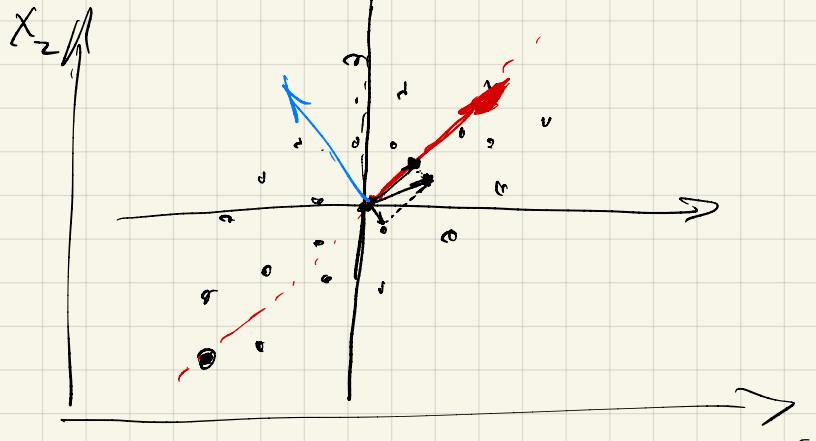


PCA = principal components analysis

$$\underline{X} \in \mathbb{R}^D$$

$$\underline{X} \rightarrow SVD$$

Singular values decomposition



$$\underline{X}_c = \underline{U} \Sigma \underline{V}^\top$$

$$\underline{T} = \underline{U} \Sigma$$

$$\underline{\tilde{X}} = \begin{bmatrix} | & | & | & | \\ X_1 & X_2 & X_3 & \dots & X_N \\ | & | & | & | \end{bmatrix} = (SVD) = \begin{bmatrix} | & | & | \\ U_1 & U_2 & \dots & U_N \\ | & | & | \end{bmatrix} \begin{bmatrix} \Sigma \\ | \\ | \end{bmatrix} \begin{bmatrix} | & | \\ V_1 & V_2 & \dots & V_N \\ | & | \end{bmatrix}^\top$$

$$\underline{U} \underline{U}^\top = \underline{U}^\top \underline{U} = \underline{I} = \underline{V} \underline{V}^\top = \underline{V}^\top \underline{V}$$

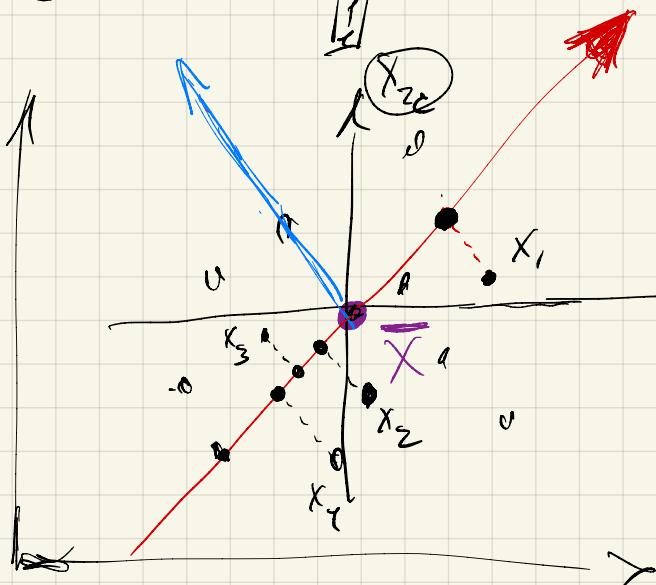
$$\bar{X} = \frac{1}{N} \sum_{i=1}^N x_i$$

$$\bar{X} \quad (1 \times f)$$

$$x_c = \bar{X} - \bar{\bar{X}}$$

$$\begin{bmatrix} & \\ & \\ & \\ & \\ & \end{bmatrix} \bar{X}$$

$$(N \times f)$$



$$x_c = \mathcal{T} V^\top$$

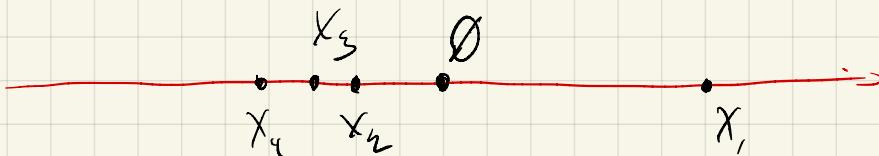
$$\mathcal{T} = U \Sigma = X_c V$$

$$V^\top \quad (k \times f)$$

$$X_c \quad (N \times f)$$

$$\tilde{x}_c = (\tilde{\mathcal{T}} \tilde{V}) \tilde{V}^\top \neq x_c$$

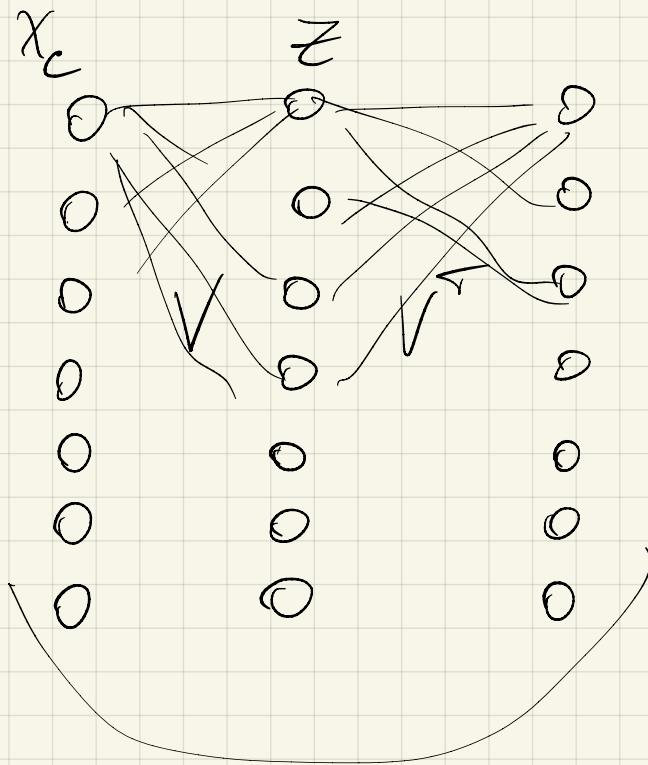
$$\mathcal{T} \quad (N \times k)$$



$$\tilde{x}_c = x_c V^* V^\top$$

$$z = x_c V$$

$$\tilde{x}_c = \tilde{z} \tilde{V}^\top$$



$$h = X_c V \quad \Theta X_c$$

$$Y_2 = h V^T$$

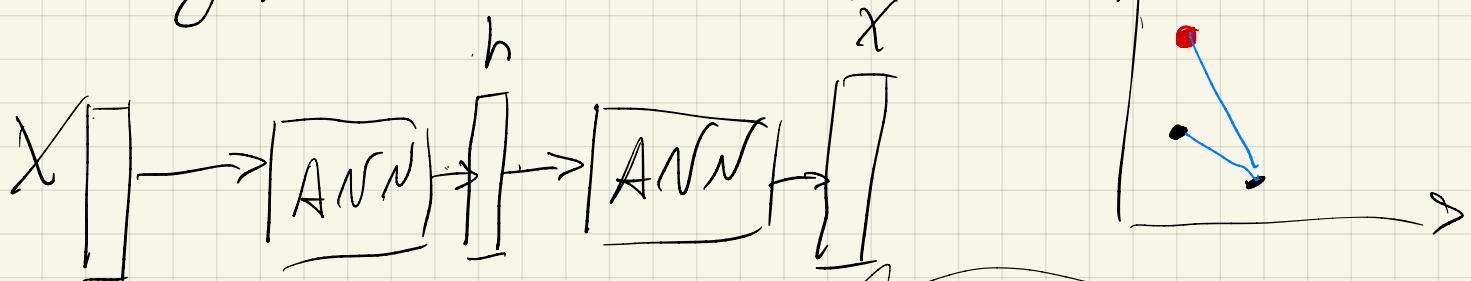
$$V V^T = I$$

$$X_c = X - \bar{X}$$

$$Z \in \mathbb{R}^k \quad k < f$$

$$\begin{matrix} V & (f \times k) \\ V^T & (k \times f) \end{matrix}$$

Abstraktionspolynomischer Autoencoder



$$X \in \mathbb{R}^{784}$$

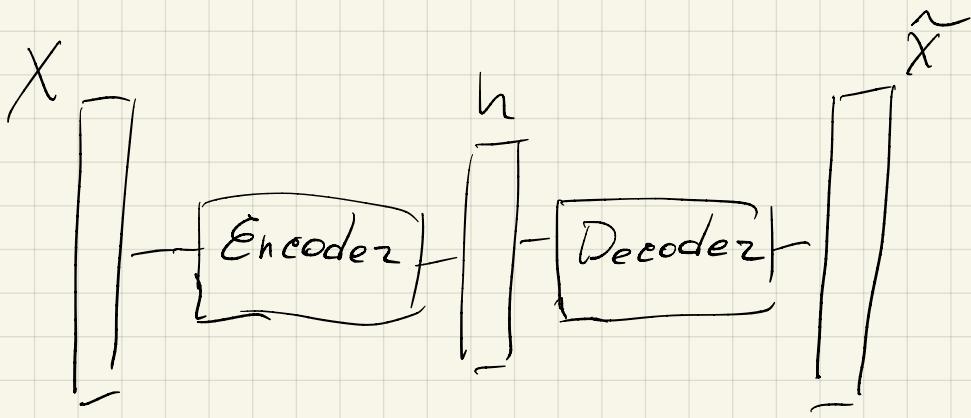
$$\begin{matrix} X_1 \\ X_2 \end{matrix}$$

$$h \in \mathbb{R}^H$$

$h$ : hidden representation  
embedding  $D=2$

$$D=10000$$

$$\sqrt{\sum_{i=1}^D (X_1^{(i)} - X_2^{(i)})^2}$$



$$E(\cdot) : \mathbb{X} \rightarrow \mathbb{H}$$

$$\mathbb{H} = \mathbb{R}^h$$

$$\mathbb{X} = \mathbb{R}^D$$

$$D = 784 \\ 260000$$

$$D(\cdot) : \mathbb{H} \rightarrow \mathbb{X}$$

$$\tilde{X} = D(E(X))$$

$$\tilde{X} = X$$

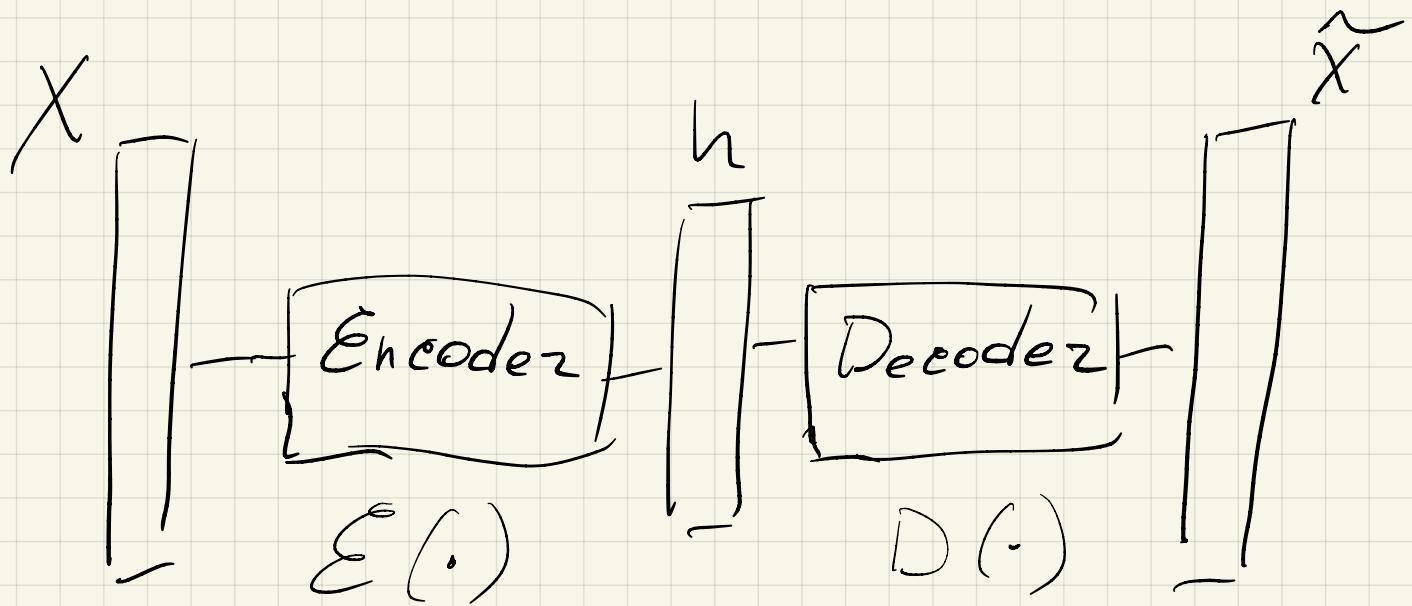
$$y = X$$

$$P = D(E(X))$$

$$J = -\frac{1}{N} \sum_{i=1}^N \sum_{j=1}^{784} (y \log(P) + (1-y) \log(1-P))$$

$$J = MAE(X, \tilde{X}) = \frac{1}{N} \sum_{i=1}^N \sum_{j=1}^D |P_{ij} - X_{ij}|$$

$$J = NSE(X, \tilde{X}) = \frac{1}{N} \sum_{i=1}^N \sum_{j=1}^D (P_{ij} - X_{ij})^2$$



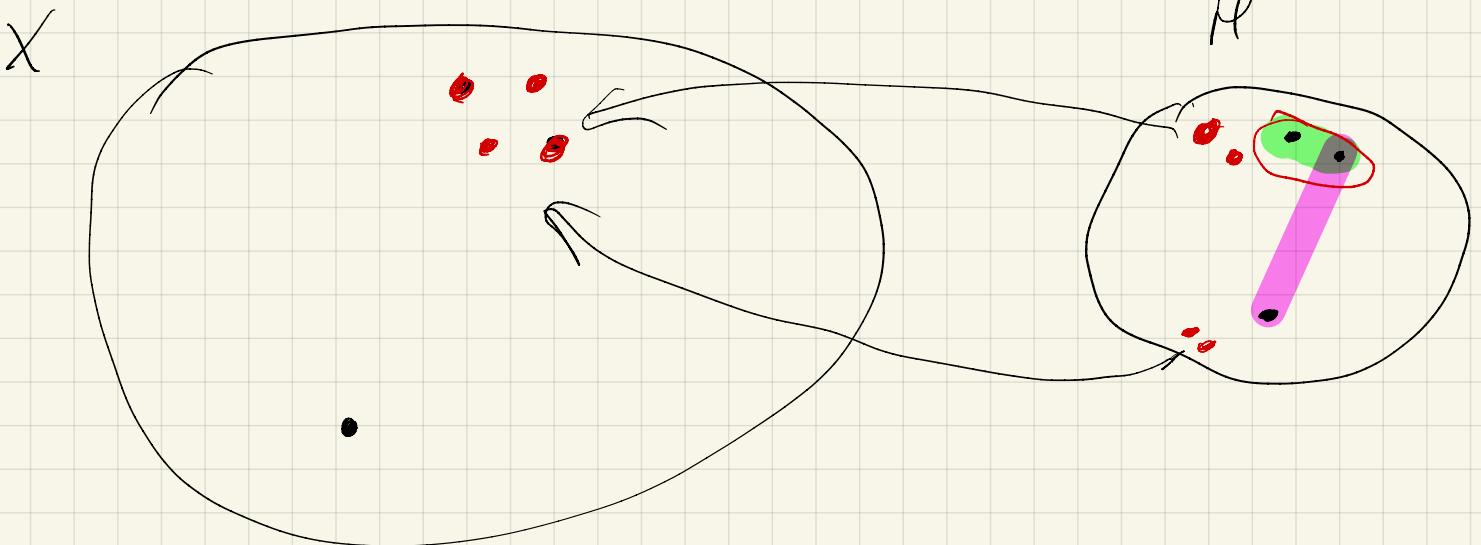
$$h = E(x)$$

$$\hat{x} = D(h)$$

$$D(E(x)) = x$$

$$h = X \theta_1$$

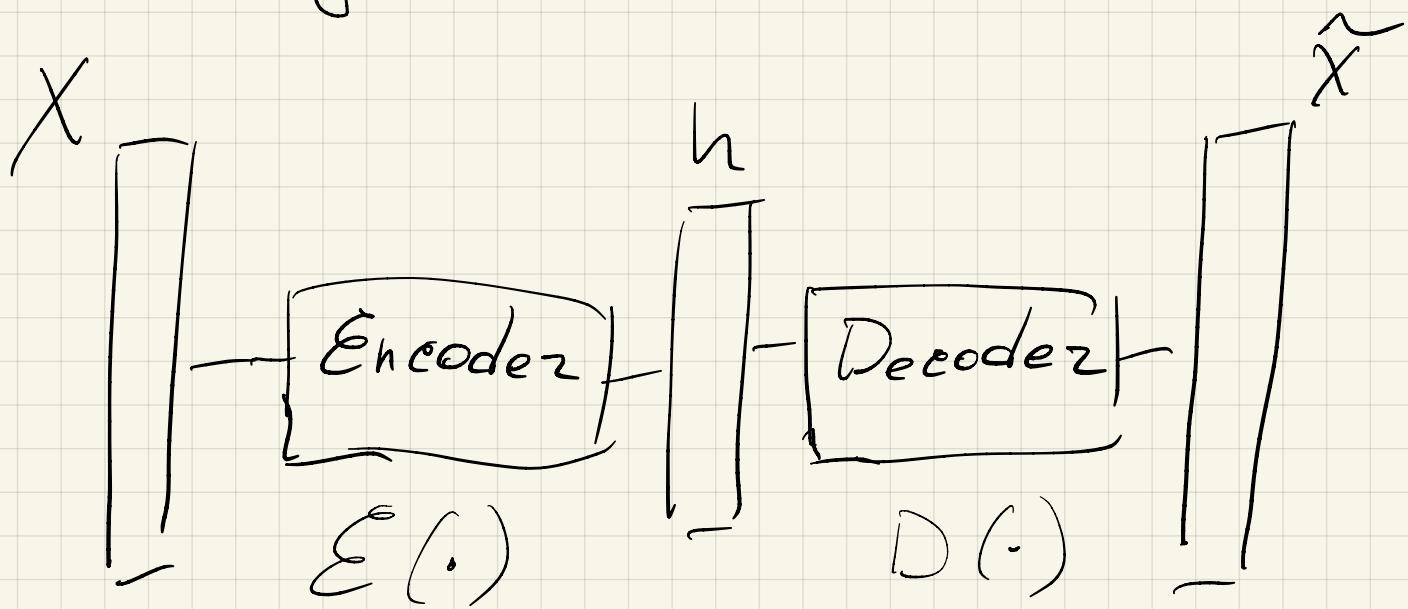
$$\hat{x} = h \theta_2$$



$$x \in \mathbb{R}^D$$

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## Denoising Autoencoder



$$x^* = x + \xi$$

$$\tilde{x} = D(E(x^*))$$

$\xi$  - noise

$$L = \text{MSE}(x, D(E(x^*)))$$