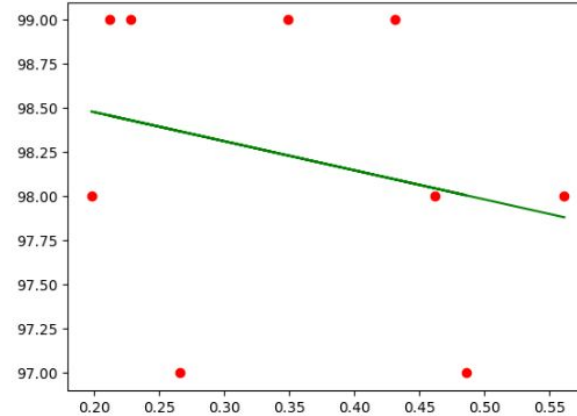
$$SpO_2 = A - B \frac{AC_{RED}/DC_{RED}}{AC_{BLUE}/DC_{BLUE}}$$

Equations Used

$$RR = \frac{AC_{\lambda 1} / DC_{\lambda 1}}{AC_{\lambda 2} / DC_{\lambda 2}}$$



$$SpO_2 = \alpha \cdot RR + \beta$$

Getting α and β





Salient Features & Advantages

- Approach works for both face and fingertip SpO₂ detection.
 - The noise reduction and filtering stages can be eliminated with no comparable reductions in result quality reducing the algorithm complexity.
 - The videos have to be at least 15 seconds long for suitable detection.
 - The α and β parameters are calibration constants that can be obtained using ground truth values for available training data set. This can be done by fitting a simple linear regression model.
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Test Results

Fingertip Videos

Face Videos

| NAME | Actual SpO2 | Predicted SpO2 | Error Percentage | Mean Error |
|-----------------------------|-------------|----------------|------------------|------------|
| HC_4.mp4 | | 97 | | 1.0125 |
| 104ajay.mp4 | | 98 | | |
| HC_5.mp4 | | 97 | | |
| HC_6.mp4 | | 97 | | |
| 20210616finger_80bpm.mp4 | | 97 | | |
| HC_7.mp4 | | 97 | | |
| 20210616finger_85bpm(1).mp4 | | 97 | | |
| HC_Dro_1.mp4 | 99 | 97 | 2.02 | |
| 20210616finger_85bpm.mp4 | | 97 | | |
| HC_Dro_2.mp4 | 98 | 98 | 0 | |
| 64hr_12rr_98spo2_ft.mp4 | | 98 | | |
| HC_Son_1.mp4 | 99 | 97 | 2.02 | |
| 65karan.mp4 | | 97 | | |
| HC_Son_2.mp4 | 99 | 98 | 1.02 | |
| 66karan.mp4 | | 98 | | |
| HC_Sum_1.mp4 | | 97 | | |
| 70karan.mp4 | | 98 | | |
| HC_nit_1.mp4 | | 98 | | |
| 71karan.mp4 | | 97 | | |
| HC_shr_1.mp4 | | 97 | | |
| KL_rj_1.mp4 | 98 | 98 | 0 | |
| 98spo2_70bpm.mp4 | | 97 | | |
| KL_sha_1.mp4 | 97 | 97 | 0 | |
| HC_1.mp4 | | 97 | | |
| VID_20210615_180226264.mp4 | | 98 | | |
| HC_10.mp4 | 98 | 97 | 1.02 | |
| HC_13.mp4 | | 98 | | |
| hr65+.mp4 | | 97 | | |
| hr65_2.mp4 | | 97 | | |
| HC_2.mp4 | | 97 | | |
| hr65_3.mp4 | | 98 | | |
| HC_3.mp4 | 99 | 97 | 2.02 | |
| hr70.mp4 | | 98 | | |

| Video Name | Actual SpO2 | RR value for signal | Predicted SpO2 | Error Percentage | Mean Error |
|------------|-------------|---------------------|----------------|------------------|------------|
| video1.mp4 | 98 | 0.05712976607 | 100 | 2.04 | 1.27 |
| video2.mp4 | 99 | 0.05453523164 | 100 | 1.01 | |
| video3.mp4 | 98 | 0.0915427457 | 99 | 1.02 | |
| video4.mp4 | 99 | 0.03758427041 | 100 | 1.01 | |

Model 3: Blood Pressure Measurement

Obtaining the PPG

Subsection of Physionet's MIMIC II dataset is used as photoplethysmograph sample

Approximation Network

One-dimensional deep supervised U-Net model

Refinement Network

One-dimensional MultiResUNet model and it outputs the desired ABP waveform

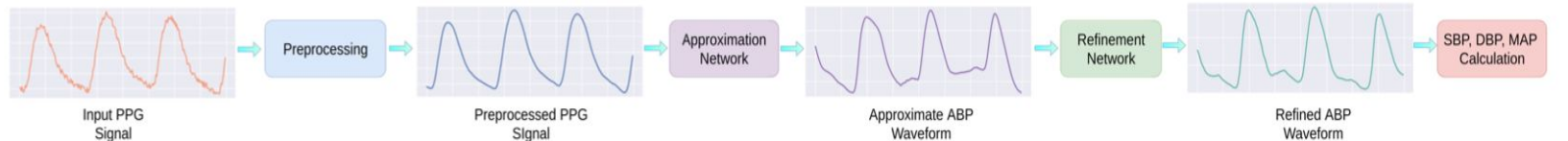
Salient Features & Advantages

- Instead of using mathematical approach like in case of previously discussed parameters, here a deep learning based approach is used.
- The ABP (Ambulatory Blood Pressure) waveform is predicted using multi layer U-Net model. A combination of two a refinement and an approximation network is used for optimal results.
- From the predicted ABP waveform, the BP parameters can be calculated using

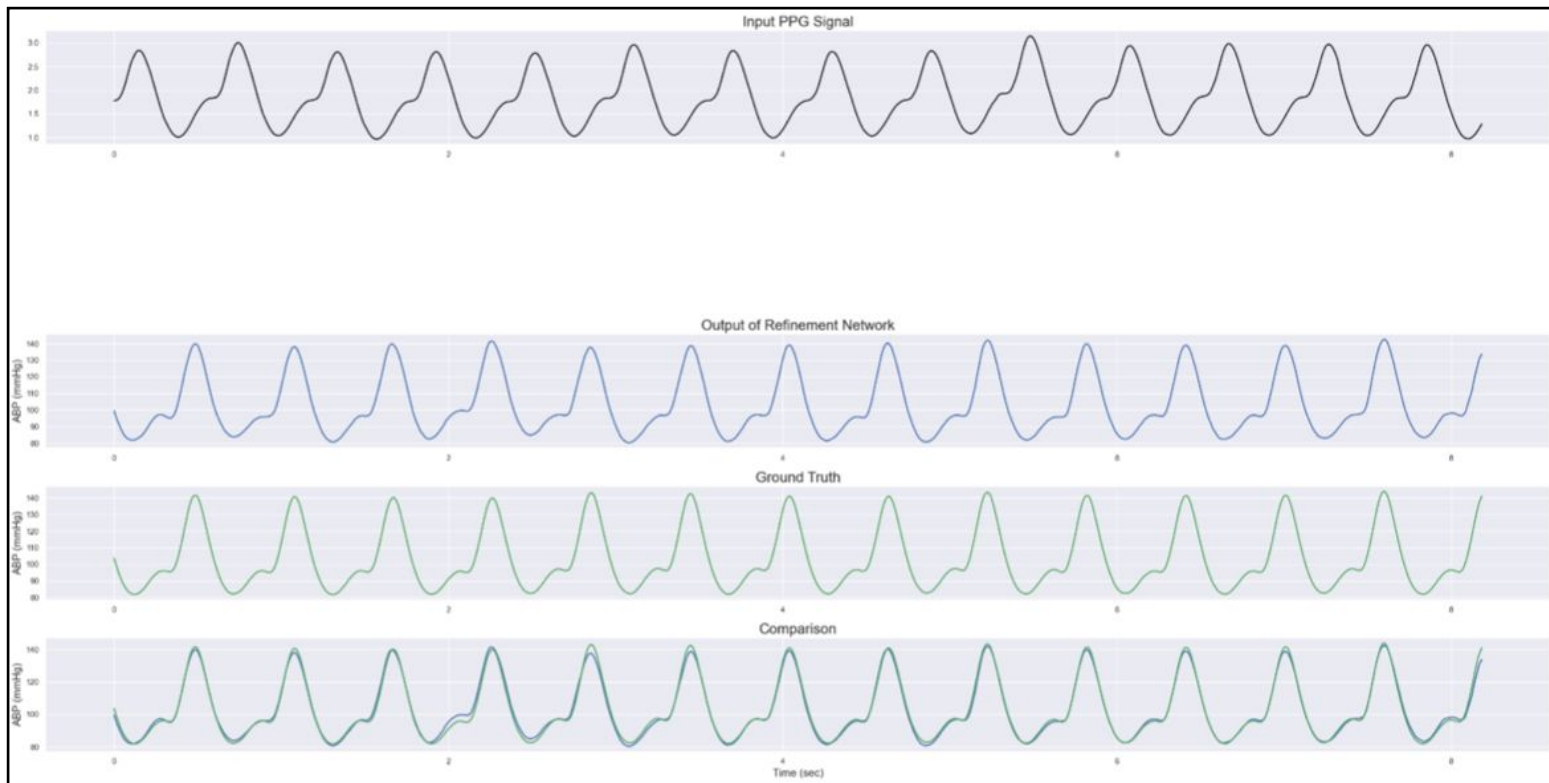
$$SBP = \max(ABP)$$

$$DBP = \min(ABP)$$

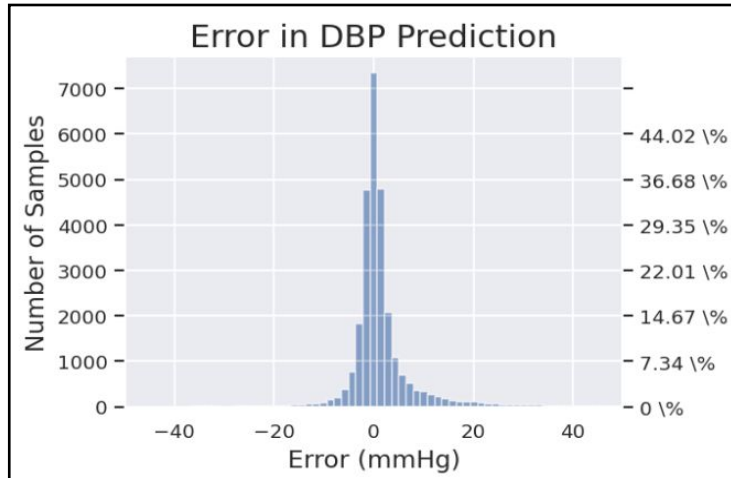
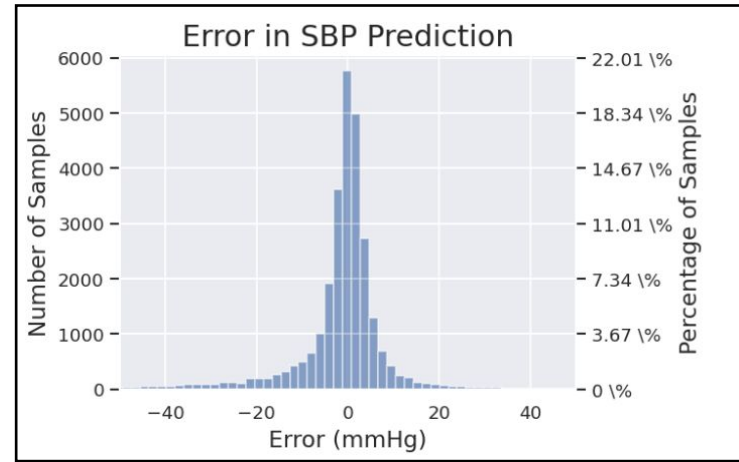
$$MAP = \text{mean}(ABP)$$



Waveform Plot



| AAMI Standard | | |
|---------------|--------|--------|
| | ME | STD |
| DBP | 1.619 | 6.859 |
| SBP | -1.582 | 10.688 |



Here, ME stands for the mean error for the test data and STD is the standard deviation for the computed errors.

Conclusions

Successful detection

All three parameters were suitably detected.

Optimal error rates

The error rates were all below the 10% mark.

Flexible signal duration

Models analysed signals for a variety of time frames without major error induction.



Thanks!

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