

Kernel PCA

use only K

$K \rightarrow$ Symmetric
 \rightarrow Positive Semi-Definite

We do want to find \vec{u}_1 : direction in feature space
that maximizes Variance
minimizes MSE

We can still project onto \vec{u}_1

$$\vec{u}_1 = \sum_{i=1}^n c_i \phi(x_i)$$

$\phi(x_j)$ project onto \vec{u}_1 :

$$\underbrace{\left(\frac{\phi(x_j)^T u_1}{u_1^T u_1} \right)}_{\text{Value}} u_1$$

$u_1^T u_1 = 1$

$$\begin{aligned}
 a_j &= \phi(x_j)^T u_1 \\
 &= \phi(x_j)^T \left(\sum_{i=1}^n c_i \phi(x_i) \right) \\
 &= \sum c_i \phi(x_j)^T \phi(x_i)
 \end{aligned}$$

Value
 a_j

$$u_1^T u_1 = 1$$

$$a_j = \sum_{i=1}^n c_i \underline{k(x_j, x_i)}$$

← we can use only K .

① Create K

① Center K

② $(C, \Lambda) = \text{eigen}(K)$

$$C = \left(\begin{array}{c|c|c|c} \vec{c}_1 & \vec{c}_2 & \dots & \vec{c}_n \\ \hline 1 & 1 & \dots & 1 \end{array} \right)$$

all eigenvectors

$$\Lambda = \begin{pmatrix} n\lambda_1 & & \\ & n\lambda_2 & \\ & & \dots \\ & & & n\lambda_d \end{pmatrix}$$

③ project: $\begin{pmatrix} \vec{c}_1 & \vec{c}_2 \\ 1 & 1 \end{pmatrix}$

new data in 2D: $\phi(x_i)$

$$n \times 2 \quad \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \\ \vdots & \vdots \end{pmatrix}$$

a_{11} is projection of $\phi(x_i)$ onto \vec{c}_1

a_{12} is projection of $\phi(x_i)$ onto \vec{c}_2

A_2

$$K \vec{c} = (n \cdot \lambda) \vec{c}$$

$$\vec{c} = \begin{pmatrix} c_1 \\ c_2 \\ \vdots \\ c_n \end{pmatrix} \in \mathbb{R}^n$$

$$u_1 = \sum_{i=1}^n c_i \phi(x_i)$$

$$K \vec{c}_1 = n_1 \vec{c}_1$$

$$n_1 = n \cdot \lambda_1$$

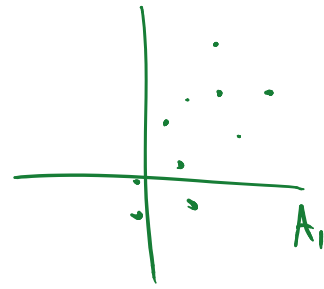
a_{11} is projection of $\phi(x_1)$ onto \vec{u}_2

$$\phi(x_1)^T \left(\sum_{i=1}^n c_i \phi(x_i) \right)$$

$$a_{11} = \sum_{i=1}^n c_i \cdot k(x_1, x_i)$$

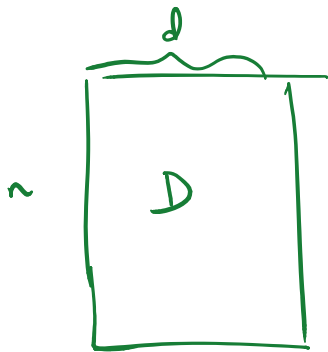
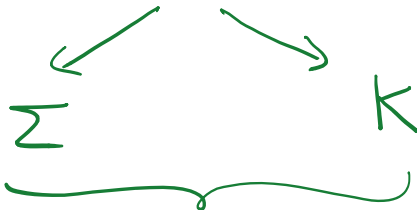
$$= \vec{c}_1 \cdot K_1$$

← 1st row of K matrix



linear kernel

exponential k



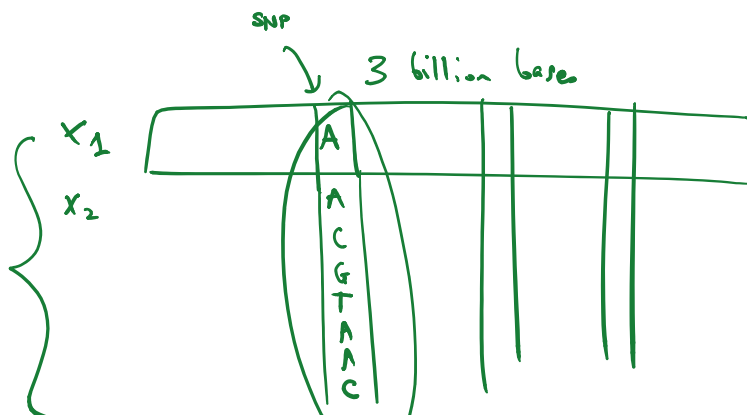
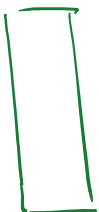
d can very large

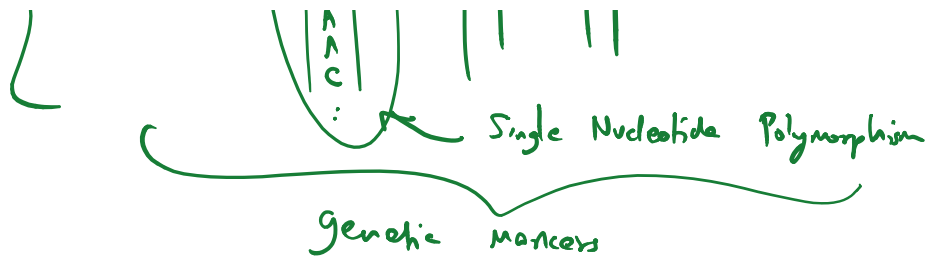


$n = 100$

$d = 100,000$

$n \gg d$

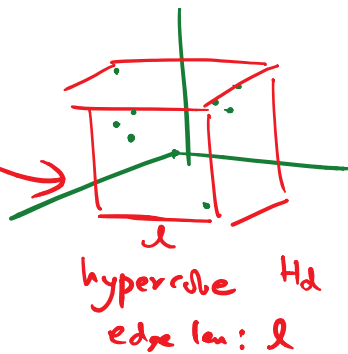




High Dimensional Space

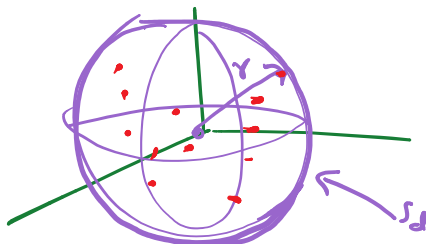
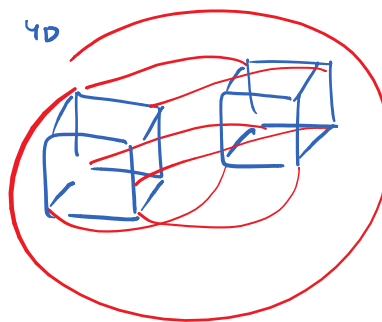
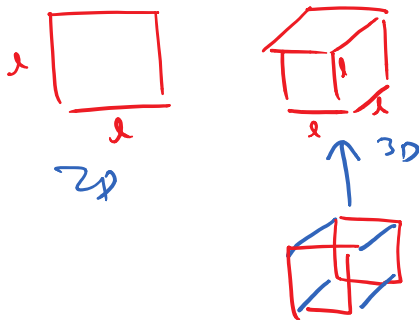
"curse" "benefits"

Universe
vs.
Sample



Volume

$$V(H_d) = l^d$$




Vol is the same

- hyper ball ← inside as well
- hyper sphere ← only the surface

$$B_d = \{ \vec{x} \mid \|\vec{x}\|_2^2 \leq r^2 \}$$

$$S_d = \{ \vec{x} \mid \|\vec{x}\|_2^2 = r^2 \}$$

the same $\left[\begin{array}{l} \text{hyper sphere} \leftarrow \text{only the surface} \end{array} \right. \quad S_d = \{ \vec{x} \mid \|x_2\|^2 = r^2 \}$

 $\Rightarrow \text{Vol}(S_2) = \pi r^2$

 $\Rightarrow \text{Vol}(S_3) = \frac{4}{3} \pi r^3$

\vdots
 $\text{Vol}(S_d) = (K_d) r^d = \left(\frac{\pi^{d/2}}{\Gamma(\frac{d}{2} + 1)} \right) r^d$

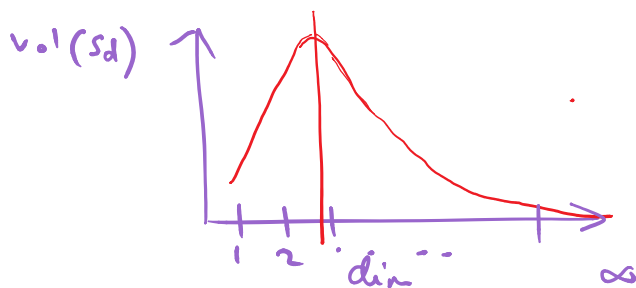
$\Gamma(\frac{d}{2} + 1) = \begin{cases} (\frac{d}{2})! & \text{if } d \text{ is even} \\ \sqrt{\pi} \left(\frac{d!!}{2^{(d+1)/2}} \right) & \text{if } d \text{ is odd} \end{cases}$ K_d

gamma function

$d!! = d(d-2)(d-4) \dots$

Vol of the unit hypersphere

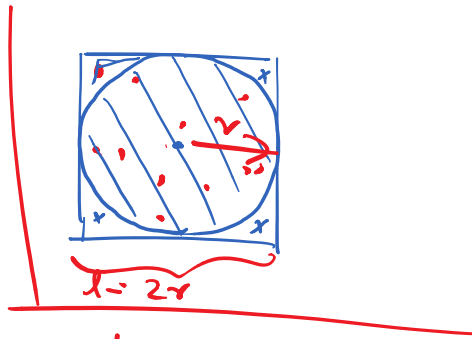
$r = 1$



$d = 5, 2, 63$

$\lim_{d \rightarrow \infty} \text{Vol}(S_d)$
 $= \lim_{d \rightarrow \infty} \frac{\pi^{d/2}}{\Gamma(\frac{d}{2} + 1)} = 0$

by about 20 dimensions
 vol is close to zero.

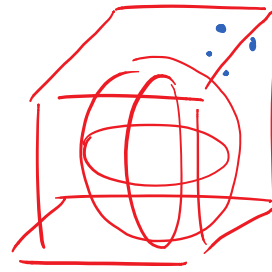


$$d=2$$

$$4 \cdot 2^d$$

$$\frac{\text{Vol}(S_2)}{\text{Vol}(H_2)} = \frac{\pi r^2}{4r^2} = \frac{\pi}{4}$$

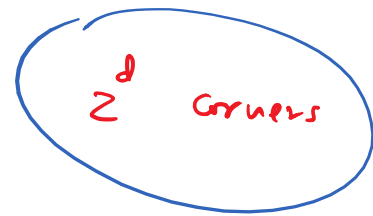
78.5%



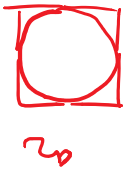
$$2^3 = 8 \text{ "corners"}$$

$$\frac{\frac{4}{3}\pi r^3}{8r^3} = 52.4\%$$

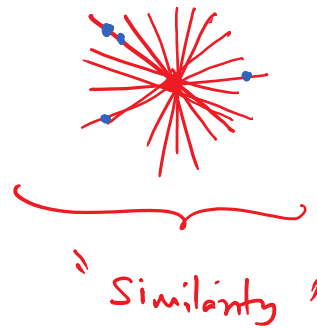
$$\lim_{d \rightarrow \infty} \frac{\text{Vol}(S_d)}{\text{Vol}(H_d)} \rightarrow 0$$



Points are in the corners
2p

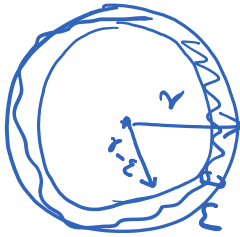


...



$$n \ll 2^d$$

Points are
On the surface



ϵ is very small $\epsilon > 0$

Vol of the thin shell

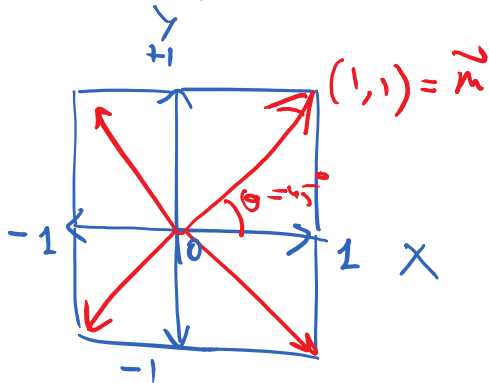
$$\frac{\text{Vol}(\text{shell}_\epsilon)}{\text{Vol}(\text{outer})} = \frac{\cancel{K_d} r^d - \cancel{K_d} (r - \epsilon)^d}{\cancel{K_d} r^d}$$

$$= 1 - \left(\frac{r - \epsilon}{r} \right)^d$$

$$\lim_{d \rightarrow \infty} \frac{\text{Vol}(\text{shell}_\epsilon)}{\text{Vol}(\text{outer})} = 1 - \underbrace{\left(1 - \frac{\epsilon}{r} \right)^d}_{0.9999 \dots 9} = 1$$

scattering effect

explosion of "new" dimension,



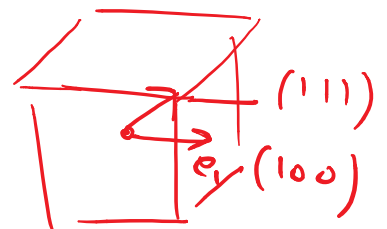
$$X = \vec{e}_1 = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \quad Y = \vec{e}_2 = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$

2D

2^d "corners" $\rightarrow 2^{d-1}$ "axes"

main diagonal: $\begin{pmatrix} 1 \\ 1 \end{pmatrix} =$

$$\cos \theta = \frac{\vec{n} \cdot \vec{e}_1}{\|\vec{n}\| \|\vec{e}_1\|}$$

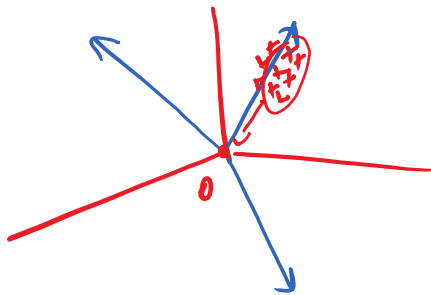


$$\lim_{A \rightarrow \infty} \cos \theta = \frac{1}{\sqrt{2}} = 0$$

$$\Rightarrow \theta = 90^\circ$$

$2^{d-1} + d$ new "orthogonal" axes

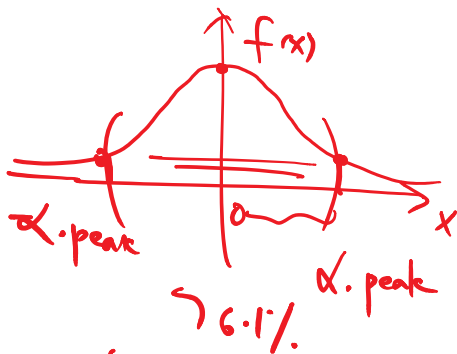
Lower-dimensional orthogonal projections



$2^{d-1} + d$
total # of "orthogonal" dimensions

Density of Normal

1D



$\alpha = 50\%$



$$f(x) = \frac{1}{(\sqrt{2\pi})^d} e^{-\frac{x^T x}{2}}$$

$\mu = 0$
 $\sigma = 1$

$$\text{Peak density} = f(0) = \frac{1}{(\sqrt{2\pi})^d}$$

$d \rightarrow \infty$

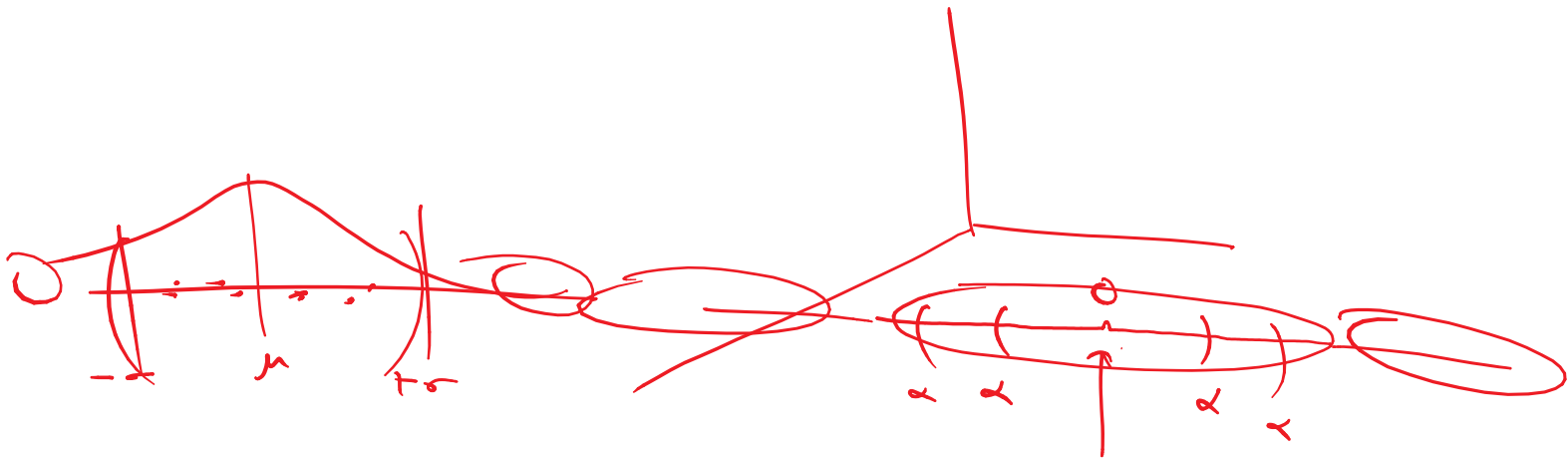
2D
50%



30%.



$d \rightarrow \infty$



most of the pub mass
lies in the tails