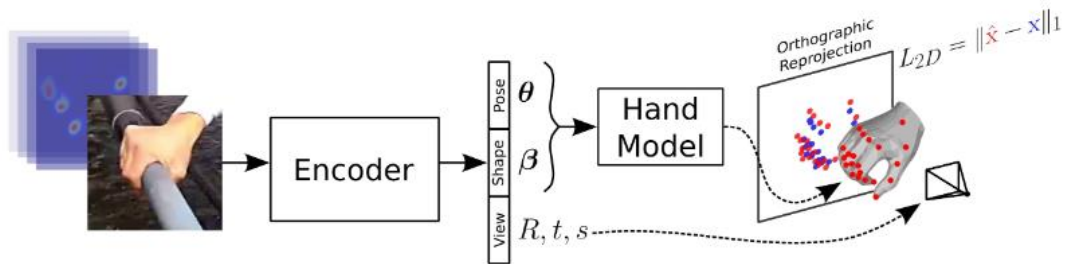


3d hand shape and pose from Images in the wild (2019)



Goal:

- Operate on: in the wild images
- Predict 3d hand shape and pose

Dataset

- Synthetic dataset to pretrain encoder:



- Use decoder, sample poses, shapes, apply rotations, scaling, translation
 - Use 3d scans to generate hand model, render on top of random background images
- Training:
 - Panoptic (15000 images) + MPII (2000) + STEREO (15000)
- Evaluation:
 - Dexter / Object Dataset
 - EgoDexter

Main obstacles:

- Lack of large datasets with reliable ground truth
- Current datasets (3d annotated) not able to generalize network well enough for in the wild pics

MANO:

- Use this model
- Handmodel parameterized by angle of rotations specified for each joint
- 16 joints + 5 for the fingertips

Network

Input:

- Optional: 2D joints detection heatmaps (from independent CNN)
- Monocular RGB images

Encoder:

- Predicts a set of hand view parameters
 - o Hand shape
 - o Pose parameters
 - o Set of view parameters
- Deep convolutional encoder
- Pre-trained with synthetic examples (self-created)
- Use res-net50 network, adjust final fully connected layer

Decoder:

- No training required
- Generate a triangulated 2d mesh and its underlying skeleton
- 2 parts:
 - o Pre-computed articulated mesh deformation hand model that generates a 3d triangulated mesh from the hand parameters (+ underlying skeleton)
 - o Re-projection module controlled by the view parameters (project generated hand into image domain)
 - Use rotation, translation and scale matrix
 - Project orthogonally

Training / loss:

- End to end using 2d and 3d joint annotations without supervision on the hand / view params, except for a regularization on the hand params to ensure their magnitude is small
- Combine multiple losses:
 - o 2d joint reprojection loss (
 - o 3d joint loss
 - o Hand mask loss (binary mask, occlusion aware, 2d)
 - Ground truth generated by GrabCut algo
 - o Model parameter regularization loss (hand model params)

Post processing:

- none