



TL;DR You can train a neural network to estimate heart rate from videos without any ground truth data

Introduction

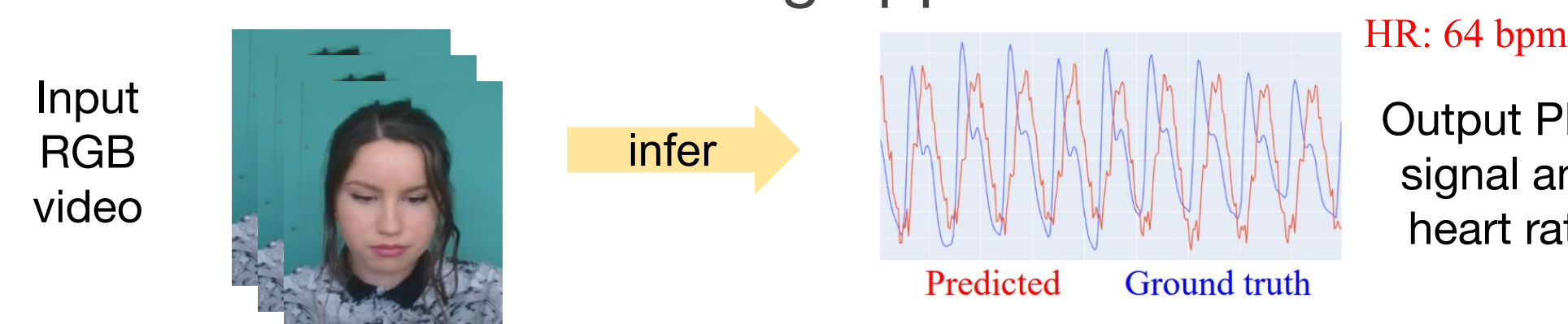
What?

Remote PhotoPlethysmoGraphy ("rPPG")

from a distance using light fluctuations write down measurements

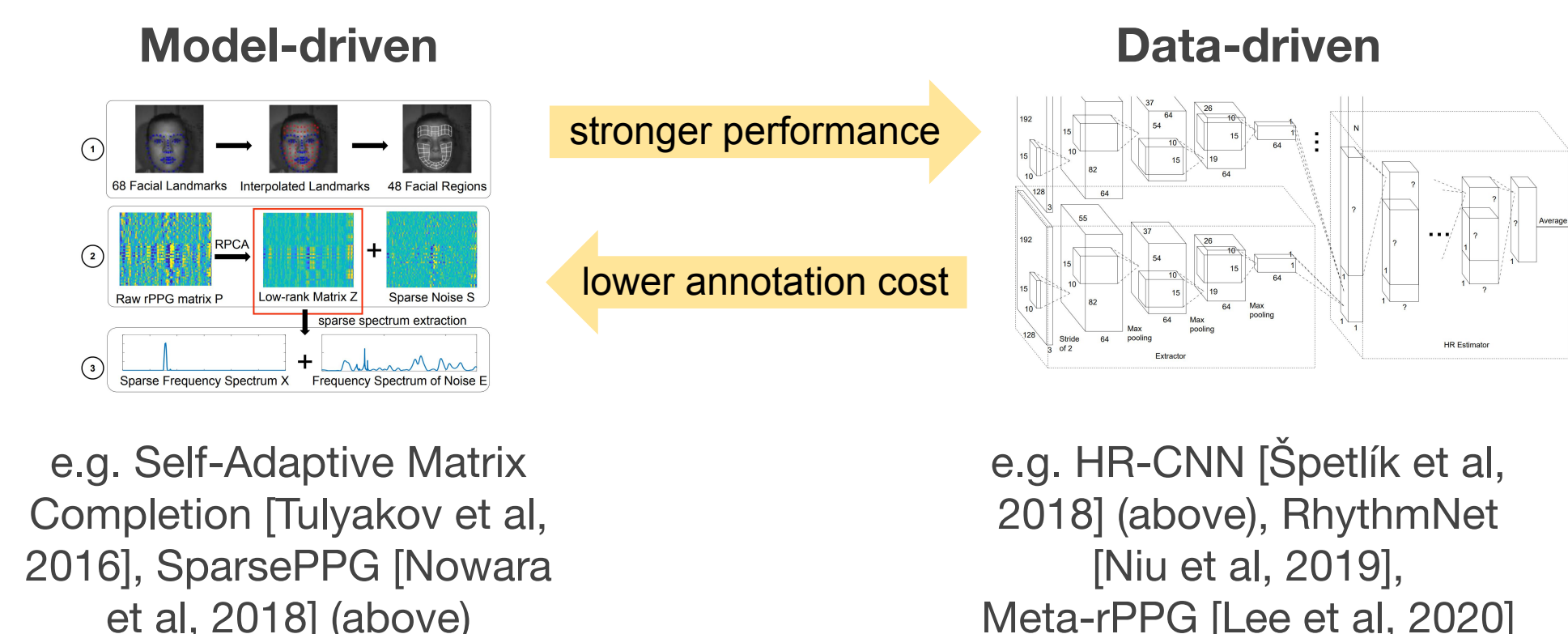
Why?

Health monitoring applications



What's new?

Show how one can train a deep neural network to detect this signal without ground truth training data

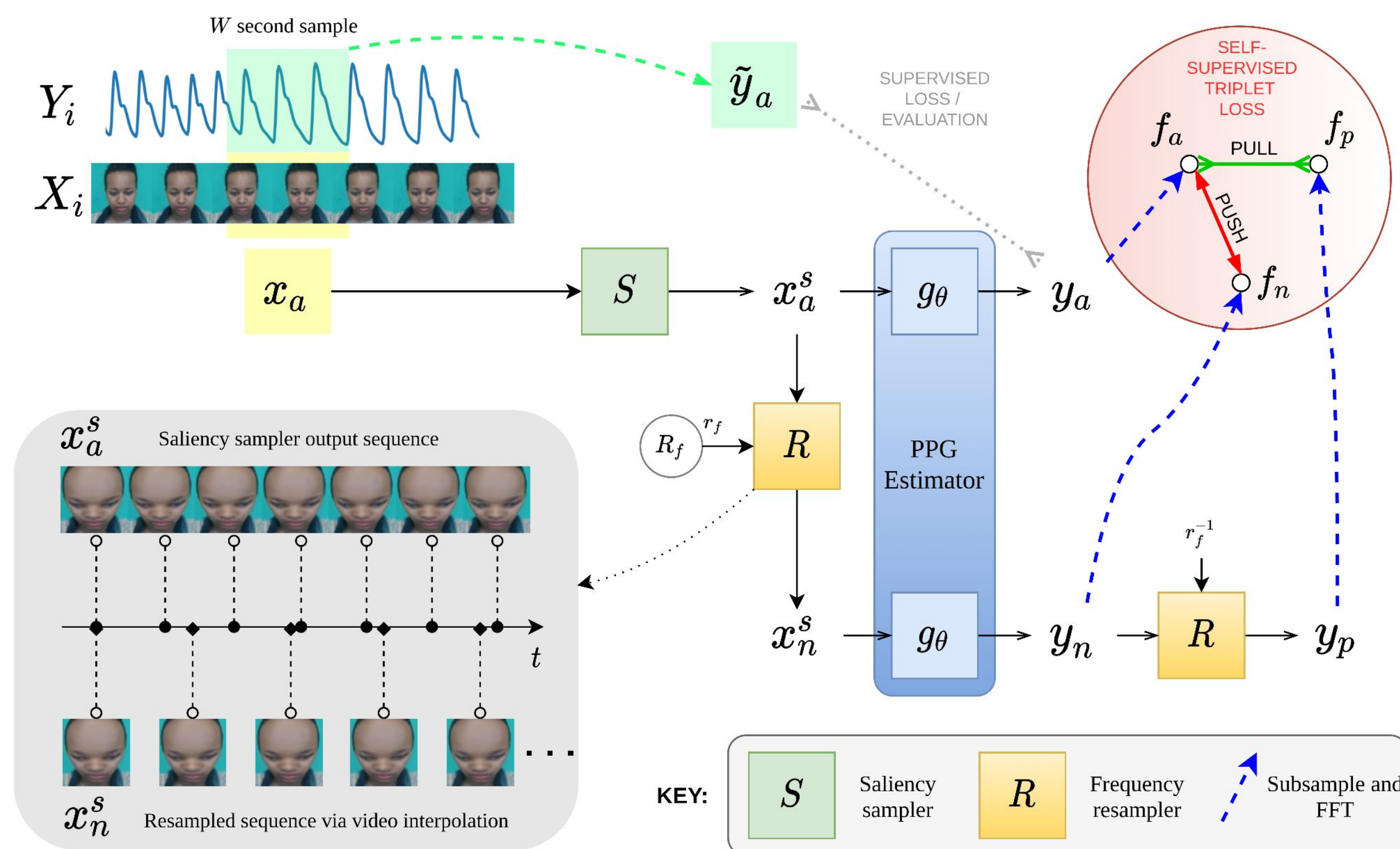
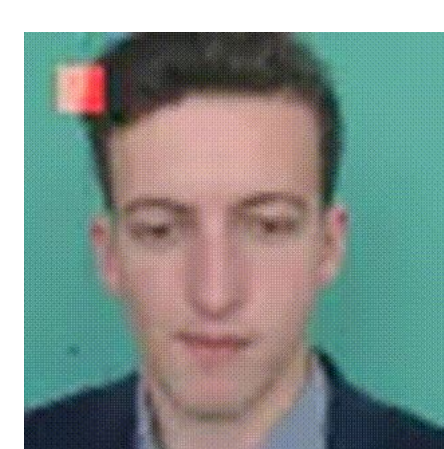
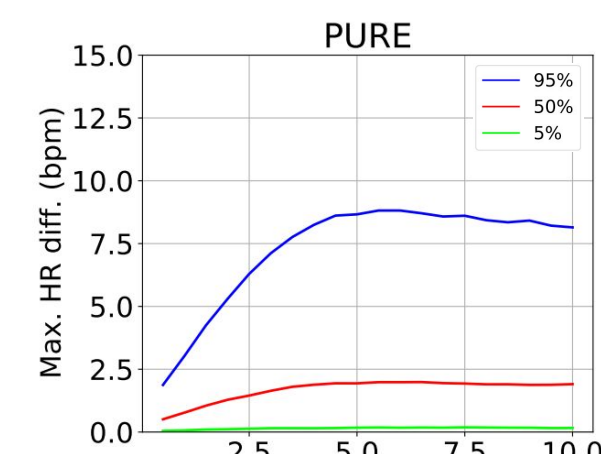
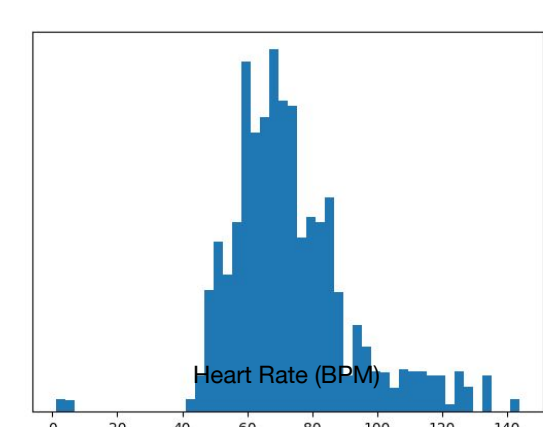


► Strong task performance with zero annotation cost

Our assumptions

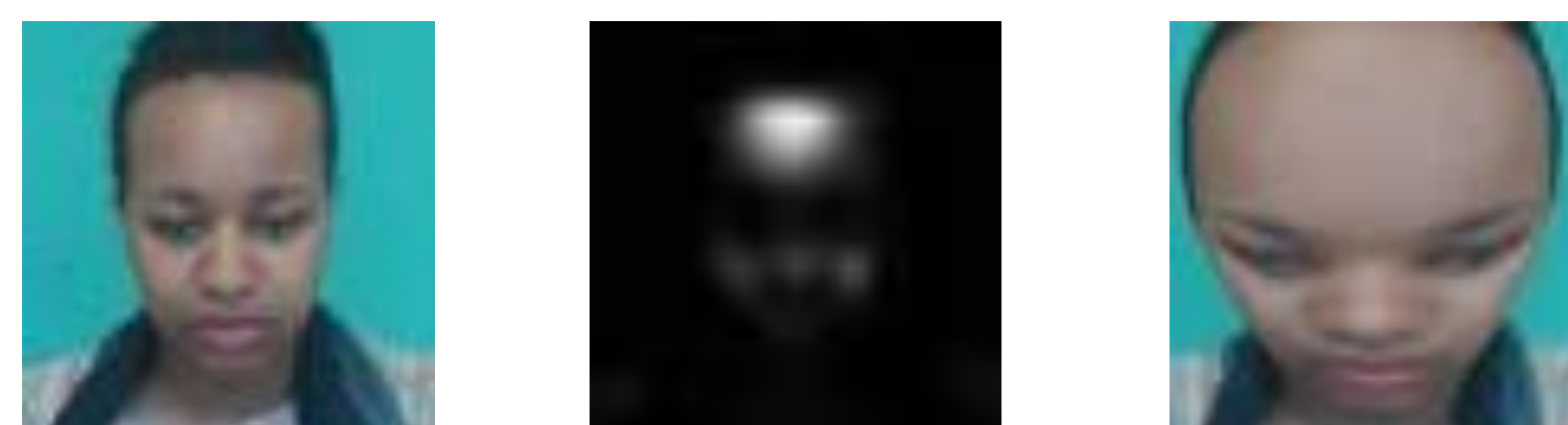
We assume that the signal of interest:

- lies within a certain frequency range (40-250 bpm)
- does not vary rapidly over short time intervals (10s)
- is the dominant visual signal within the target freq. band



7 steps to understand our method

- 10 second input video $\rightarrow x_a$
- Video optionally passed through saliency sampler (S), which learns a task-salient mask while resampling input according to that mask $\rightarrow x_a^s$

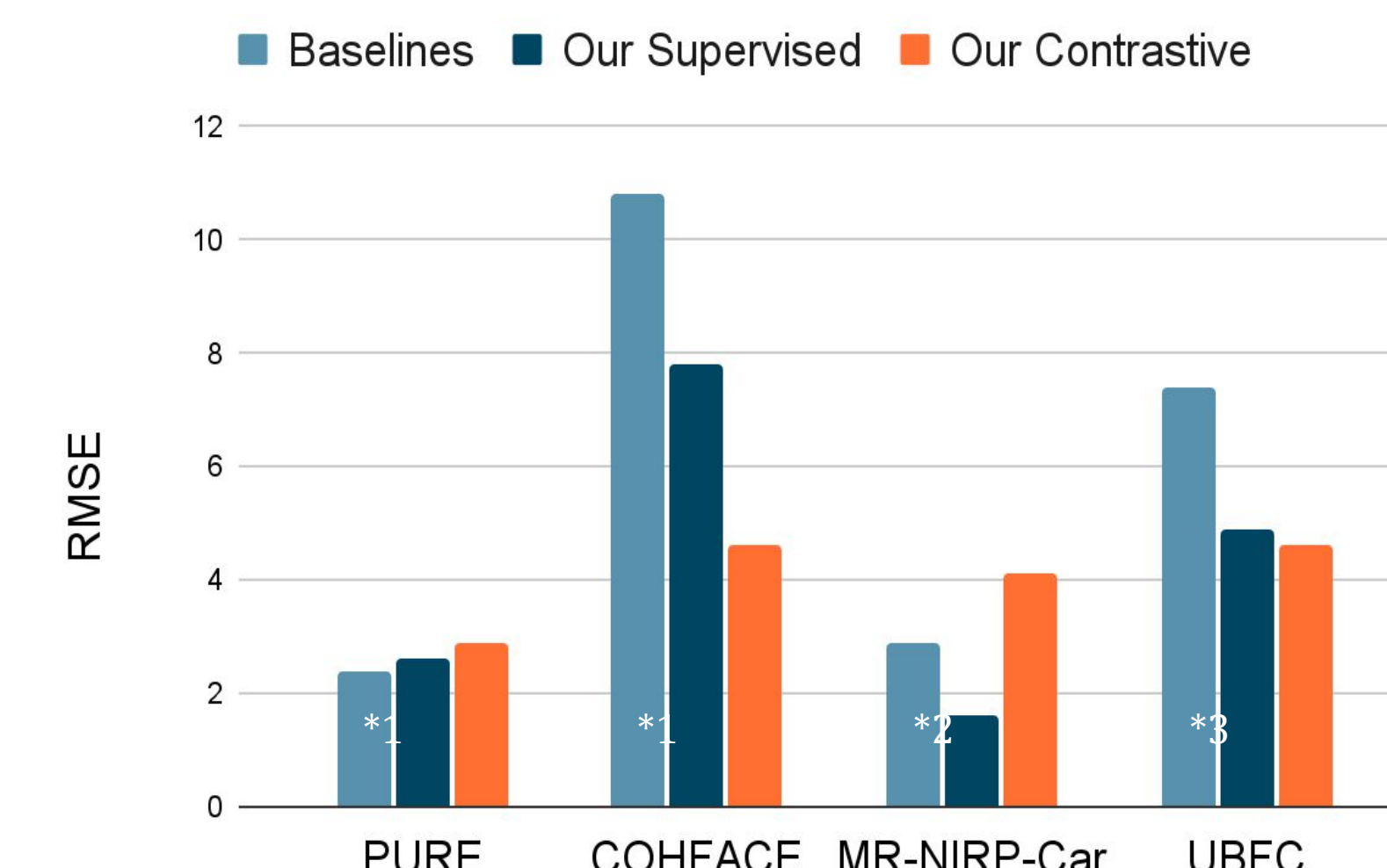


- Video resampled in time using a factor $R_f \rightarrow$ negative sample x_n^s
- Both x_a^s and x_n^s are fed to a PPG estimator (3D CNN) $\rightarrow y_a$ and y_n
- y_n resampled to original rate via $R_f^{-1} \rightarrow$ positive sample y_p
- PPGs y_a, y_n, y_p are randomly subsampled to 5s views and FFTed $\rightarrow f_a, f_n, f_p$
- Triplet loss encourages $\text{dist}(f_a, f_n)$ to be large and $\text{dist}(f_a, f_p)$ to be small

Results

Baselines

- 1) HR-CNN [Špetlík et al, 2018]
- 2) SparsePPG [Nowara et al, 2018]
- 3) Meta-rPPG [Lee et al, 2020]



- Compares favorably to supervised approaches
- Saliency sampler improves interpretability

We will present our winning entry, based on this contrastive approach, to the Vision 4 Vitals workshop challenge on Saturday. Join us to find out more!

Conclusions

- A self-supervised neural network to estimate heart rate from video
- Our approach performs well on existing datasets and is interpretable
- Future extensions: stabilize for input motion and estimate uncertainty

Further results, paper links, and code:

<https://github.com/ToyotaResearchInstitute/RemotePPG>

