

Img 2 Mesh

Estevam Fernandes Arantes 9763105
João Victor Almeida de Aguiar 8503986

Goal



Regression

\vec{X}



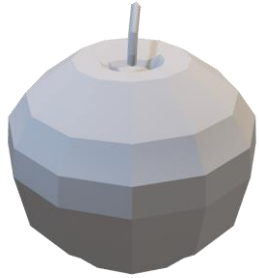
$\hat{f}(\vec{X})$



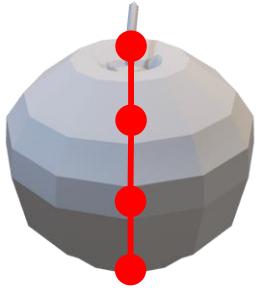
\vec{Y}



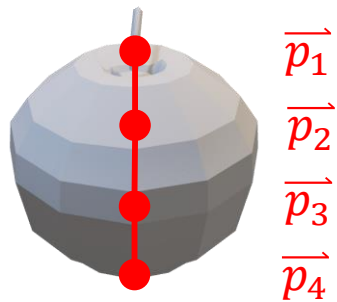
Synthetic Dataset



Synthetic Dataset



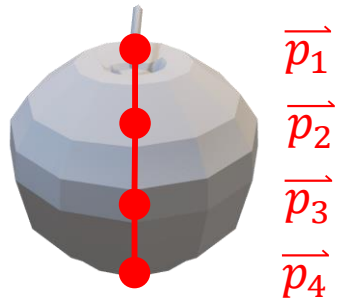
Synthetic Dataset



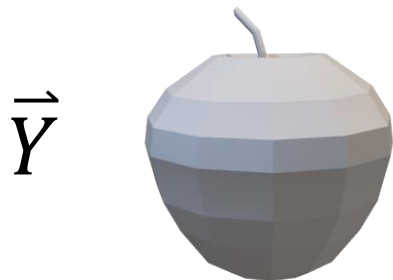
$$\vec{p_i} = \text{random}(\min_i, \max_i)$$



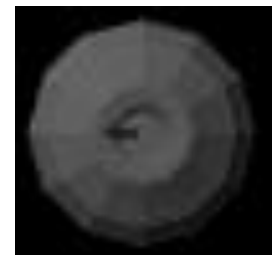
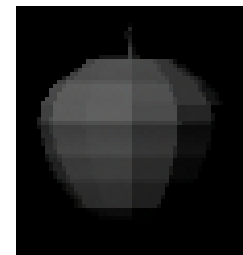
Synthetic Dataset



$\vec{p_i} = \text{random}(\text{min}_i, \text{max}_i)$

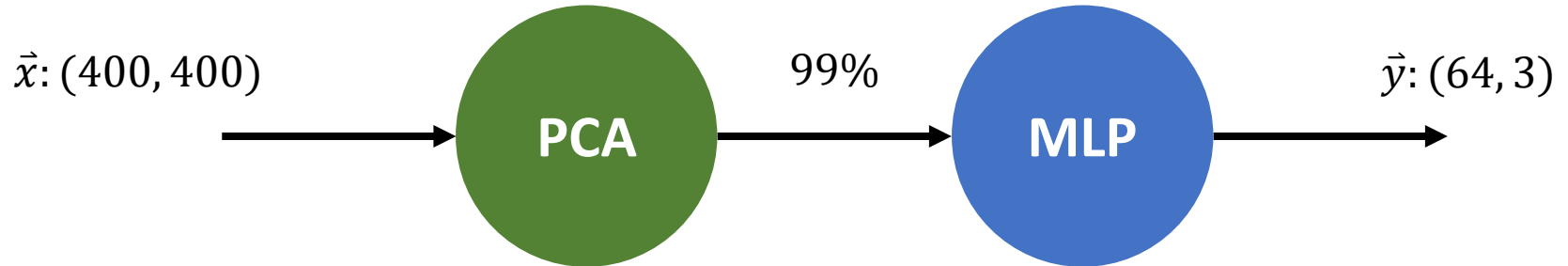


render

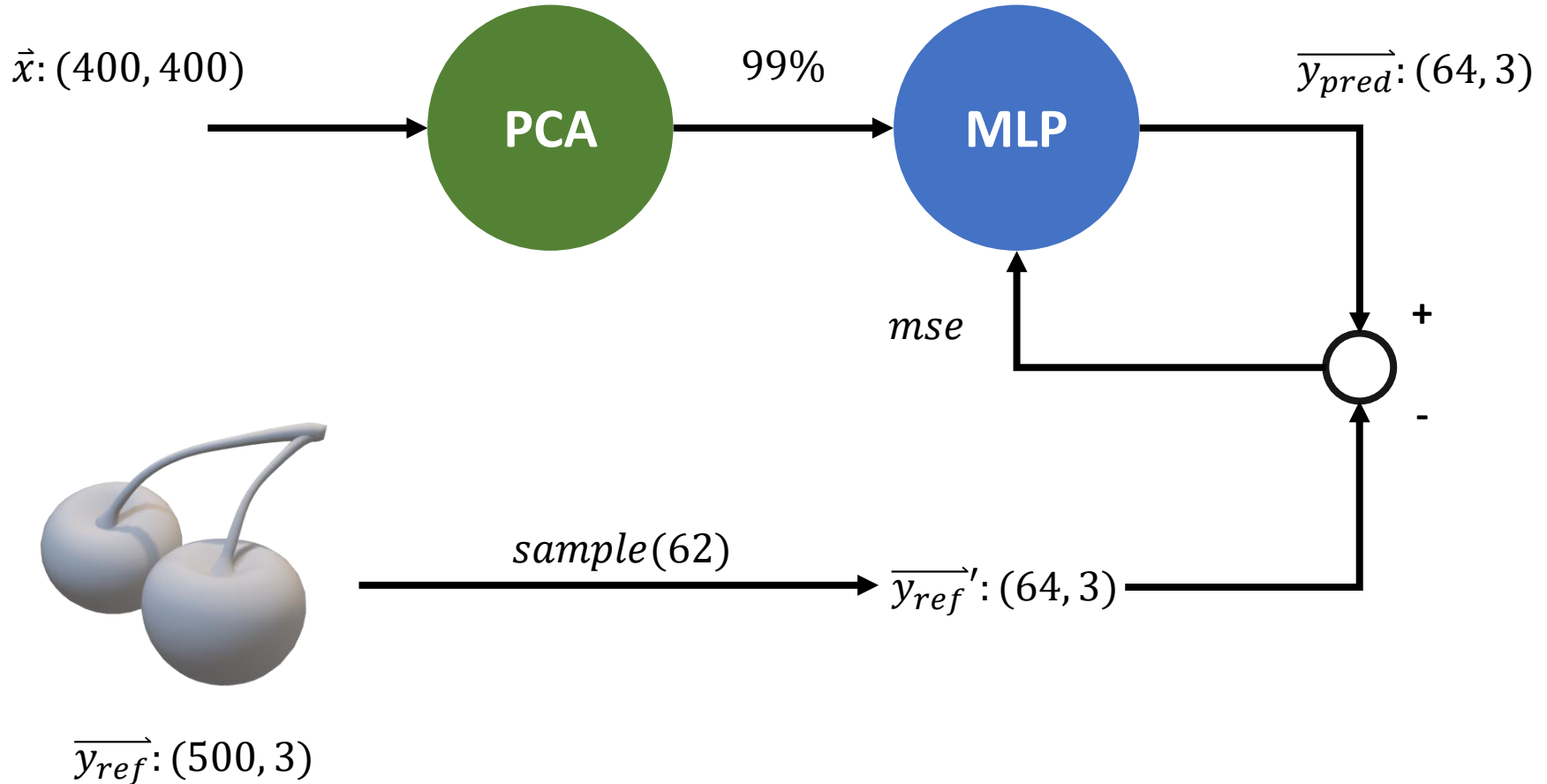


\vec{X}

Naïve Solution

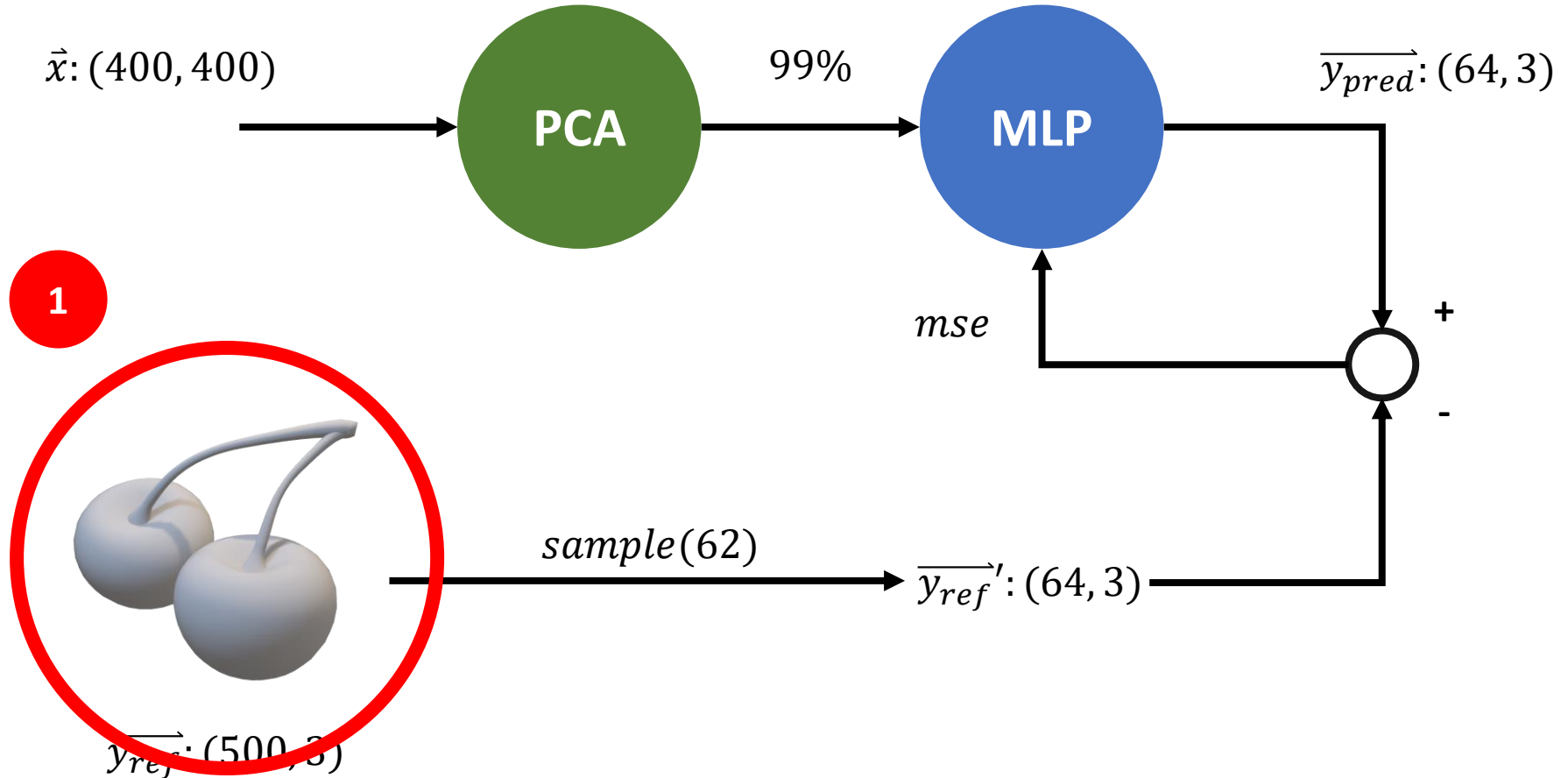


Naïve Solution (Training)



Problem 1 : Dataset

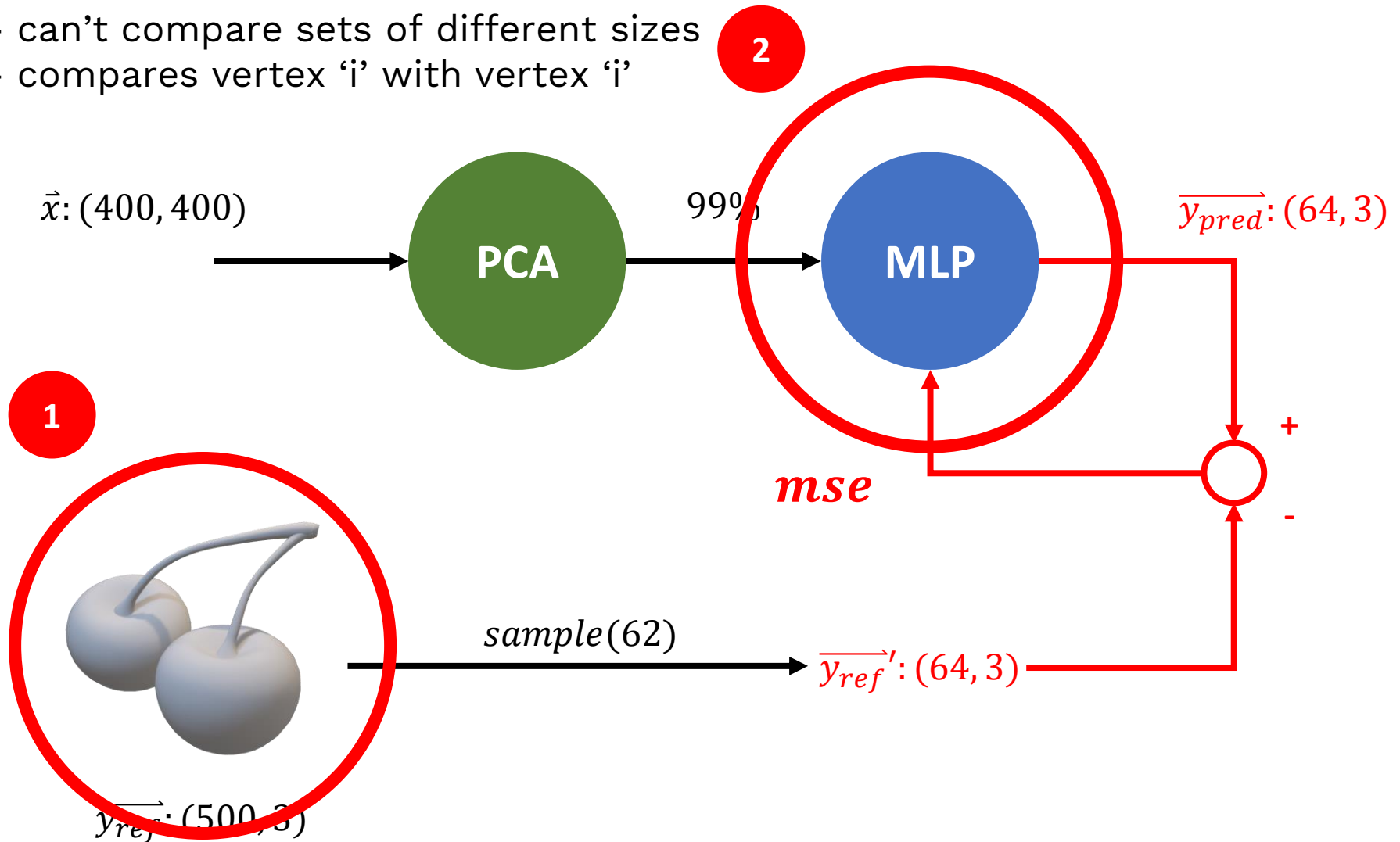
Too biased dataset



Problem 2 : Loss Metric

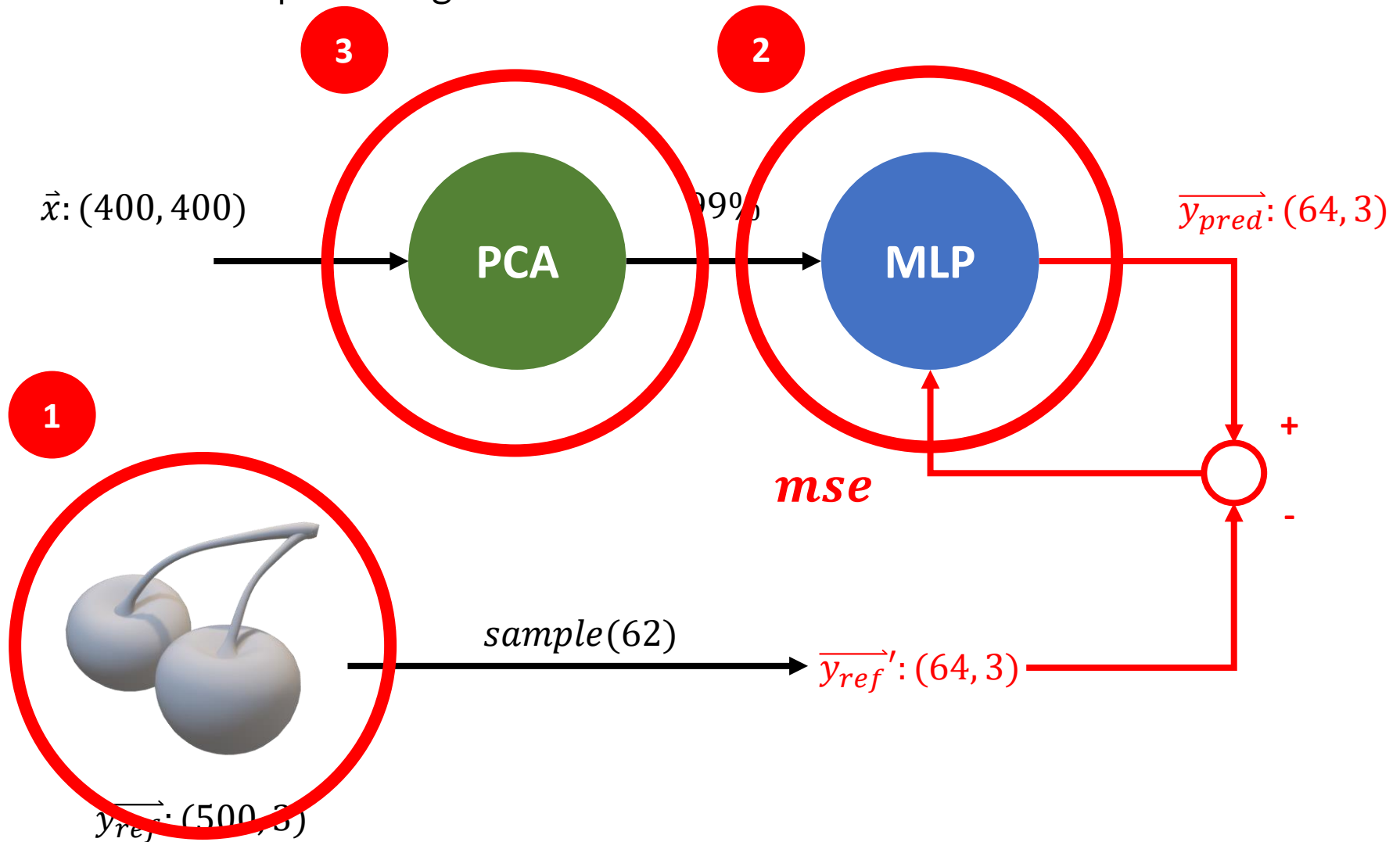
mse

- can't compare sets of different sizes
- compares vertex 'i' with vertex 'i'



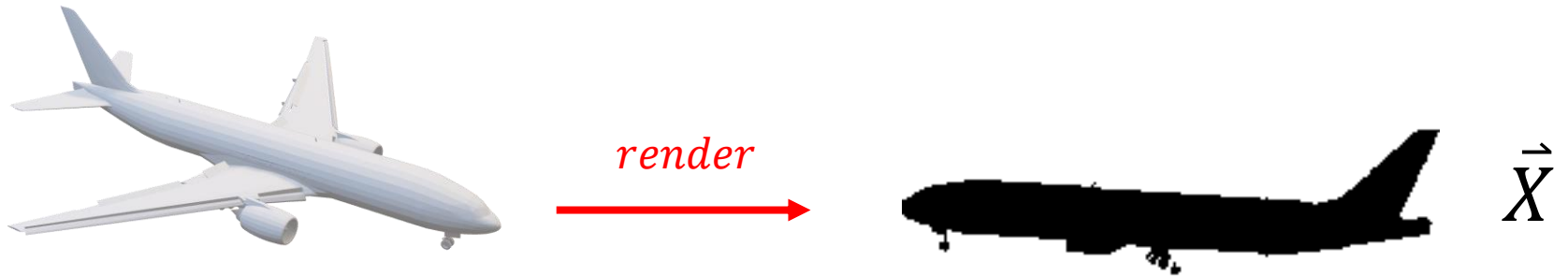
Problem 3 : Encoding

PCA don't use pixel neighborhood information



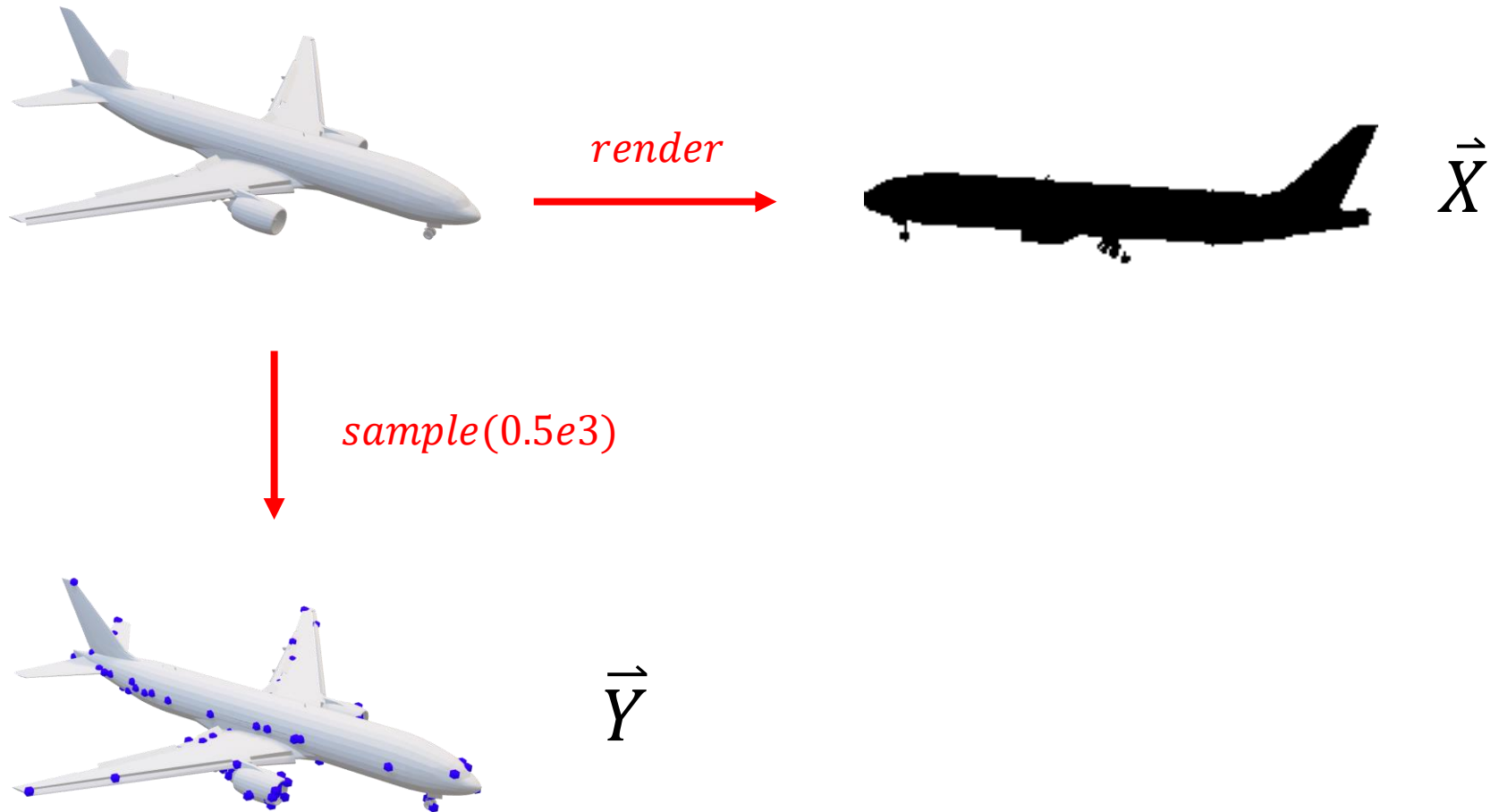
Solution 1 : ModelNet Dataset

Each object is rendered with fixed cameras



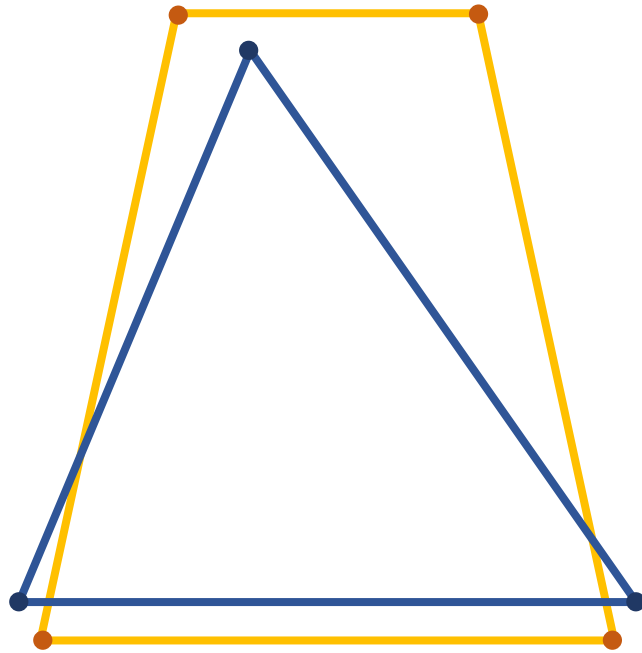
Solution 1 : ModelNet Dataset

Models are downsampled for faster training



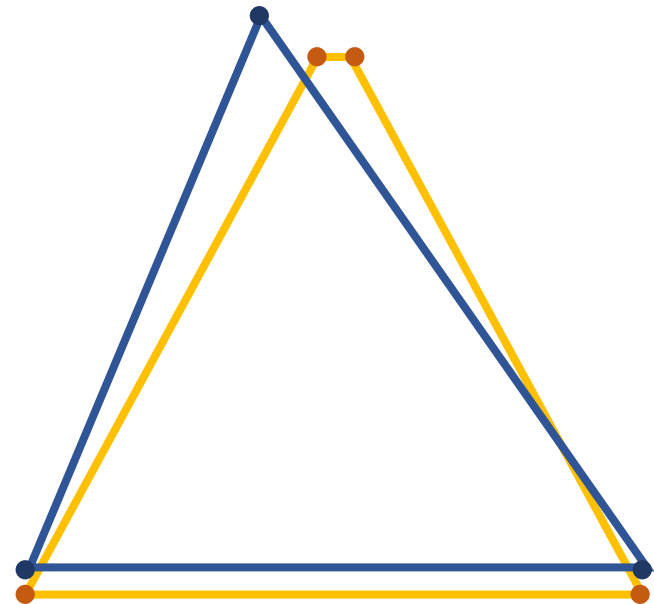
Solution 2 : Chamfer Loss

- Vertices order independent
- Compares != number of vertices



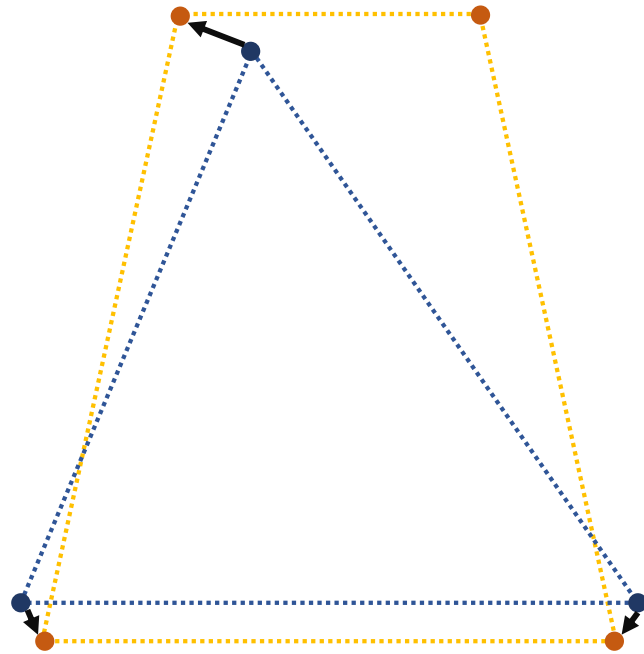
$\overrightarrow{y_{pred}}$ vs $\overrightarrow{y_{real\ 1}}$

>

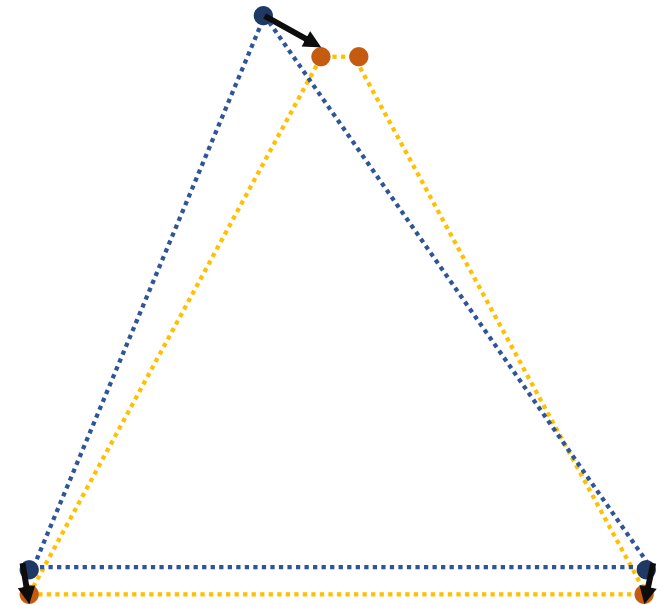


$\overrightarrow{y_{pred}}$ vs $\overrightarrow{y_{real\ 2}}$

Solution 2 : Chamfer Loss (Pred)

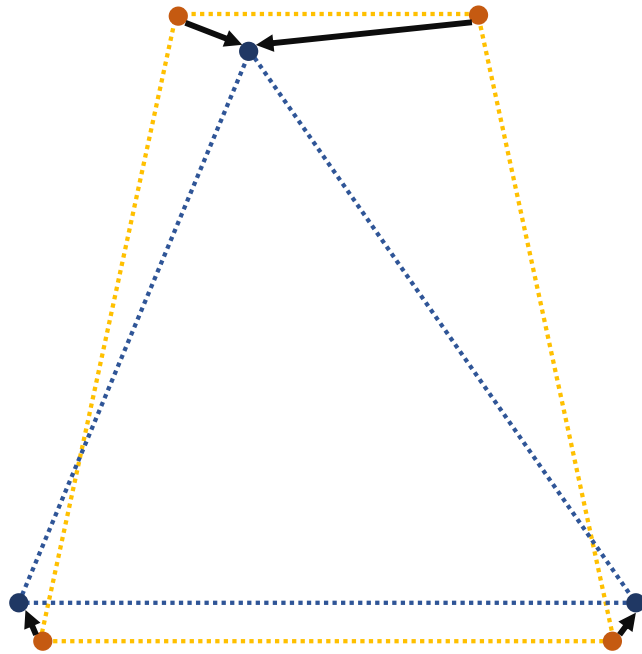


$\overrightarrow{y_{pred}}$ vs $\overrightarrow{y_{real\ 1}}$

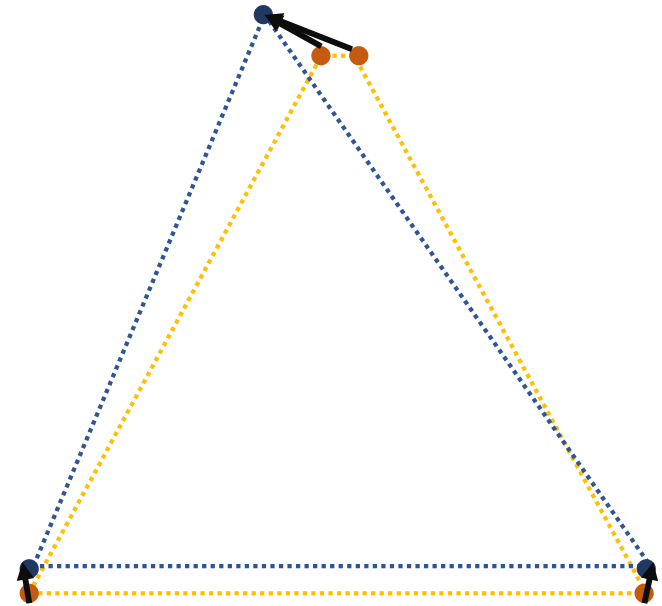


$\overrightarrow{y_{pred}}$ vs $\overrightarrow{y_{real\ 2}}$

Solution 2 : Chamfer Loss (Real)

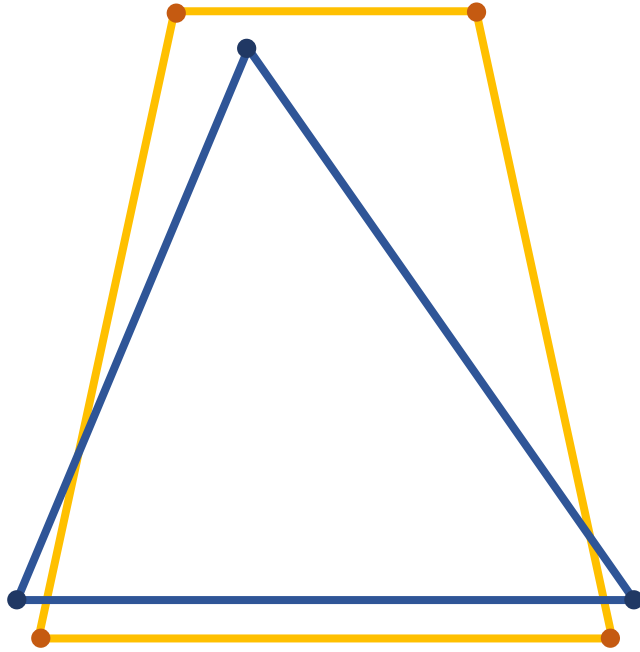


$\overrightarrow{y_{pred}}$ vs $\overrightarrow{y_{real\ 1}}$



$\overrightarrow{y_{pred}}$ vs $\overrightarrow{y_{real\ 2}}$

Solution 2 : Chamfer Loss

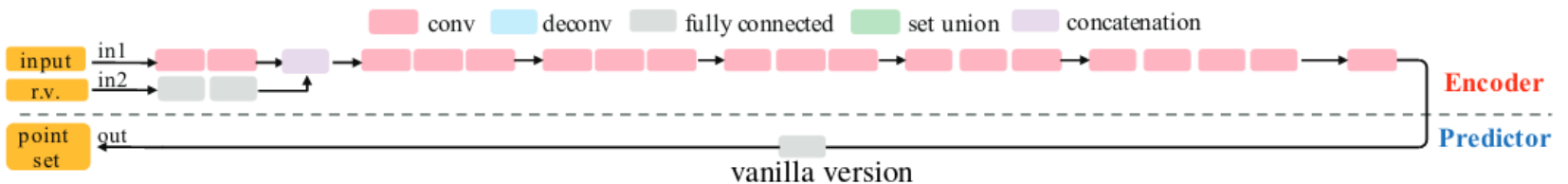
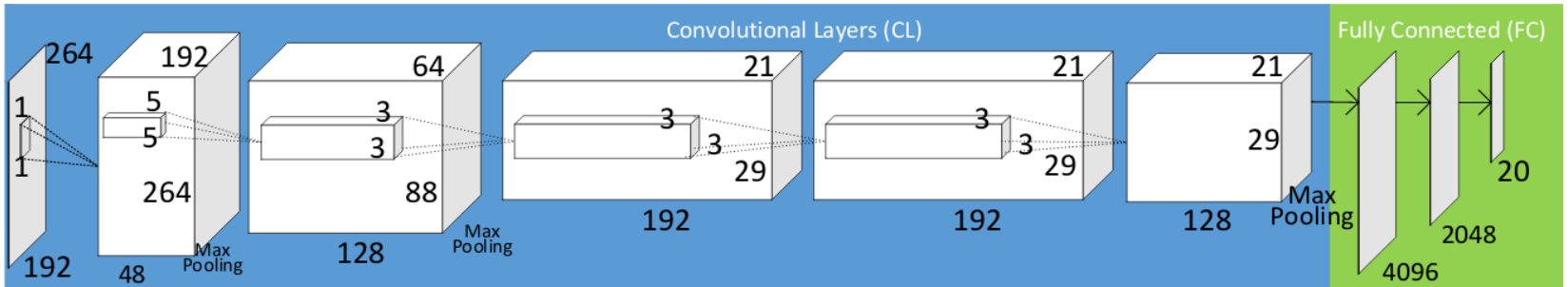
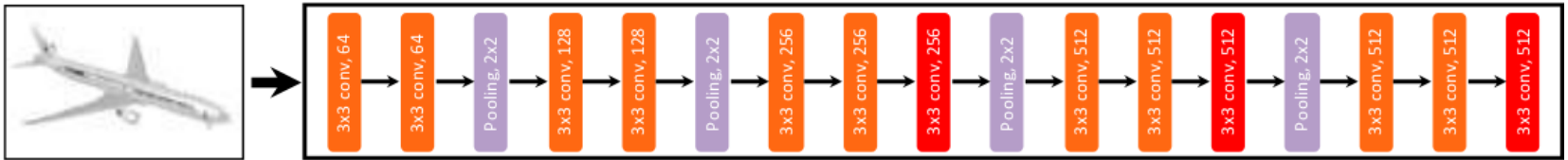


$\overrightarrow{y_{pred}}$ vs $\overrightarrow{y_{real\ 1}}$

$$\begin{aligned} loss(p, r) = & 0,5 \cdot \frac{1}{N_p} \cdot \sum_{i=1}^{N_p} \min_j |\overrightarrow{p_i} - \overrightarrow{q_j}| \\ & + 0,5 \cdot \frac{1}{N_q} \cdot \sum_{j=1}^{N_q} \min_i |\overrightarrow{p_i} - \overrightarrow{q_j}| \end{aligned}$$

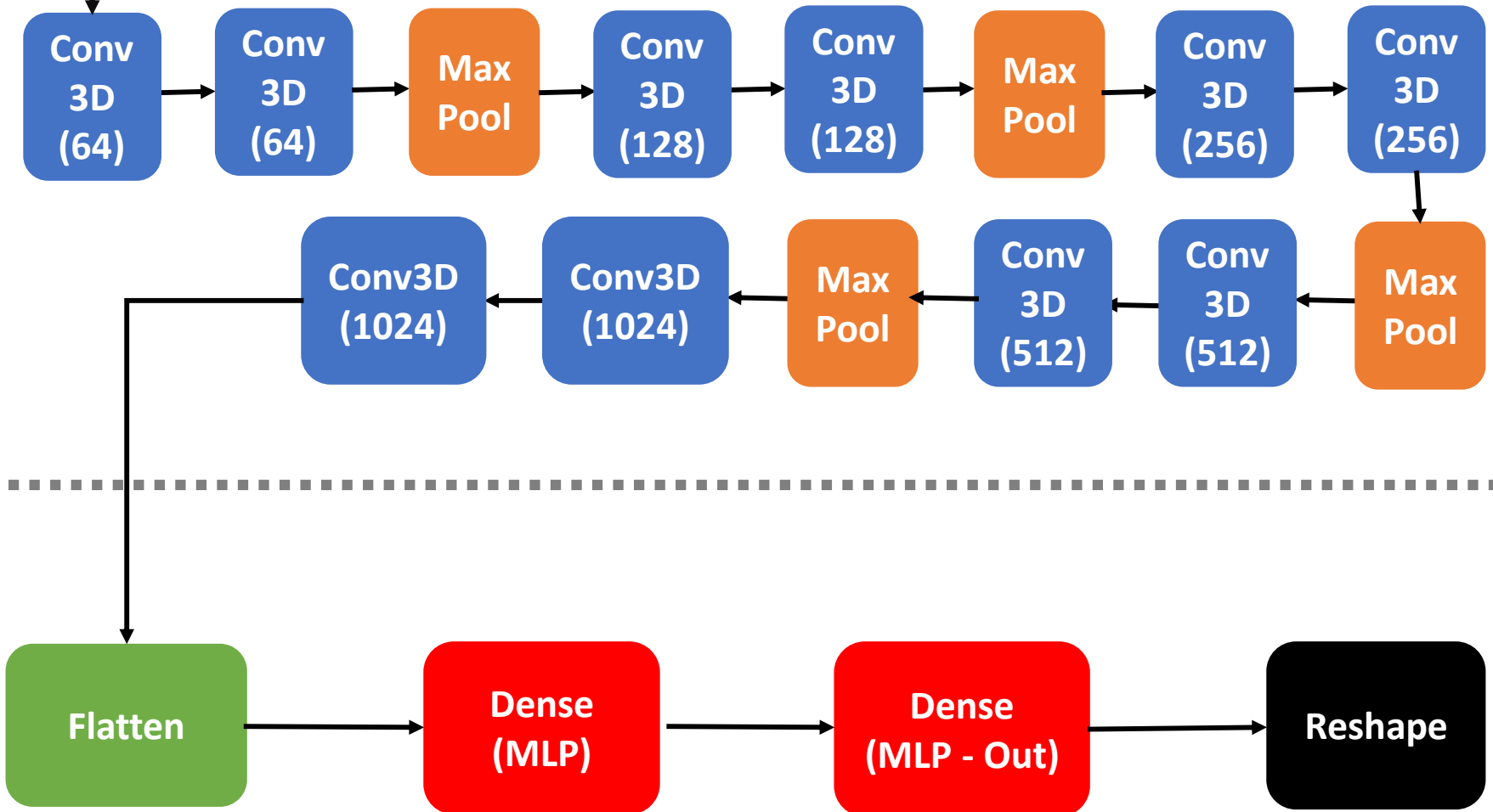
Solution 3: Deep Learning

- PCA - Linear
- Most articles - CNNs [1, 2, 3]



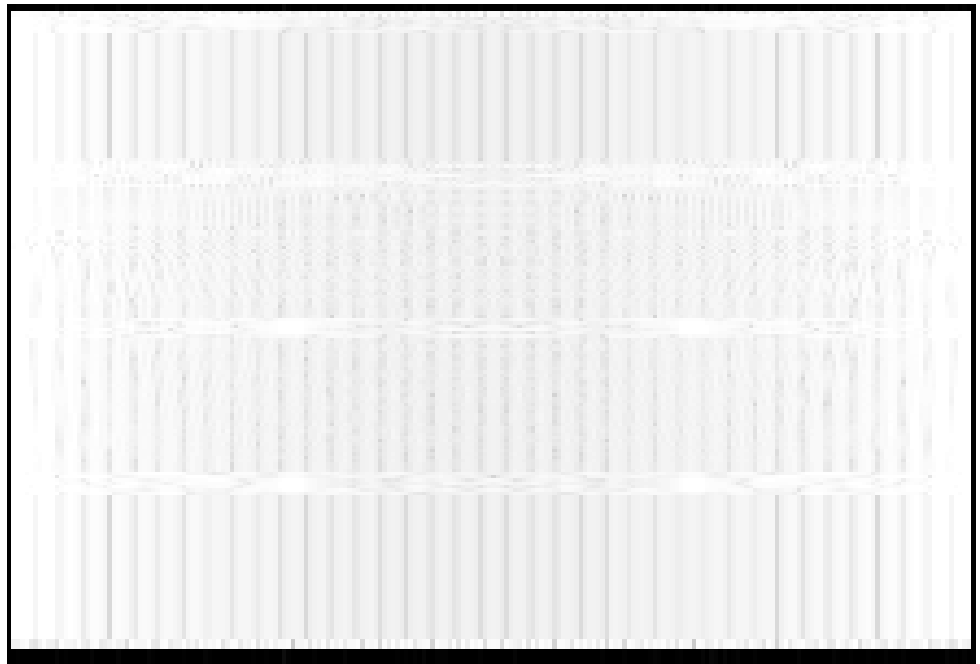
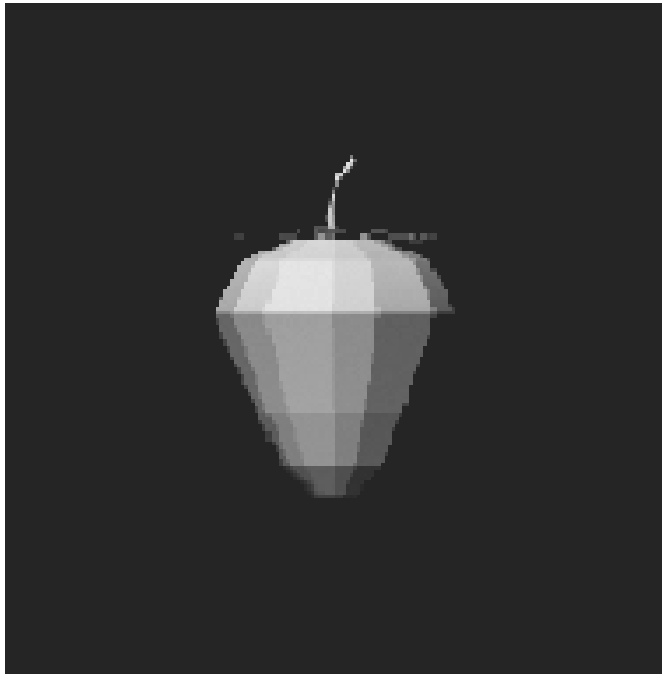
VGG Architecture (inspired)

$\vec{x}: (400, 400, 6 \text{ channels})$



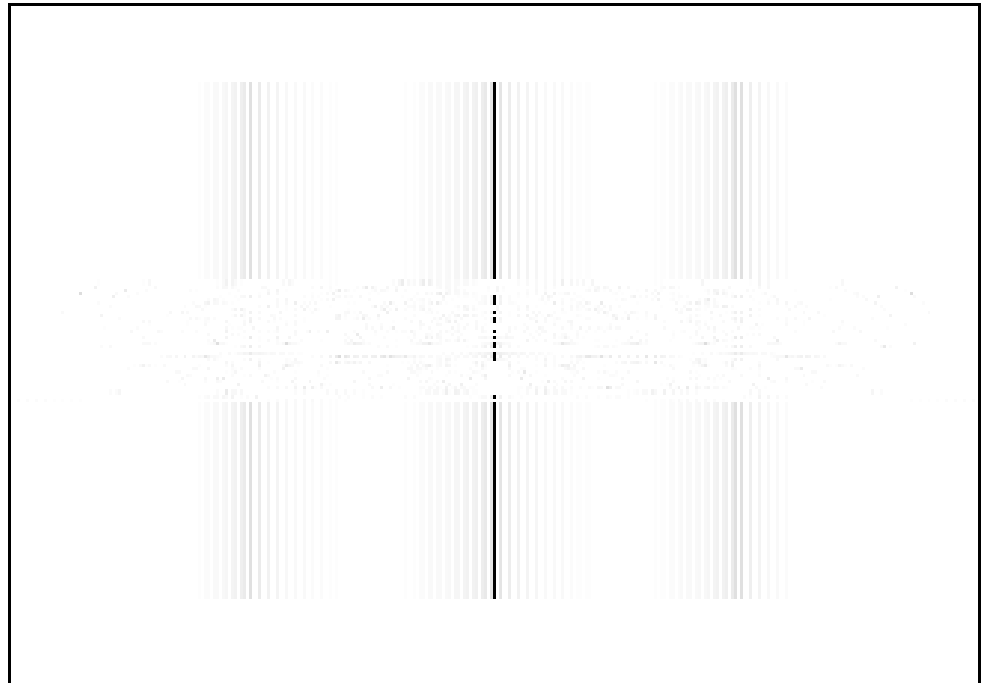
VGG Architecture : Filtering

Most articles : render with lighting => power spectrum in many frequencies

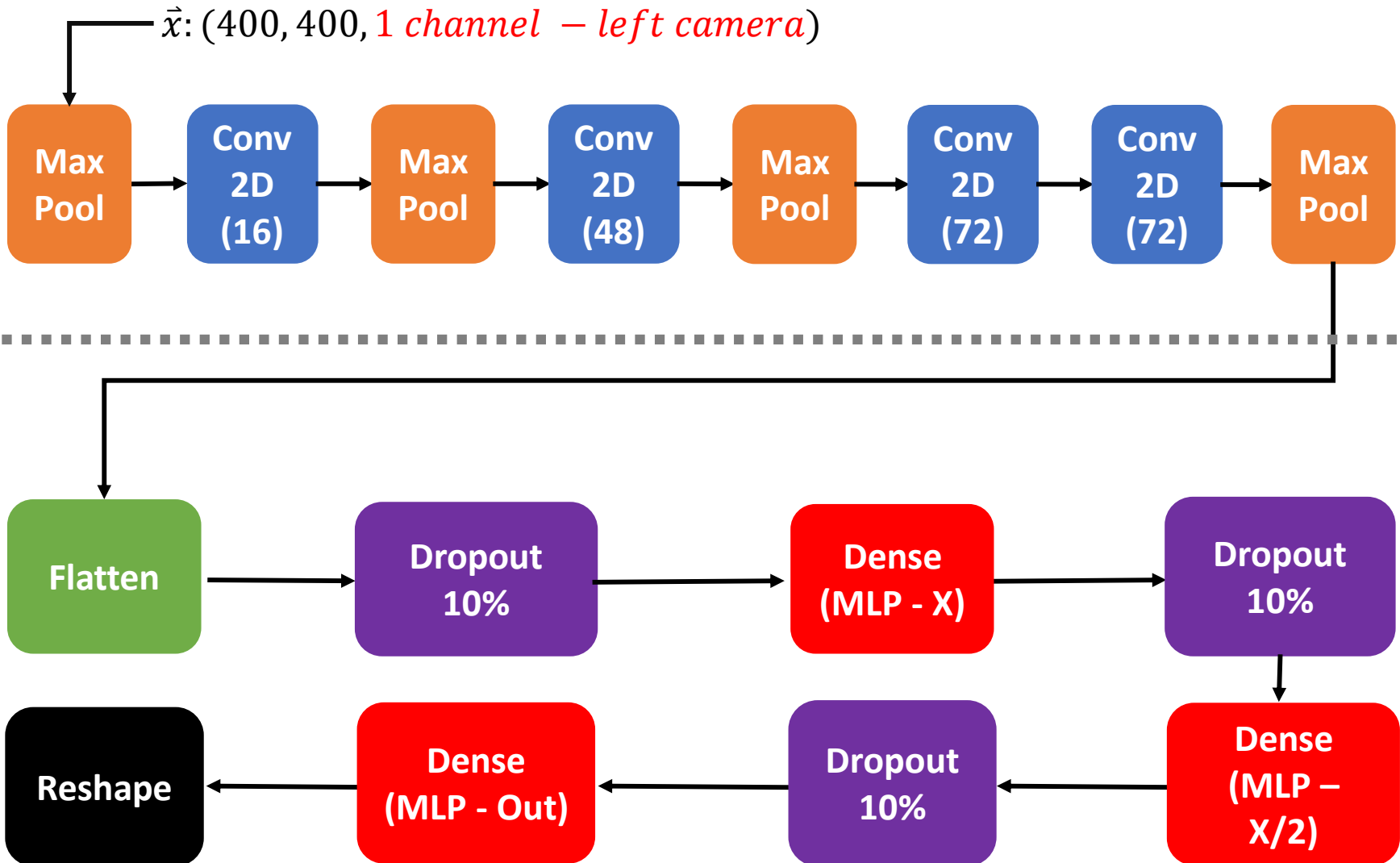


VGG Architecture : Filtering

Our solution: silhouette => high frequency information
Less CNN / MaxPool layers may give a better result



Silhouette Architecture



\vec{X}



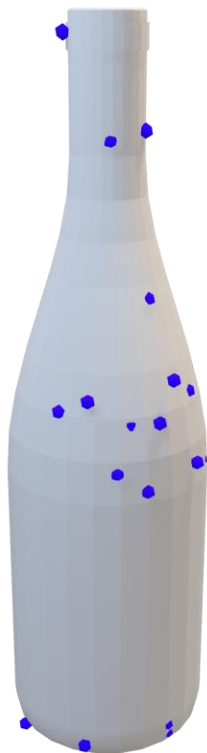
\vec{Y}



\vec{X}



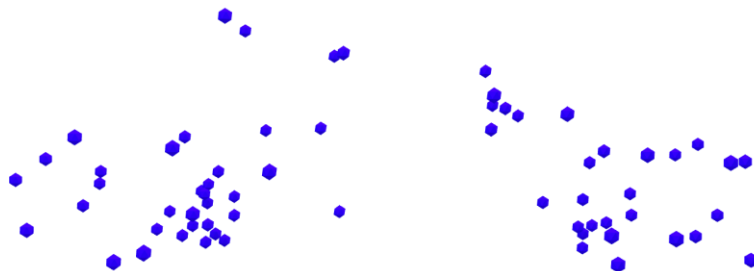
\vec{Y}



\vec{X}

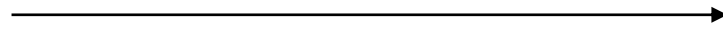


\vec{Y}

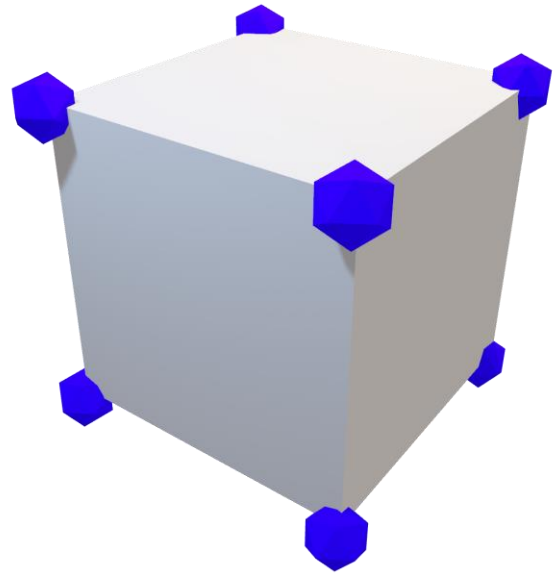
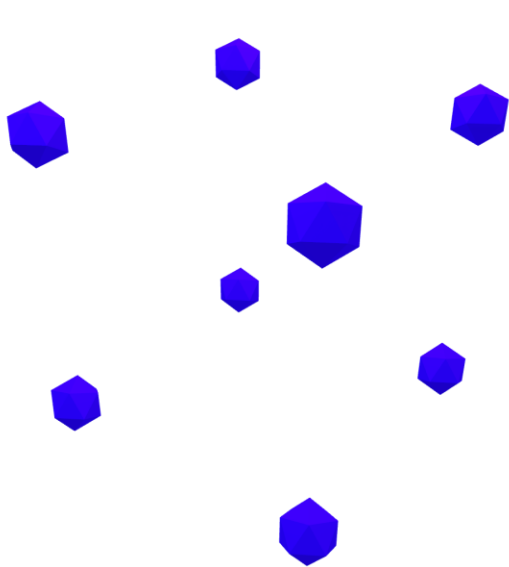


Mesh ?

$\overrightarrow{y_{pred}}: (64, 3)$



$\{vertices, faces\}$

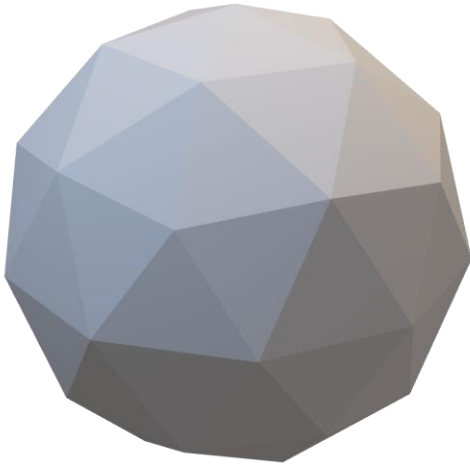


Mapping!

$\overrightarrow{y_{pred}}: (64, 3)$

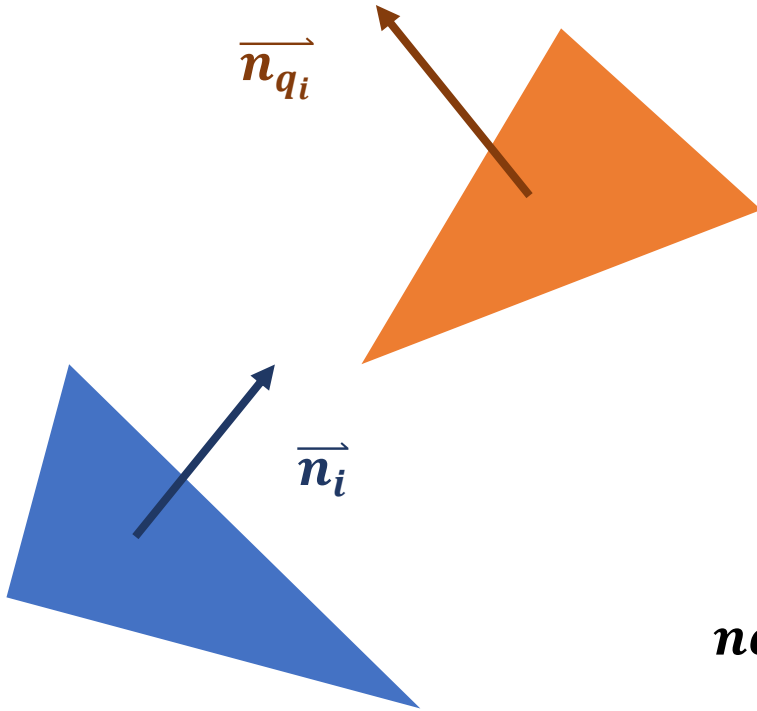


$\{vertices = y_{pred}\}$



$\{faces\}$

Normal Loss

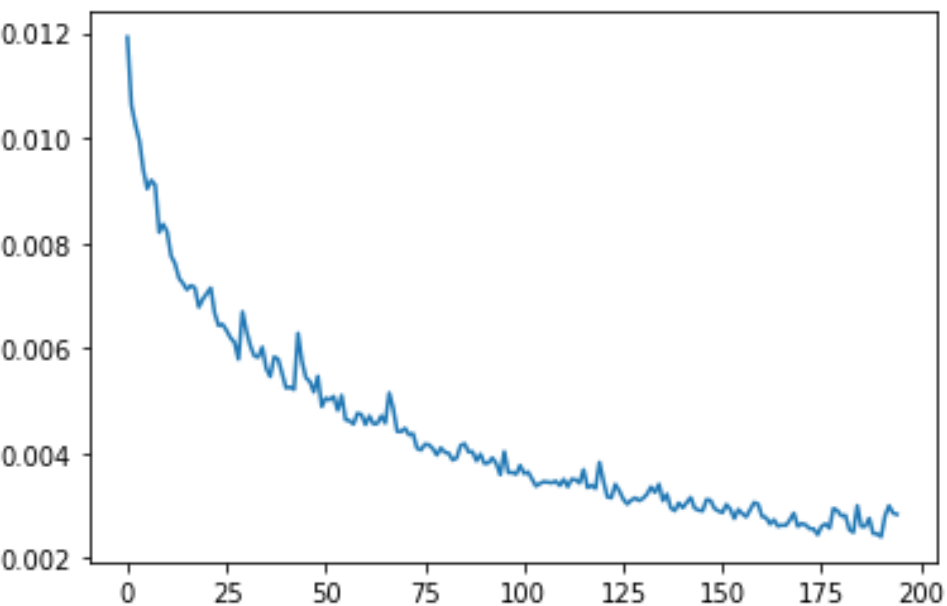


$$normal_loss(p, q) = 1 - \frac{1}{N_p} \cdot \sum_{i=1}^{N_p} \langle \overrightarrow{n_i}, \overrightarrow{n_{q_i}} \rangle$$

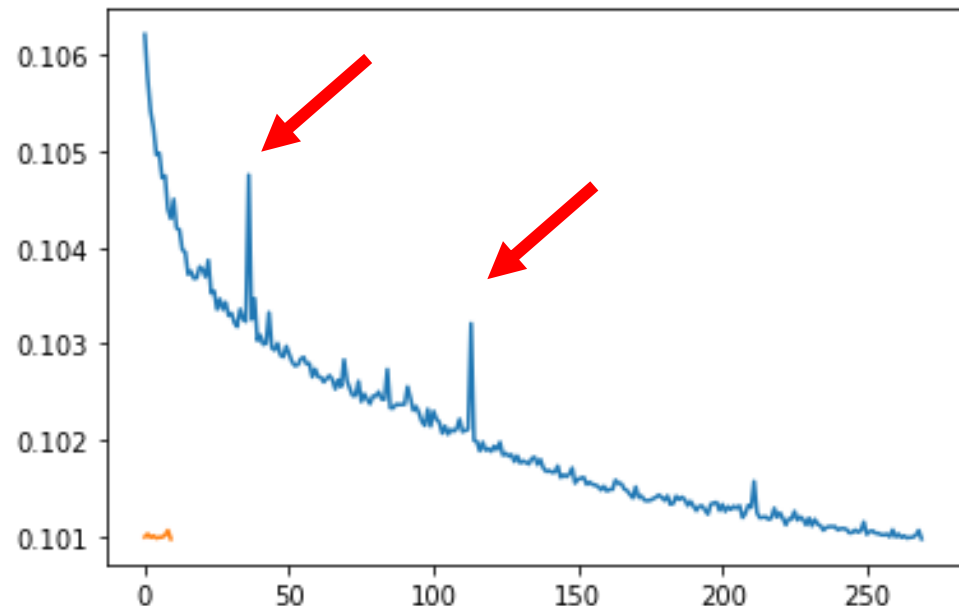
$$q_i = \arg \min_j |p_i, r_j|_2^2$$



- Normal loss as defined may give a non-convex loss



chamfer



chamfer + normal

Questions ?

Bibliography

- [1] Wang, N., Zhang, Y., Li, Z., Fu, Y., Liu, W. and Jiang, Y.G., 2018. Pixel2mesh: Generating 3d mesh models from single rgb images. In *Proceedings of the European Conference on Computer Vision (ECCV)* (pp. 52-67).
- [2] Dibra, E., Jain, H., Öztireli, C., Ziegler, R. and Gross, M., 2016, October. Hs-nets: Estimating human body shape from silhouettes with convolutional neural networks. In *2016 Fourth International Conference on 3D Vision (3DV)* (pp. 108-117). IEEE.
- [3] Fan, H., Su, H. and Guibas, L.J., 2017. A point set generation network for 3d object reconstruction from a single image. In *Proceedings of the IEEE conference on computer vision and pattern recognition* (pp. 605-613).