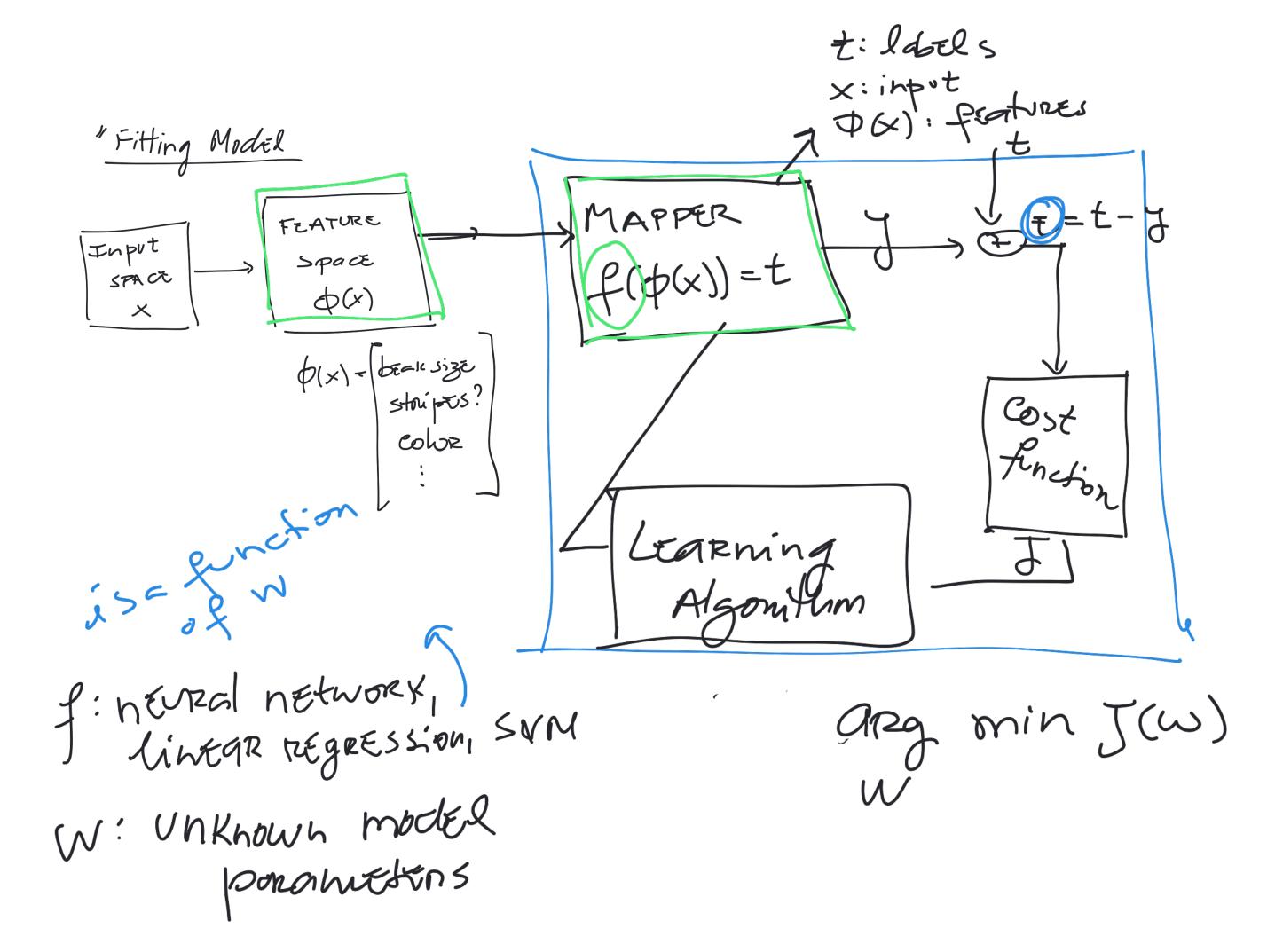
1: test point w/out label Class, fication Genorative Discriminative



Cost function

J(W)

W

arg min JCw)

The stypically an ERREDR

Function

Input data: Yolynomial Régréssion. Linear regression w/ polynomial
features y wo. x +w, x +... + wm x $y = (x^2 + x^2 +$ polynomial input fichnes M: model exder $X = \left[\begin{array}{c} \phi(x_1)^T \\ \phi(x_2)^T \\ \hline \phi(x_N)^T \end{array} \right] = \left[\begin{array}{c} x_1 & x_1 & \dots & x_1^M \\ x_2 & x_2 & \dots & x_2^M \\ \vdots & & & & \\ x_N & x_N & \dots & x_N \end{array} \right] \quad W = \left[\begin{array}{c} x_1 & x_1 & \dots & x_1^M \\ x_2 & x_2 & \dots & x_2^M \\ \vdots & & & & \\ x_N & x_N & \dots & x_N \end{array} \right] \quad W = \left[\begin{array}{c} x_1 & x_1 & \dots & x_1^M \\ x_2 & x_2 & \dots & x_2^M \\ \vdots & & & & \\ x_N & x_N & \dots & x_N \end{array} \right] \quad W = \left[\begin{array}{c} x_1 & x_1 & \dots & x_1^M \\ x_2 & x_2 & \dots & x_2^M \\ \vdots & & & & \\ x_N & x_N & \dots & x_N \end{array} \right] \quad W = \left[\begin{array}{c} x_1 & x_1 & \dots & x_1^M \\ x_2 & x_2 & \dots & x_2^M \\ \vdots & & & & \\ x_N & x_N & \dots & x_N \end{array} \right] \quad W = \left[\begin{array}{c} x_1 & x_1 & \dots & x_1^M \\ \vdots & & & & \\ x_N & x_N & \dots & x_N \end{array} \right] \quad W = \left[\begin{array}{c} x_1 & x_1 & \dots & x_1^M \\ \vdots & & & & \\ x_N & x_N & \dots & x_N \end{array} \right] \quad W = \left[\begin{array}{c} x_1 & x_1 & \dots & x_1^M \\ \vdots & & & & \\ x_N & x_N & \dots & x_N \end{array} \right] \quad W = \left[\begin{array}{c} x_1 & x_1 & \dots & x_1^M \\ \vdots & & & & \\ x_N & x_N & \dots & x_N \end{array} \right] \quad W = \left[\begin{array}{c} x_1 & x_1 & \dots & x_1^M \\ \vdots & & & & \\ x_N & x_N & \dots & x_N \end{array} \right] \quad W = \left[\begin{array}{c} x_1 & x_1 & \dots & x_1^M \\ \vdots & & & & \\ x_N & x_N & \dots & x_N \end{array} \right] \quad W = \left[\begin{array}{c} x_1 & x_1 & \dots & x_1^M \\ \vdots & & & & \\ x_N & x_N & \dots & x_N \end{array} \right] \quad W = \left[\begin{array}{c} x_1 & x_1 & \dots & x_1^M \\ \vdots & & & & \\ x_N & x_N & \dots & x_N \end{array} \right] \quad W = \left[\begin{array}{c} x_1 & x_1 & \dots & x_N \\ \vdots & & & & \\ x_N & x_N & \dots & x_N \end{array} \right] \quad W = \left[\begin{array}{c} x_1 & x_1 & \dots & x_N \\ \vdots & & & & \\ x_N & x_N & \dots & x_N \end{array} \right] \quad W = \left[\begin{array}{c} x_1 & x_1 & \dots & x_N \\ \vdots & & & & \\ x_N & x_N & \dots & x_N \end{array} \right] \quad W = \left[\begin{array}{c} x_1 & x_1 & \dots & x_N \\ \vdots & & & & \\ x_N & x_N & \dots & x_N \end{array} \right] \quad W = \left[\begin{array}{c} x_1 & x_1 & \dots & x_N \\ \vdots & & & & \\ x_N & x_N & \dots & x_N \end{array} \right] \quad W = \left[\begin{array}{c} x_1 & x_1 & \dots & x_N \\ \vdots & & & & \\ x_N & x_N & \dots & x_N \end{array} \right] \quad W = \left[\begin{array}{c} x_1 & x_1 & \dots & x_N \\ \vdots & & & & \\ x_N & x_N & \dots & x_N \end{array} \right] \quad W = \left[\begin{array}{c} x_1 & x_1 & \dots & x_N \\ \vdots & & & & \\ x_N & x_N & \dots & x_N \end{array} \right] \quad W = \left[\begin{array}{c} x_1 & x_1 & \dots & x_N \\ \vdots & & & & \\ x_N & x_N & \dots & x_N \end{array} \right] \quad W = \left[\begin{array}{c} x_1 & x_1 & \dots & x_N \\ \vdots & & & & \\ x_N & x_N & \dots & x_N \end{array} \right] \quad W = \left[\begin{array}{c} x_1 & x_1 & \dots & x_N \\ \vdots & & & & \\ x_N & x_N & \dots & x_N \end{array} \right] \quad W = \left[\begin{array}{c} x_1 & x_1 & \dots & x_N \\ \vdots & & & & \\ x_N & x_N & \dots & x_N \end{array} \right] \quad W = \left[\begin{array}{c} x_1 & x_1 &$

Y is our Estimate of labels t
$$E = t - y = \begin{bmatrix} e_1 \\ e_2 \\ \vdots \\ e_N \end{bmatrix}$$

$$J(w) = \frac{1}{2} \underbrace{\sum_{i=1}^{N} e_i^2}_{i=1}$$