

DETECTING PULSE FROM HEAD MOTIONS IN VIDEO

CVPR 2013

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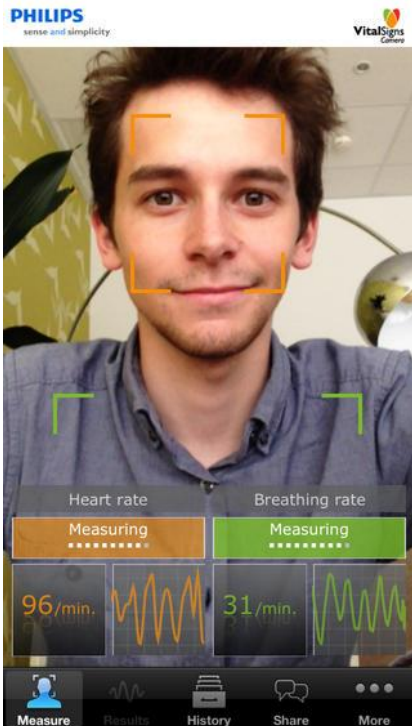
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April 9th, 2014

Motivation

- Diagnosis
 - Pulse rate → acute problems
 - Heart Rate Variability (HRV) → longer-term cardiac assessment
- Electrocardiogram (ECG)
 - Accurate, but...
 - Electrodes must be attached to the skin
- Doing it **non-invasively**
 - Premature newborns
 - Elderly
 - Long-term statistical analyses

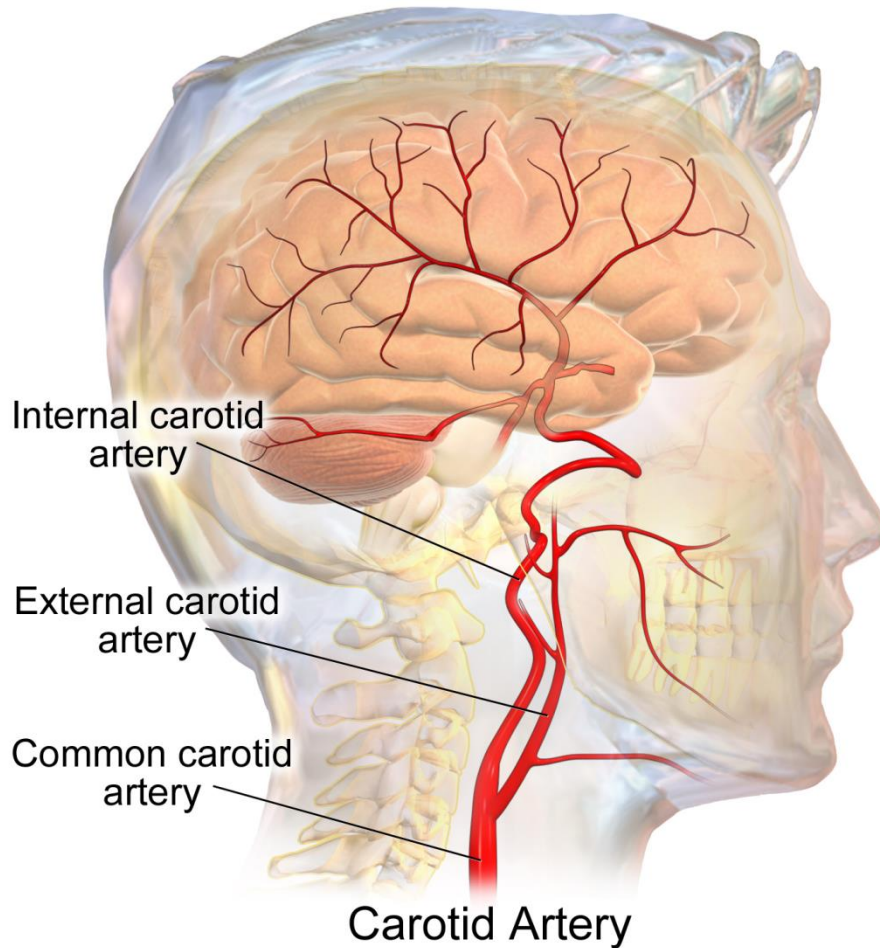
Previous works

- Color changes in the skin



→ Include **head motion** information!

Physiology

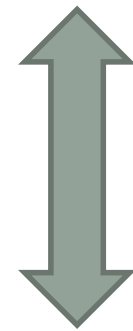


<http://www.allinahealth.org/>

Newton's Third Law
 $F_{\text{blood} \rightarrow \text{head}} = F_{\text{head} \rightarrow \text{blood}}$



$F_{\text{blood} \rightarrow \text{head}}$

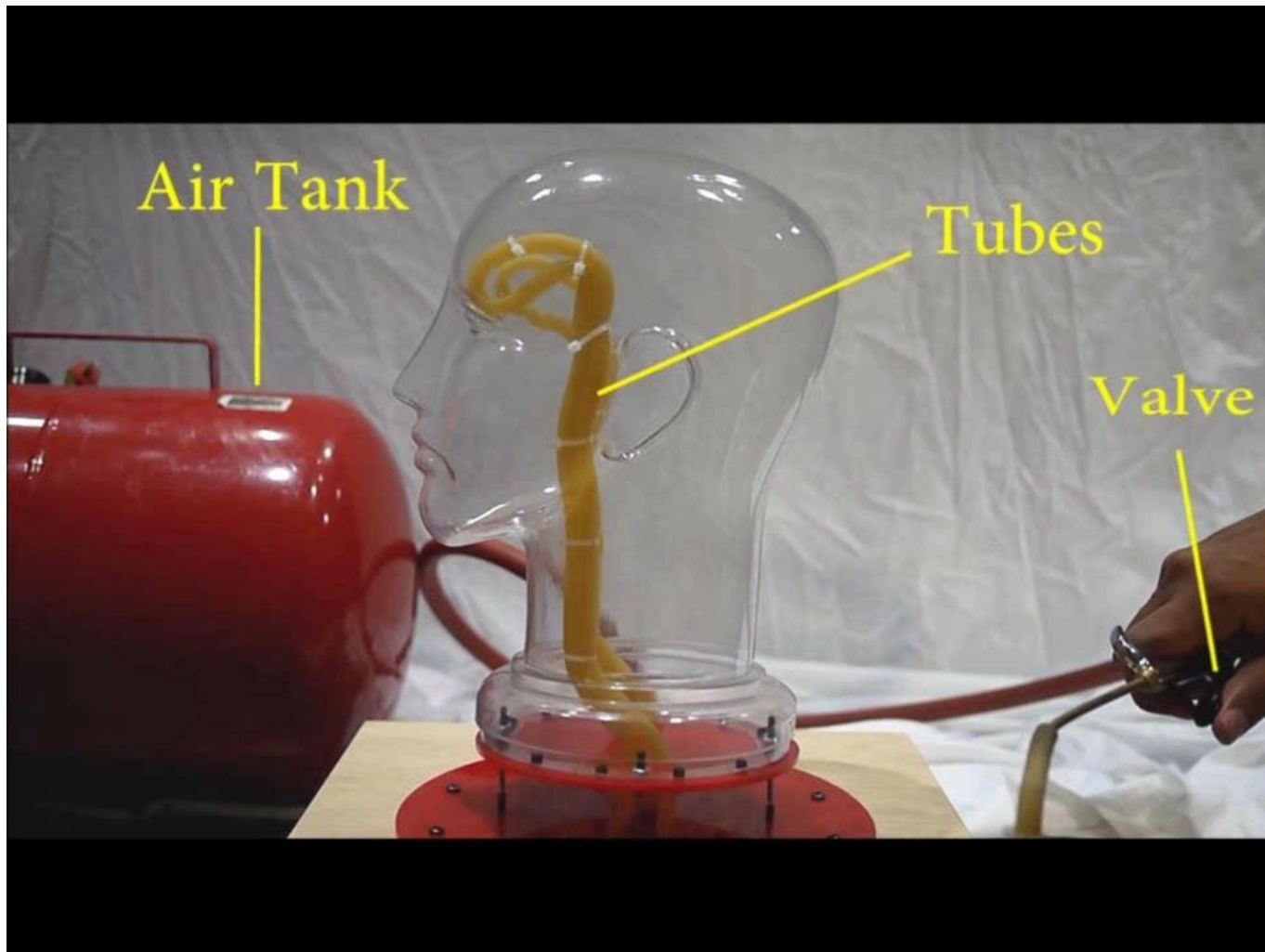


Head motions



Blood pumped to the head

Physiology (2)



Physiology (3)

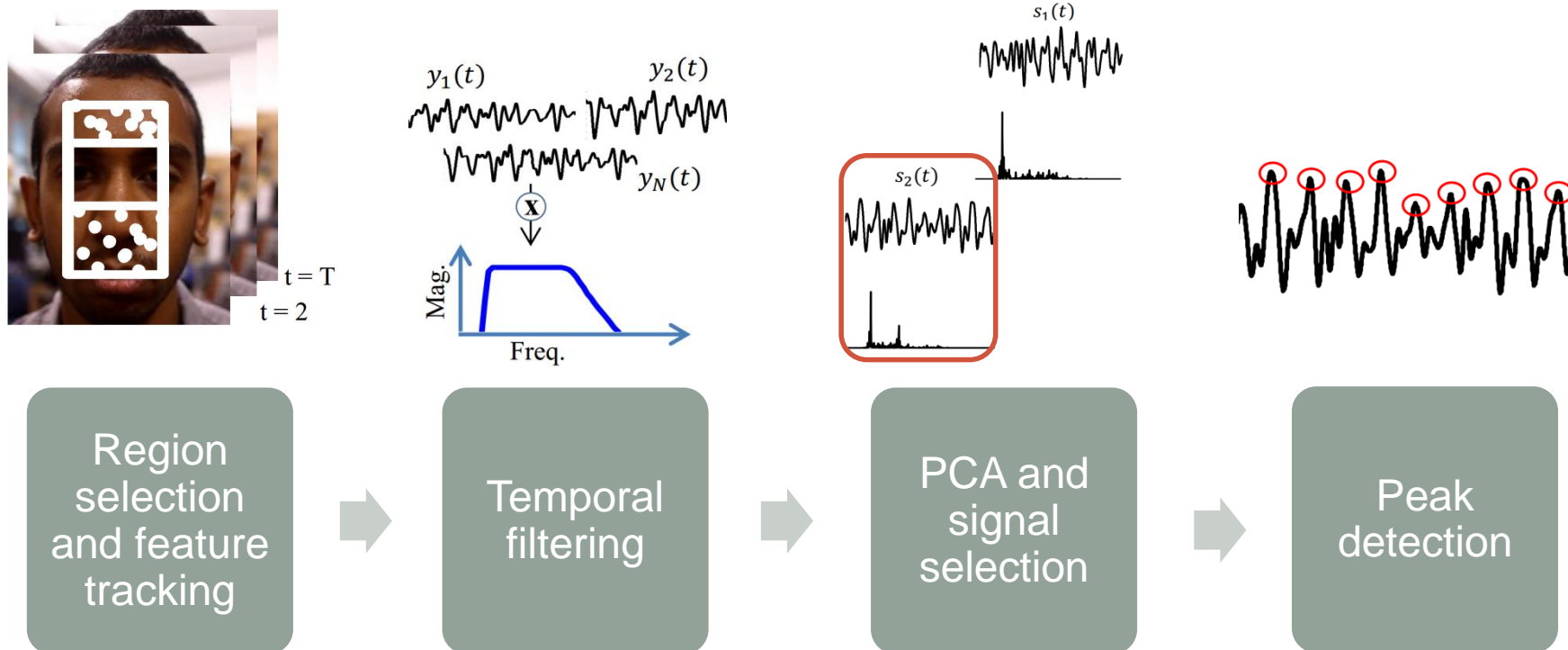
Eulerian Video Magnification (MIT CSAIL 2012)

Input Video



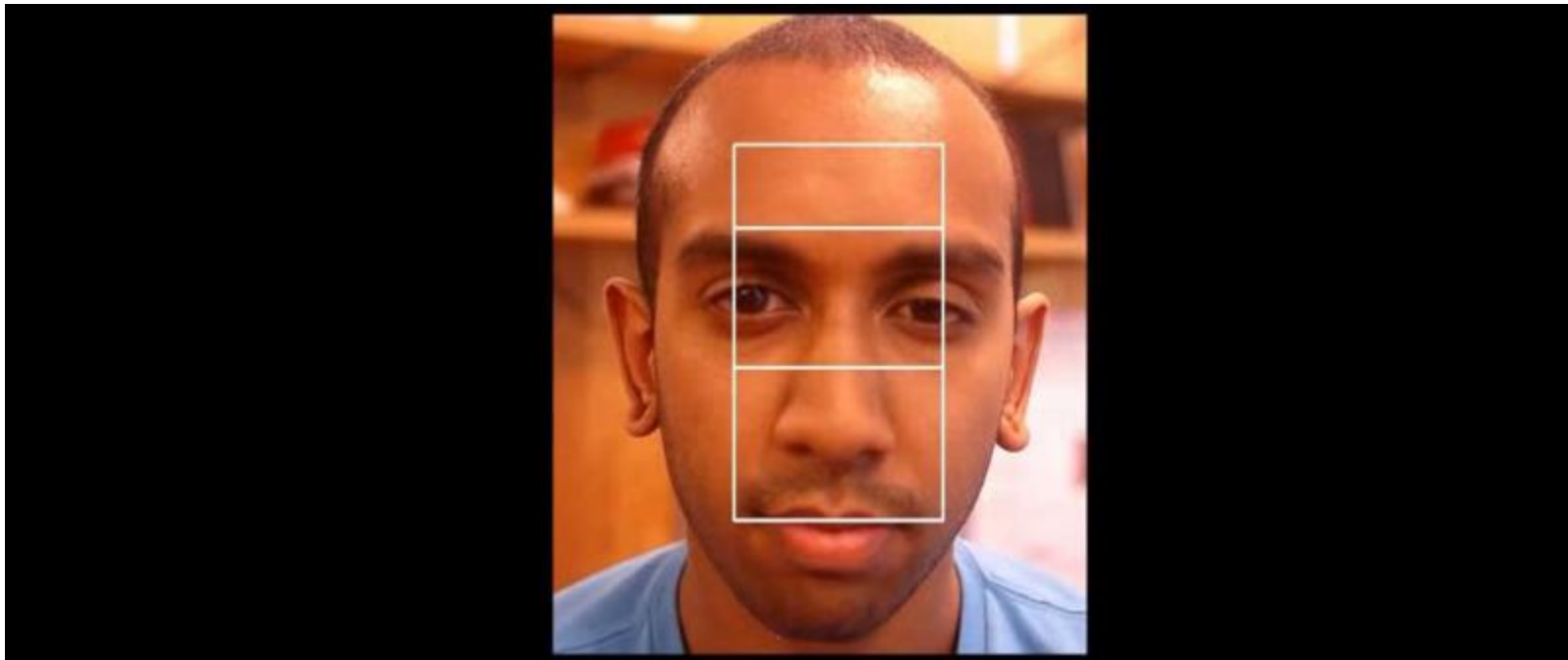
Method

- Whole framework



Method (2)

- Region selection and feature tracking
 - **Viola-Jones** face detector; remove the eyes
 - **Lucas-Kanade** tracker; select stable feature points



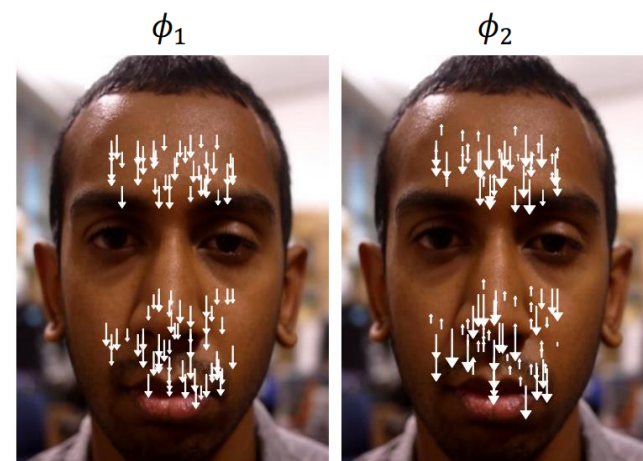
Method (3)

- Temporal filtering
 - 5th order butterworth bandpass filter: **[0.75, 5] Hz**
- PCA decomposition

$$m_t = [y_1(t), y_2(t), \dots, y_N(t)]$$

$$\Sigma_m = \frac{1}{T} \sum_{i=1}^T (m_t - \bar{m})(m_t - \bar{m})^T$$

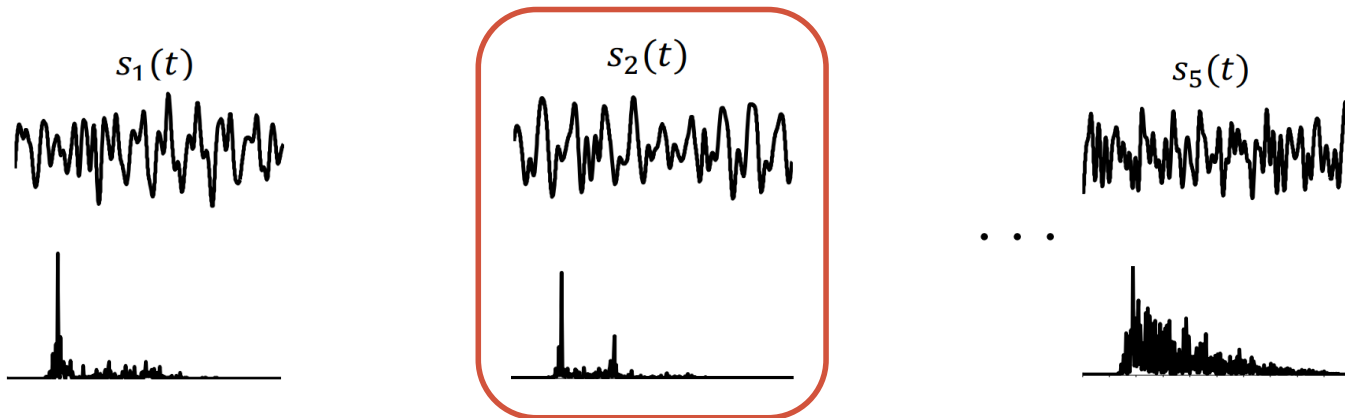
$$\Sigma_m \Phi_m = \Phi_m \Lambda_m$$



$$s_i(t) = \begin{pmatrix} m_1 \\ m_2 \\ \vdots \\ m_T \end{pmatrix} \cdot \phi_i$$

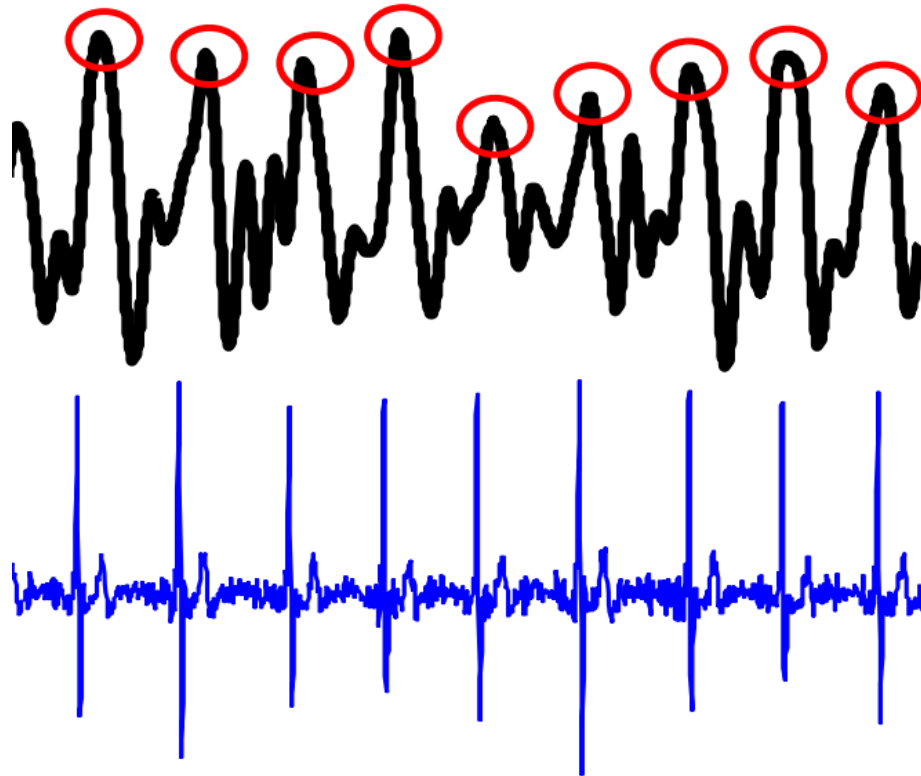
Method (4)

- Signal selection
 - Choose component that is **most periodic**
 - Analyze frequency content



Method (5)

- Peak detection
 - Largest value in a sample-centered window

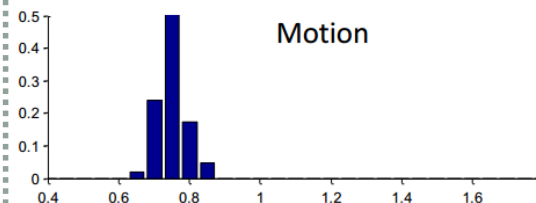
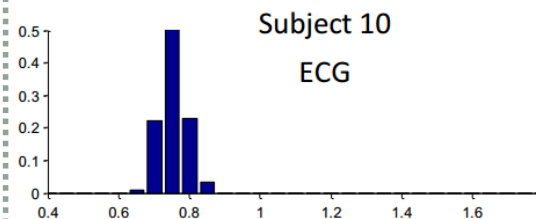
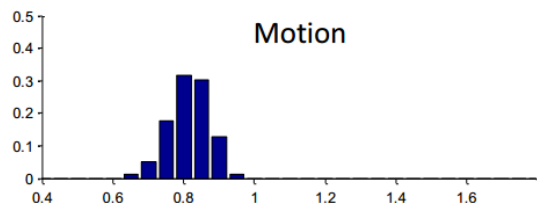
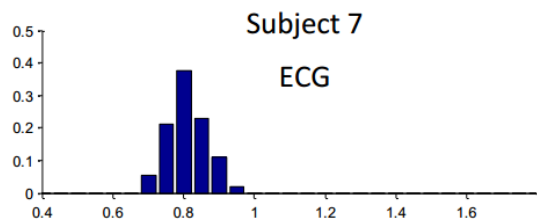
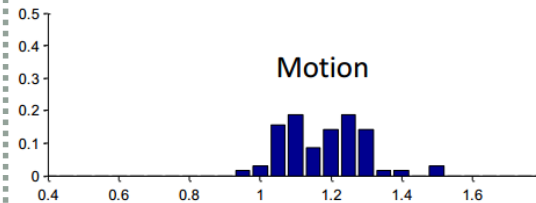
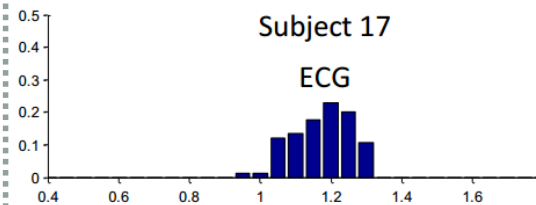
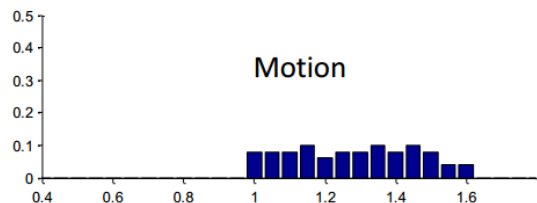
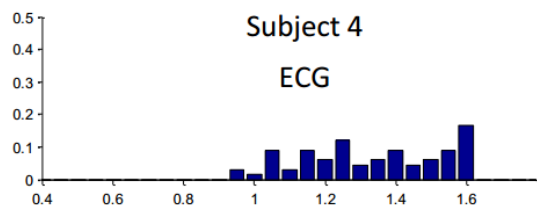


ECG ground truth

Results – Pulse rate

Sub.	Avg. Pulse (beats per minute)		Number of peaks	
	ECG	Motion (% error)	ECG	Motion(% error)
1	66.0	66.0(0)	99	98(1.0)
2	54.7	55.3(1.1)	82	84(2.4)
3	81.3	82.6(1.6)	122	116(4.9)
4	44.7	46.0(2.9)	67	70(4.5)
5	95.3	96.0(0.7)	143	142(0.7)
6	78.9	78.0(1.1)	92	78(15.2)
7	73.3	71.3(2.7)	110	100(9.1)
8	59.3	58.6(1.2)	89	88(1.1)
9	56.7	58.6(3.4)	85	84(1.2)
10	78.7	79.3(0.8)	118	117(0.8)
11	84.7	86.6(2.2)	127	121(4.7)
12	63.3	62.6(1.1)	95	95(0)
13	59.3	60.0(1.2)	89	89(0)
14	60.0	61.3(2.2)	90	89(1.1)
15	80.0	81.3(1.6)	120	114(5.0)
16	74.7	74.6(0.1)	112	110(1.8)
17	50.0	49.3(1.4)	75	76(1.3)
18	77.1	78.8(2.2)	90	85(5.6)

Results – Beat length

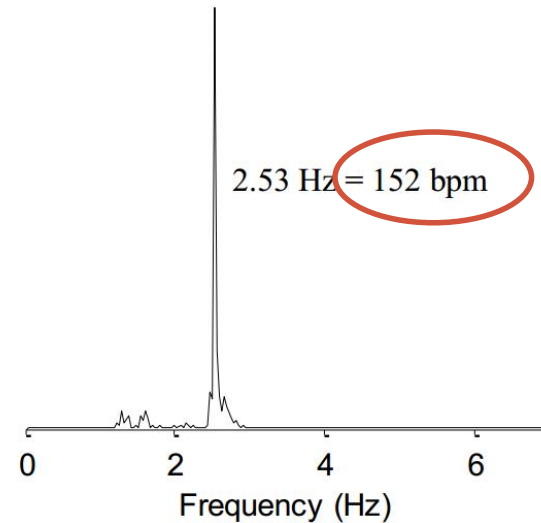
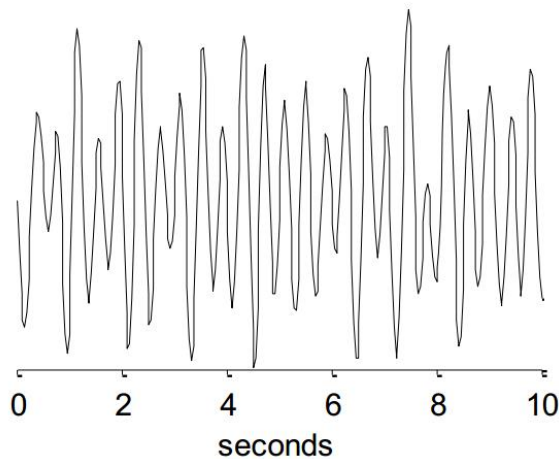


x-axis is time (sec), y-axis is fraction of total beats

Sub.	ECG $\mu(\sigma)$	Motion $\mu(\sigma)$	KS-Test p -value
1	0.91(0.06)	0.90(0.06)	0.89
2	1.08(0.08)	1.06(0.11)	0.52
3	0.73(0.04)	0.73(0.08)	0.05
4	1.34(0.19)	1.28(0.18)	0.14
5	0.62(0.03)	0.63(0.07)	<0.01
6	0.76(0.04)	0.76(0.04)	0.64
7	0.81(0.05)	0.81(0.06)	0.85
8	1.01(0.04)	1.02(0.09)	0.16
9	1.04(0.07)	1.04(0.11)	0.27
10	0.75(0.04)	0.75(0.04)	0.75
11	0.70(0.06)	0.70(0.08)	0.30
12	0.94(0.08)	0.94(0.09)	0.85
13	0.99(0.04)	0.98(0.12)	<0.01
14	0.99(0.11)	0.98(0.12)	0.47
15	0.74(0.05)	0.75(0.06)	0.95
16	0.80(0.05)	0.80(0.06)	0.60
17	1.18(0.08)	1.18(0.11)	0.70
18	0.76(0.05)	0.76(0.06)	0.24

Results – Applications

- Newborn monitoring



Results – Applications (2)

- Hidden faces
 - No need for the skin to be visible (unlike color-based approaches)



Discussion

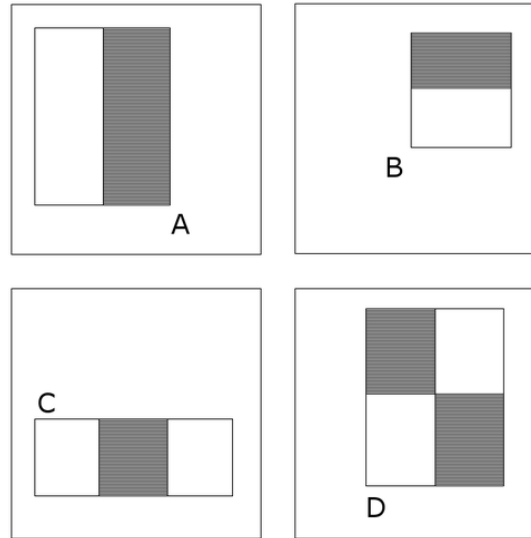
- **Accurate** pulse and beat rate measurements
- Limitations
 - Only **healthy** individuals tested
 - Camera's **sampling rate** (\ll ECG) and **acquisition time**
 - Pulse **variability** from the aorta to the head
 - **Lighting** conditions
- Possible future improvements
 - Filtering to account for **macroscopic head movements**
 - Other motion-derived cardiac metrics?
 - Combine **motion and color** information

Thank you!

<http://newsoffice.mit.edu/2013/seeing-the-human-pulse-0620>

Extra slides – Viola Jones

- Haar-like features
 - Integral image



- AdaBoost (feature selection)
- Cascade of classifiers (focus on positive windows)

Extra slides – Lucas-Kanade

- Optical flow equation (brightness constancy assumption)

$$I_x(x, y) \cdot u + I_y(x, y) \cdot v = -I_t(x, y)$$

- One equation, two unknowns (*aperture problem*)
- Use pixel neighborhood (assume displacement is the same)
- Least-squares minimization