

2. Prove that for X, Y : $W^* = \arg \min_{W \in O_d(\mathbb{R})} \|WX - Y\|_F = UV^T$ with $U, Z, V^T = \text{SVD}(YX^T)$

Minimizing $\|WX - Y\|_F \Leftrightarrow$ minimizing $\|WX - Y\|_F^2$

$$\text{Let } X, Y \in \mathbb{R}^{d \times n}: \min_{W \in O_d(\mathbb{R})} \|WX - Y\|^2 = \min_{W \in O_d(\mathbb{R})} (WX^T)WX - 2(WX^T)Y + Y^T Y$$

$$= \min_{W \in O_d(\mathbb{R})} X^T X + Y^T Y - 2YX^T W^T$$

$$\text{because } W^T W = I$$

$$= \min_{W \in O_d(\mathbb{R})} -2YX^T W^T$$

$$= \max_{W \in O_d(\mathbb{R})} YX^T W^T$$

$$= \max_{W \in O_d(\mathbb{R})} \sum V^T W^T U$$

$$V^T, W^T, U \in O_d(\mathbb{R})^3 \Rightarrow V^T W^T U \in O_d(\mathbb{R})$$

$$\text{we should then choose } V^T W^T U = I_{d \times d} \Rightarrow W = UV^T$$

4. Which loss did you use?

I used a categorical cross-entropy loss function because we have a multiclass classification problem

$$H(p, q) = - \sum_{i=1}^S y_i \log(y_i')$$

with y_i true probability distribution
and y_i' predicted one

Be creative, use another encoder

I used XGBoost classifier, however it did not seem to add much value when submitting to Kaggle

3. Training and dev error

Training set accuracy: 0.48

Dev set accuracy: 0.42