First Project update

Proposed topic

Reconstructing 3D Human Mesh Model from a Single Wild Image

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Summary of work

Completed work

- 1. Study papers about SMPL[1], ResNet[2], HMR[3].
- 2. Set up the environment using Anaconda3:

Tensorflow 2.4, Python 3.6, Other libraries: keras, opency-python, numpy, h5py, ipykernel, matplotlib, scipy, tqdm, pyglet, etc.

3. Test and analysis codes of Resnet50V2 and SMPL method.

3.1 Resnet50V2

From tensorflow.keras.applications.resnet_v2 import ResNet50V2 with pre-trained weights on ImageNet dataset.

3.2 SMPL generator

Construct a generator by regressors and test with the pre-trained regressor model on COCO dataset.

3.2.1 Regression module

Resnet50V2 extracts features [1*2048] to generate SMPL pose_and_shape parameters (24*3 pose and 10 shape parameters) by a regressor. It consists of 2 fully-connected layers with 1024 neurons each with dropout layer (Dropout(0.5)) in between, followed by a final fc layer of 85D neurons.

3.2.2 Blend shape parameters

SMPL generation module converts pose_and_shape parameters to 3D vertices, joints and rotations of the human model. SMPL is based on linear blend skinning (LBS) models. There are 6890 vertices and K=23 joints (23 joints and 1 root orientation) in a human SMPL model. Each vertex t_i in the mesh is linearly controlled by all joints with lbs weights $\omega_{k,i}$.

$$\mathbf{\bar{t}}_i' = \sum_{k=1}^K w_{k,i} G_k'(\vec{\theta}, J(\vec{\beta})) (\mathbf{\bar{t}}_i + \mathbf{b}_{S,i}(\vec{\beta}) + \mathbf{b}_{P,i}(\vec{\theta}))$$

 G_k is the world transformation of joint k. G_k ' is the following transformation after removing G_k due to the rest pose. θ is a pose vector, β is a shape vector, and J is the collection of all pose vectors.

a) For shape blend shapes, B_{S,i} presents different body shapes of people. It is used to infer shape dependent SMPL joints locations by applying smpl_joint_regressor (J_regressor in pretrained paired base_model).

- b) For pose blend shapes, B_{P,i} is used in preparation for the split pose. It adds pose features to SMPL vertices, producing reposed parameters [6890*3] to reshape joints_controlled vertices.
- c) Optimization objective for shape and pose parameters is to minimize the vertex reconstruction error. Shape parameter training: PCA. Pose parameter training: gradient-based, non-negative least squares.
- d) Deforming vertices in a mesh model by dual quaternion skinning from weighted joints (lbs weights are loaded from pretrained paired base_model).

Achieved results

Test Image(from COCO dataset):



• Resnet50V2 extracted features: pose (1, 72)

```
1.16436847e-01 2.08306819e-01 -4.53822553e-01 -5.32653481e-02
-2.65590429e-01 3.11724305e-01 -1.14719130e-01 3.53369489e-02
1.25984192e-01 -3.03998053e-01 -3.64122652e-02 5.81212878e-01
7.08573163e-02 8.50350857e-02 2.01636460e-02 6.76949248e-02
-7.10003972e-02 -1.43410251e-01 2.14893341e-01 -8.87283385e-02
-2.24928170e-01 -3.06966990e-01 2.19144985e-01 5.71423545e-02
-2.15562377e-02 2.09669545e-02 -1.61535442e-01 1.01694748e-01
1.02624893e-01 -5.31504676e-02 2.47309972e-02 -3.05911452e-02
1.38713658e-01 -3.28828335e-01 -1.45027369e-01 -9.00704861e-02
-1.97478533e-02 2.45430976e-01 -1.50686502e-03 3.01675275e-02
-7.75373131e-02 -3.75955105e-02 -3.25453818e-01 -2.77324378e-01
4.18180376e-02 6.98037744e-02 -1.90767422e-02 -6.05258346e-01
1.14554919e-01 -4.92527783e-01 2.32917756e-01 1.33113265e-02
-1.79201871e-01 -3.84014547e-02 5.85794225e-02 1.23685651e-01
-1.20442621e-02 -3.81480977e-02 -1.30594328e-01 -4.52433154e-03
-7.11707696e-02 -8.62428397e-02 4.39000130e-02 7.35650957e-02]]
```

Unexpected complications

- 1. Tensorflow 2.6 doesn't support Mac M1 chip, exports 'Illegal hardware instruction' error.
- 2. Tensorflow 2.6 for Windows can cause 'Value error / InvalidArgumentError' in excuting ResNet50V2 module from keras.

Solution: use tensorflow 2.4 on Windows.

Plan for completion

- 1. Learning the Dual Quaternion Skinning method (SMPL model deformation approach): how shape and pose parameters affect the mesh surface vertices' coordinates (6890*3).
- 2. Add a discriminator to optimize shape and pose parameters based on adversarial learning. By 11/18, second update report.
- 3. Do training and testing on different datasets, parameter adjustment. By 12/7, final report and project presentation.

REFERENCES

- [1] Loper, Matthew, Naureen Mahmood, Javier Romero, Gerard Pons-Moll, and Michael J. Black. "SMPL: A skinned multi-person linear model." ACM transactions on graphics (TOG) 34, no. 6 (2015): 1-16.
- [2] He, Kaiming, Xiangyu Zhang, Shaoqing Ren, and Jian Sun. "Identity mappings in deep residual networks." In European conference on computer vision, pp. 630-645. Springer, Cham, 2016.
- [3] Kanazawa, Angjoo, Michael J. Black, David W. Jacobs, and Jitendra Malik. "End-to-end recovery of human shape and pose." In Proceedings of the IEEE conference on computer vision and pattern recognition, pp. 7122-7131. 2018.