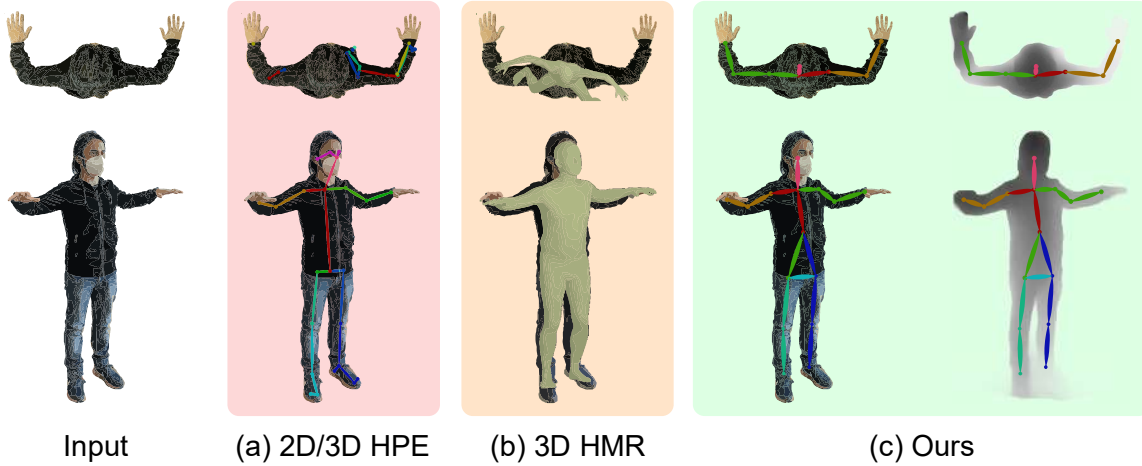


# DECA: Deep viewpoint-Equivariant human pose estimation using Capsule Autoencoders

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## Overview



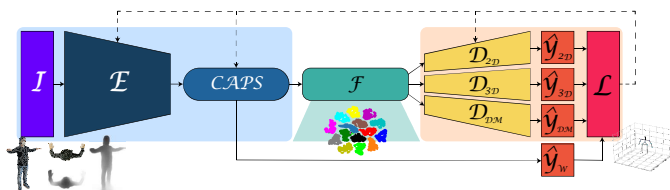
## Context

Current 3D HPE methods suffer a lack of viewpoint equivariance, namely they tend to fail or perform poorly when dealing with viewpoints unseen at training time.

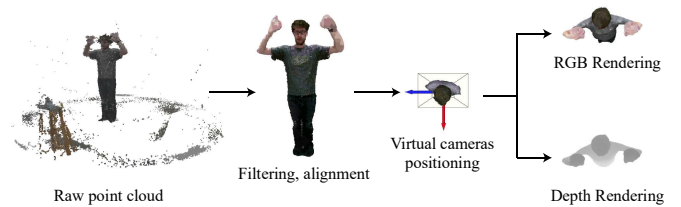
We propose DECA, a novel capsule autoencoder network that allows to drastically reduce the network data dependency at training time, resulting in an improved ability to generalize to unseen viewpoints.

$$\mathcal{L} = \sum_{\tau \in \mathcal{T}} (s_{\tau} + e^{-s_{\tau}} \mathcal{L}_{\tau}) \quad \mathcal{L}_{\tau} = \begin{cases} \mathcal{L}_{2D,3D} = \frac{1}{BS} \sum_{i=0}^{BS} (y_i - \hat{y}_i)^2 \\ \mathcal{L}_{DM} = \frac{\sum_{i=0}^{BS} [mask |y_i - \hat{y}_i| + |y_i - \hat{y}_i|]}{2 * BS} \\ \mathcal{L}_{\mathcal{W}} = \|\hat{y}_{\mathcal{W}} W_{ij}\|_F \end{cases}$$

## Network



## PanopTOP dataset [2]



## Latent space and results

