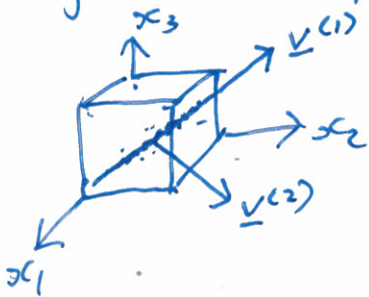


PCA

High-dim ball of points



$\underline{v}^{(k)}$ eigenvector of $\text{cov}[X]$

$$V = \begin{bmatrix} | & | & \dots & | \\ \underline{v}^{(1)} & \underline{v}^{(2)} & \dots & \underline{v}^{(k)} \\ | & | & \dots & | \end{bmatrix}$$

$D \times K$

Individual vector

$$\underline{x} \rightarrow \underbrace{V^T}_{K \times D} \underline{x}$$

$N \times D$ Design matrix:

$$X \rightarrow \underbrace{XV}_{N \times K}$$

$N \times D \quad D \times K$

If $k=2$

Lossily reconstruct

$$\hat{X} = \underbrace{XVV^T}_{N \times D}$$



All reconstructed points exactly on k -dim subspace.

Gaussian model for PCA

Assume there is a process in k -dim:

$$\underline{h}^{(n)} \sim N(0, I_k)$$

$$\underline{x}^{(n)} = \underbrace{W \underline{h}^{(n)}}_{D \times k} + \text{Gaussian noise, zero mean, covariance } \sigma_n^2 I$$

These lie exactly on k -dim linear subspace

$$\underline{x} \sim N(0, WW^T + \sigma_n^2 I)$$

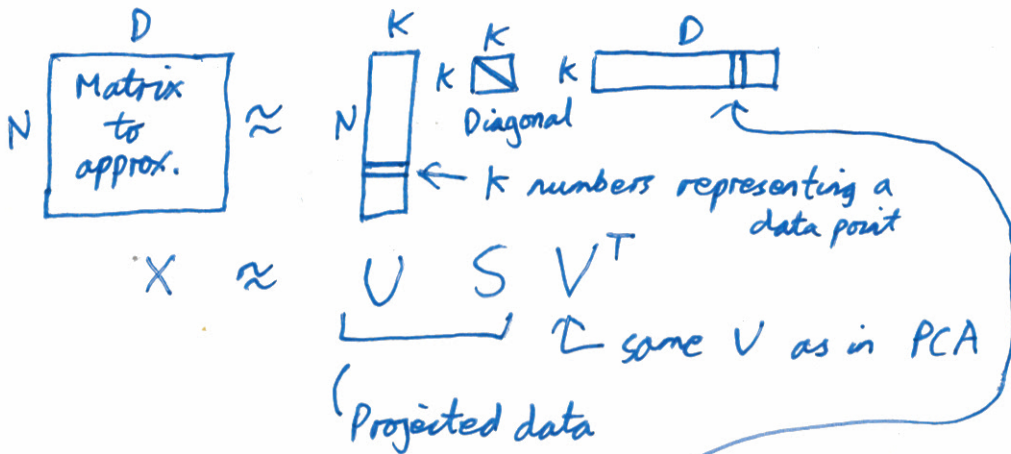
$$\left. \begin{aligned} & \text{cov}[W \underline{h}^{(n)}] \\ &= E[W \underline{h}^{(n)} (W \underline{h}^{(n)})^T] \\ &= W E[\underline{h}^{(n)} \underline{h}^{(n)T}] W^T \\ &= W I W^T \end{aligned} \right\}$$

"Probabilistic PCA"

"Factor Analysis" where noise has arbitrary diagonal covariance

Truncated SVD

SVD is some standard linear algebra method,



K numbers representing an original feature.

[See typeset notes for links to papers with figures.]

[Also for discussion of privacy.]

Truncated SVD

$$\begin{bmatrix} X_{11} & X_{12} & \cdots & X_{1D} \\ X_{21} & X_{22} & \cdots & X_{2D} \\ X_{31} & X_{32} & \cdots & X_{3D} \\ X_{41} & X_{42} & \cdots & X_{4D} \\ X_{51} & X_{52} & \cdots & X_{5D} \\ \vdots & \vdots & \ddots & \vdots \\ X_{N1} & X_{N2} & \cdots & X_{ND} \end{bmatrix} \approx$$

$$\begin{bmatrix} U_{11} & \cdots & U_{1K} \\ U_{21} & \cdots & U_{2K} \\ U_{31} & \cdots & U_{3K} \\ U_{41} & \cdots & U_{4K} \\ U_{51} & \cdots & U_{5K} \\ \vdots & \ddots & \vdots \\ U_{N1} & \cdots & U_{NK} \end{bmatrix} \begin{bmatrix} S_{11} & 0 & 0 \\ 0 & \cdots & 0 \\ 0 & 0 & S_{KK} \end{bmatrix} \begin{bmatrix} V_{11} & V_{21} & \cdots & V_{D1} \\ \vdots & \vdots & \ddots & \vdots \\ V_{1K} & V_{2K} & \cdots & V_{DK} \end{bmatrix}$$

$$X \approx U S V^T$$

```
% PCA via SVD,  
% for zero-mean X:  
[U, S, V] = svd(X, 0);  
U = U(:, 1:K);  
S = S(1:K, 1:K);  
V = V(:, 1:K);  
X_kdim = U*S;  
X_proj = U*S*V';
```