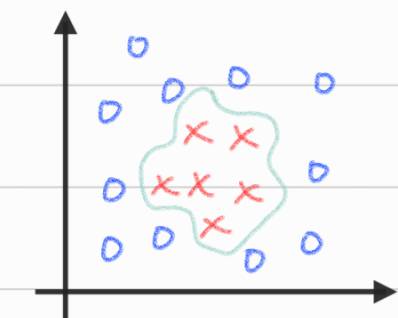


# Kernels

## Non-linear Decision Boundary



Given  $x$ :

$$f_1 = \text{similarity}(x, \underline{l^{(1)}}) = \exp\left(-\frac{\|x - l^{(1)}\|^2}{2\sigma^2}\right)$$

$$\text{If } x \approx l^{(1)}: f_1 \approx \exp\left(-\frac{0^2}{2\sigma^2}\right) \approx 1$$

$$\text{If } x \text{ is far from } l^{(1)}: f_1 = \exp\left(-\frac{(\text{large})^2}{2\sigma^2}\right) \approx 0$$

$\sigma^2$  (sigma squared) 크면 , 작으면 

Define extra features using landmarks  
and similarity functions to learn more  
complex nonlinear classifier

## SVM with Kernels

Given  $(x^{(1)}, y^{(1)}), (x^{(2)}, y^{(2)}), \dots, (x^{(m)}, y^{(m)})$

Choose  $l^{(1)} = x^{(1)}, l^{(2)} = x^{(2)}, \dots, l^{(m)} = x^{(m)}$

Given example  $x$ :

$f_1 = \text{similarity}(x, l^{(1)})$

$f_2 = \text{similarity}(x, l^{(2)})$

$$f = \begin{bmatrix} f_0 \\ f_1 \\ f_2 \\ \vdots \\ f_m \end{bmatrix} \quad f_0 = 1$$

For training example  $(x^{(i)}, y^{(i)})$ :

$$f_1^{(i)} = \text{sim}(x^{(i)}, l^{(1)})$$

$$x^{(i)} \rightarrow f_2^{(i)} = \text{sim}(x^{(i)}, l^{(2)})$$

$$\vdots \leftarrow f_i^{(i)} = \text{sim}(x^{(i)}, l^{(i)}) = \exp(-\frac{0}{2\sigma^2}) = 1$$

$$f_m^{(i)} = \text{sim}(x^{(i)}, l^{(m)})$$

Hypothesis: Given  $x$ , compute features  $f \in \mathbb{R}^{m+1}$

Predict "y=1" if  $\theta^T f \geq 0$

Training:

$$\min_{\theta} C \sum_{i=1}^m y^{(i)} \text{cost}_1(\theta^T f^{(i)}) + (1 - y^{(i)}) \text{cost}_0(\theta^T f^{(i)}) + \frac{1}{2} \sum_{j=1}^m \theta_j^2$$

$C (= \frac{1}{\lambda})$  { Large  $C$ : Lower Bias, Higher Variance (=small  $\lambda$ )  $\nearrow$   $\nearrow$   $\downarrow$   
small  $C$ : Higher Bias, Lower Variance (=large  $\lambda$ )  $\nearrow$   $\nearrow$   $\uparrow$

$\sigma^2$  { Large  $\sigma^2$ : Higher Bias, Lower Variance features vary smoothly  $\nearrow$   $\nearrow$   
small  $\sigma^2$ : Lower Bias, Higher Variance " less smoothly  $\nearrow$   $\downarrow$