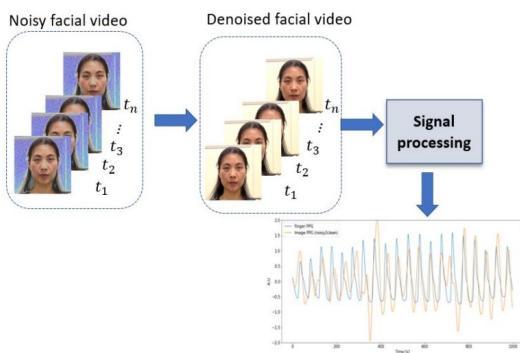


Deep Video Denoising for Facial Signal Processing

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Motivation

- Spatial-temporal denoising. Temporal coherency.
- Video contains much more information.
- FastDVDnet (supervised) and UDVD (unsupervised).
- Apply two advanced video denoising methods to facial videos.
- Ground-truth video is not available.
- Image photoplethysmography (iPPG) reveals the change of blood flow.
- Clean video could reflect blood flow better.



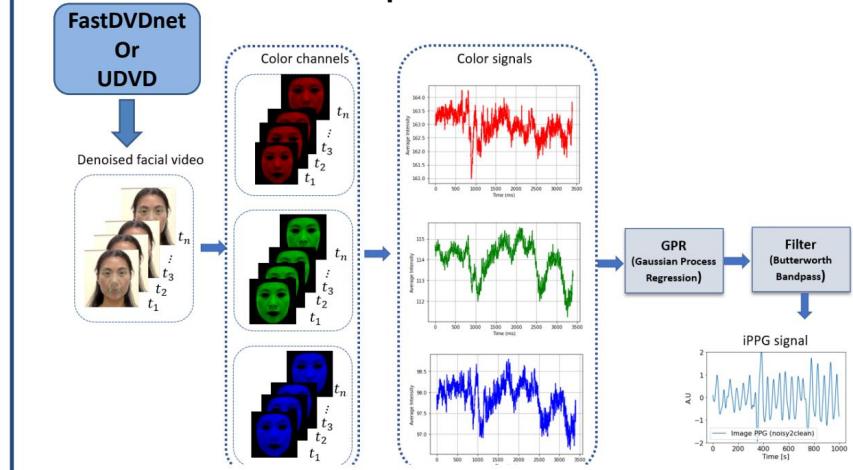
Related Work

- FastDVDnet**
 - Supervised learning
 - Fast, a wide range of noise level.
- UDVD**
 - Unsupervised learning.

References

- [1] M. Tassano, J. Delon, and T. Veit, "Fastdvdnet: Towards real-time deep video denoising without flow estimation," in CVPR, 2020.
- [2] D. Y. Sheth, S. Mohan, J. Vincent, R. Manzorro, P. A. Crozier, M. M. Khapra, E. P. Simoncelli, and C. Fernandez-Granda, "Unsupervised deep video denoising," in ICCV, 2021.
- [3] R. H. Goudarzi, S. Somayeh Mousavi, and M. Charmi, "Using imaging photoplethysmography (ippg) signal for blood pressure estimation," in MVIP, 2020.

Proposed Method



UDVD

$$\mathcal{L}(\mu_x, \Sigma_x) = \frac{1}{2}[(y - \mu_x)^T (\Sigma_x + \sigma^2 I)^{-1} (y - \mu_x)] + \frac{1}{2} \log |\Sigma_x + \sigma^2 I|$$

Where $\mu_x \in R^3$ and $\sigma_x \in R^3$ denote the mean vector and covariance matrix; y is observed noisy pixel

FastDVDnet

$$\mathcal{L}(\theta) = \frac{1}{2m_t} \sum_{j=1}^{m_t} \|\hat{P}_t^j - P_t^j\|^2, \quad (2)$$

where $\hat{P}_t^j = \mathcal{F}((S_t^j, M^j); \theta)$ is the output of the network, and θ is the set of all learnable parameters.

Experimental Results

Results of frame denoising



Results of iPPG extraction

