# PCA for FetalECG solution

December 20, 2022

## 1 PCA for Source Separation of Abdominal ECG Signals

#### 1.1 Introduction

In this exercise we use PCA for the separation of maternal and fetal electrocardiography (ECG) signals in abdominal ECG (aECG) data recorded on the belly of a pregnant woman. Due to the low signal strength of fetal ECG (fECG) signals it is an "algorithmic challenge" to properly separate fECG from much stronger maternal ECG (mECG) signals [1].

The present example uses a simplified version of the method proposed by Varanini et al. [2].

### 1.2 References

- [1] R. Kahankova et al., "A Review of Signal Processing Techniques for Non-Invasive Fetal Electrocardiography," IEEE Reviews in Biomedical Engineering, vol. 13, pp. 51–73, 2020, doi: 10.1109/RBME.2019.2938061.
- [2] M. Varanini, G. Tartarisco, L. Billeci, A. Macerata, G. Pioggia, and R. Balocchi, "An efficient unsupervised fetal QRS complex detection from abdominal maternal ECG," Physiol. Meas., vol. 35, no. 8, pp. 1607–1619, Aug. 2014, doi: 10.1088/0967-3334/35/8/1607.
- [3] Source of data: https://physionet.org/content/challenge-2013/1.0.0/

```
[1]: import numpy as np
  import pandas as pd
  import matplotlib.pyplot as plt
  from scipy.signal import filtfilt, butter
  from biosppy.signals import ecg
  from sklearn.decomposition import PCA

import plotly.graph_objects as go
  from plotly.offline import init_notebook_mode, iplot
  from plotly.subplots import make_subplots
  init_notebook_mode(connected=True) # initiate notebook for offline plot

import sys
  sys.path.append(r"..")
  from mqrs_utils import cancel_mqrs
```

```
[2]: # load abdominal ECG (aECG) data
     # transformed from initial source: https://physionet.org/content/challenge-2013/
      \hookrightarrow 1.0.0/set-a/a13.dat
     filename = '../aecg a13.hdf5'
     aecg = pd.read_hdf(filename, key='signals').values
     fs = 1000
     t = np.arange(aecg.shape[0]) / fs
[3]: # bandpass filter data
     b, a = butter(4, np.asarray([3, 45])/(fs/2), btype='bandpass')
     aecg = filtfilt(b, a, aecg.transpose()).transpose()
     # plot
     fig = go.Figure()
     for i in range(aecg.shape[1]):
         fig.add_trace(go.Scatter(x=t, y=aecg[:, i], name='AECG{:d}'.format(i)))
     fig.update_xaxes(title='Time (s)')
     fig.update yaxes(title='AECG Amplitude (A.U.)')
     fig.update_layout(title='Bandpass-Filtered AECG Signals')
     fig.show()
[4]: # apply first PCA for enhancing maternal ECG component
     pca1 = PCA()
     pc1 = pca1.fit_transform(aecg)
     # maternal ECG as the first principal component, note that this
     # remains a quess and would need to be automated in the final solution
     maternal_ecg = pc1[:, 0]
     # detect maternal QRS peaks
     ts, filtered, mqrs_peaks = ecg.ecg(maternal_ecg, fs, show=False)[:3]
     # plot
     fig = go.Figure()
     for i in range(pc1.shape[1]):
         fig.add_trace(go.Scatter(x=t, y=pc1[:, i], name='PC1[:,{:d}]'.format(i)))
         if i == 0:
             fig.add_trace(go.Scatter(x=t[mqrs_peaks], y=pc1[mqrs_peaks, i],_

¬name='mQRS-Peaks',
                                      mode='markers', marker_color='red',__
      →marker_symbol='circle-open'))
     fig.update_xaxes(title='Time (s)')
     fig.update yaxes(title='PC1 (A.U.)')
     fig.update_layout(title='Principal Components of First PCA Used to Enhance mECG_

Signal')
     fig.show()
```

```
\hookrightarrowsignal
     x_residual, mecg_estimations = cancel_mqrs(fs, pc1, mqrs_peaks.astype(np.

¬float64))
     # plot
     fig = make_subplots(rows=3, cols=1, shared_xaxes=True)
     fig.add_trace(go.Scatter(x=t, y=pc1[:,0], name='Maternal ECG'), row=1, col=1)
     fig.add_trace(go.Scatter(x=t[mqrs_peaks], y=pc1[mqrs_peaks, 0],__
      ⇔name='mQRS-Peaks', marker_color='red',
                              legendgroup='mQRS', mode='markers',
      →marker_symbol='circle-open'), row=1, col=1)
     fig.add_trace(go.Scatter(x=t, y=mecg_estimations[:, 0], name='Interpolated mQRS_L
      ⇒Signal'), row=2, col=1)
     fig.add_trace(go.Scatter(x=t[mqrs_peaks], y=mecg_estimations[mqrs_peaks, 0],__

¬name='mQRS-Peaks', marker_color='red',
                              legendgroup='mQRS', showlegend=False, mode='markers', __
      →marker_symbol='circle-open'), row=2, col=1)
     fig.add_trace(go.Scatter(x=t, y=x_residual[:, 0], name='mQRS-free Signal'),u
      \rightarrowrow=3, col=1)
     fig.add_trace(go.Scatter(x=t[mqrs_peaks], y=x_residual[mqrs_peaks, 0],_u

¬name='mQRS-Peaks', marker_color='red',
                              legendgroup='mQRS', showlegend=False, mode='markers', __
      →marker_symbol='circle-open'), row=3, col=1)
     fig.update_xaxes(title='Time (s)', row=3, col=1)
     fig.update_layout(title='Maternal QRS Cancellation')
     fig.show()
[6]: # apply second PCA for enhancing fetal ECG component in residual signal
     pca2 = PCA()
     pc2 = pca2.fit_transform(x_residual)
     # fetal ECG as the first principal component, note that this
     # remains a guess and would need to be automated in the final solution
     fetal_ecg = pc2[:, 0]
     # detect fetal QRS peaks
     ts, filtered, fqrs_peaks = ecg.ecg(fetal_ecg, fs, show=False)[:3]
     # plot
     fig = go.Figure()
     for i in range(pc2.shape[1]):
         fig.add_trace(go.Scatter(x=t, y=pc2[:, i], name='PC2[:,{:d}]'.format(i)))
         if i == 0:
             fig.add_trace(go.Scatter(x=t[fqrs_peaks], y=pc2[fqrs_peaks, i],_

¬name='fQRS-Peaks',
```

[5]: # remove maternal QRS complexes from signal to obtain a best possible fetal ECG\_

```
mode='markers', marker_color='black', u

smarker_symbol='triangle-down-open'))

fig.update_xaxes(title='Time (s)')

fig.update_yaxes(title='PC2 (A.U.)')

fig.update_layout(title='Principal Components of Second PCA Used to Enhanceu

sfECG Signal')

fig.show()
```

```
[7]: # plot for summarizing all
     fig = make_subplots(rows=2, cols=1, shared_xaxes=True)
     # maternal ECG with mQRS
     fig.add_trace(go.Scatter(x=t, y=maternal_ecg, name='Maternal_ECG'), row=1,__
      \hookrightarrowcol=1)
     fig.add_trace(go.Scatter(x=t[mqrs_peaks], y=maternal_ecg[mqrs_peaks],_

¬name='mQRS-Peaks',
                              marker color='red', mode='markers',
     →marker_symbol='circle-open'), row=1, col=1)
     # fetal ECG with fQRS
     fig.add_trace(go.Scatter(x=t, y=fetal_ecg, name='Fetal_ECG'), row=2, col=1)
     fig.add_trace(go.Scatter(x=t[fqrs_peaks], y=fetal_ecg[fqrs_peaks],_u
      marker_color='black', mode='markers',__
     →marker_symbol='triangle-down-open'), row=2, col=1)
     fig.update_xaxes(title='Time (s)', row=2, col=1)
     fig.update_layout(title='Maternal vs. Fetal ECG')
     fig.show()
```

```
[8]: from IPython.display import display, Math, Latex display(Latex(r"\newpage"))
```

# 2 Exercise Questions

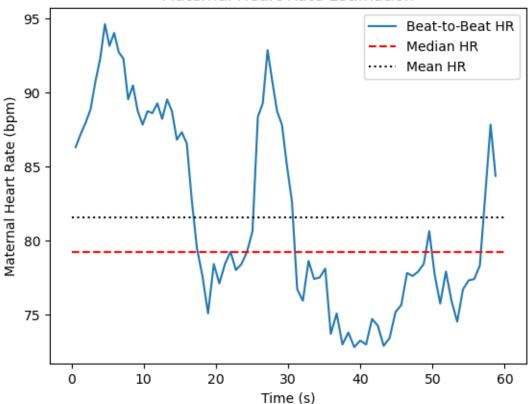
Please provide your answers directly below each question.

## 2.1 Question 1

Determine the maternal heart rate, both expressed in Hz and beats/min.

```
[9]: # estimate mgrs
      mrr = np.diff(mqrs_peaks) / fs
      mhr_median = 60 / np.median(mrr)
      mhr_mean = 60 / np.mean(mrr)
      print('Maternal HR: mean={:.1f} bpm; median={:.1f} bpm'.format(mhr_mean,__
       →mhr_median))
      print('Maternal HR: mean={:.1f} Hz; median={:.1f} Hz\n'.format(mhr_mean/60,_
       →mhr_median/60))
     Maternal HR: mean=81.6 bpm; median=79.3 bpm
     Maternal HR: mean=1.4 Hz; median=1.3 Hz
[10]: plt.plot(t[mqrs_peaks][:-1], 60/np.diff(t[mqrs_peaks]), label='Beat-to-Beat HR')
      plt.plot(t[[0, -1]], [60/np.median(np.diff(t[mqrs_peaks]))]*2, '--r', __
       ⇔label='Median HR')
      plt.plot(t[[0, -1]], [60/np.mean(np.diff(t[mqrs_peaks]))]*2, ':k', label='Mean_
       →HR')
      plt.legend()
      plt.xlabel('Time (s)')
      plt.ylabel('Maternal Heart Rate (bpm)')
      plt.title('Maternal Heart Rate Estimation')
      plt.show()
```





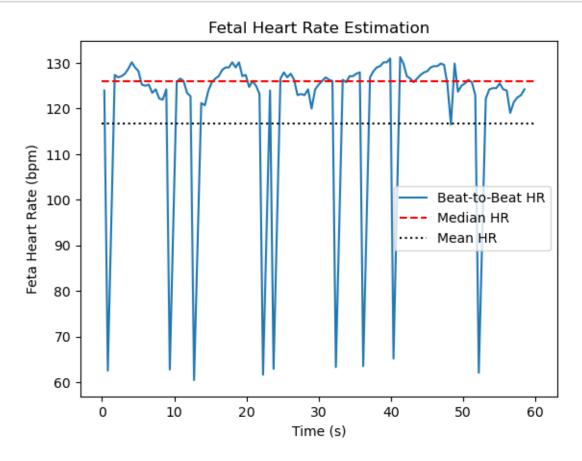
## 2.2 Question 2

Determine the fetal heart rate, both expressed in Hz and beats/min.

Fetal HR: mean=116.7 bpm; median=125.9 bpm Fetal HR: mean=1.9 Hz; median=2.1 Hz

Due to missing detections of certain fQRS peaks the the mean value is underestimating the real  ${\tt HR}$ .

Therefore the median value shall be used.



### 2.3 Question 3

Determine the following three values:

- i) the average amplitude of the maternal QRS peaks (mQRS);
- ii) the average amplitude of the fetal QRS peaks (fQRS);
- iii) the ratio between the average amplitudes of i) mQRS and ii) fQRS peaks.

```
Maternal MQRS Amplitude: mean=103.2; median=102.1
Fetal MQRS Amplitude: mean=20.1; median=19.9
Ratio Fetal/Maternal QRS Amplitude: 0.195
Ratio Maternal/Fetal QRS Amplitude: 5.14
```

## 2.4 Question 4

How many of the principal components of the first PCA clearly show a maternal ECG signal? Which ones?

The first two: PC1[:, 0] and PC1[:, 1]

### 2.5 Question 5

How many of the principal components of the second PCA clearly show a fetal ECG signal? Which ones?

The first one: PC2[:, 0]

#### 2.6 Question 6

Not all of the fetal QRS peaks seem to be detected properly. Do you have an explanation why this happens and under which circumstances? Is it a problem of the fQRS detector, the mQRS cancellation or of another block of the algorithm?

Mainly all fQRS peaks which do not get properly detected occur closely to mQRS peaks. This indicates that the mQRS cacellation is not perfect, i.e., it is difficult to separate the mQRS of high amplitude with the fQRS of much smaller amplitude but of very similar morphology. While some missed fQRS are virtually not visible in the fetal\_ecg signal, others are of smaller amplitude than the surroundin fQRS peaks. For the latter the missing detections can thus also be partly attributed to the QRS peak detector.