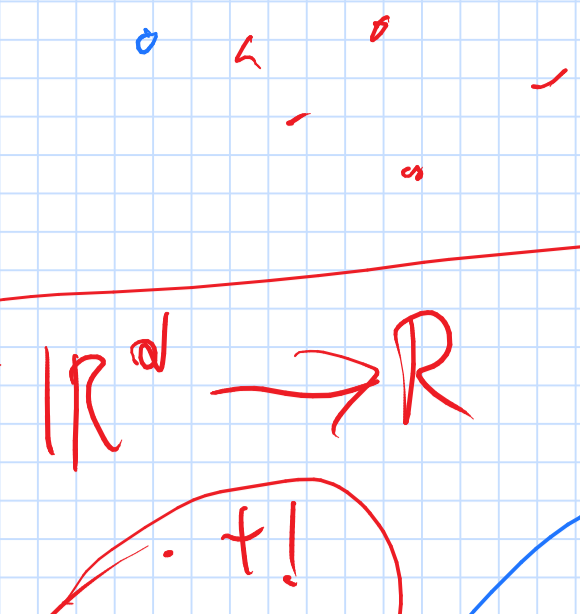
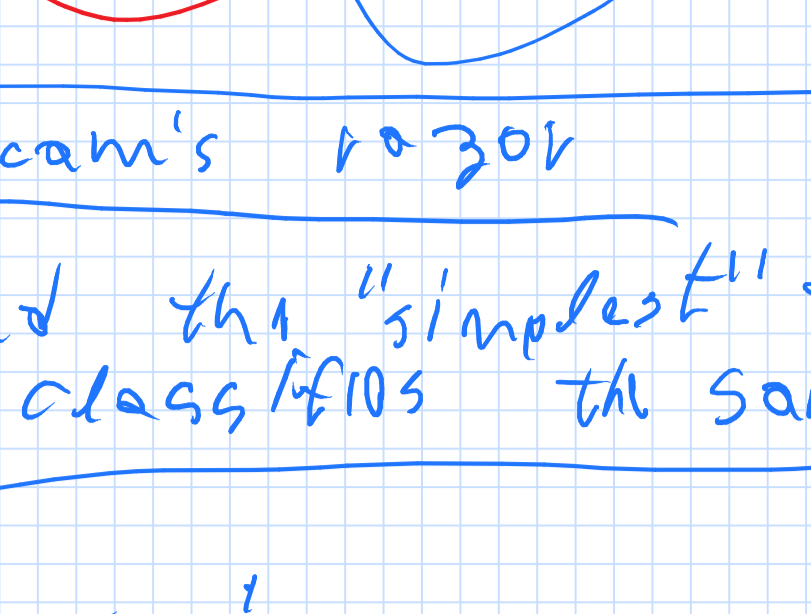


Perceptron

\mathbb{R}^d d large

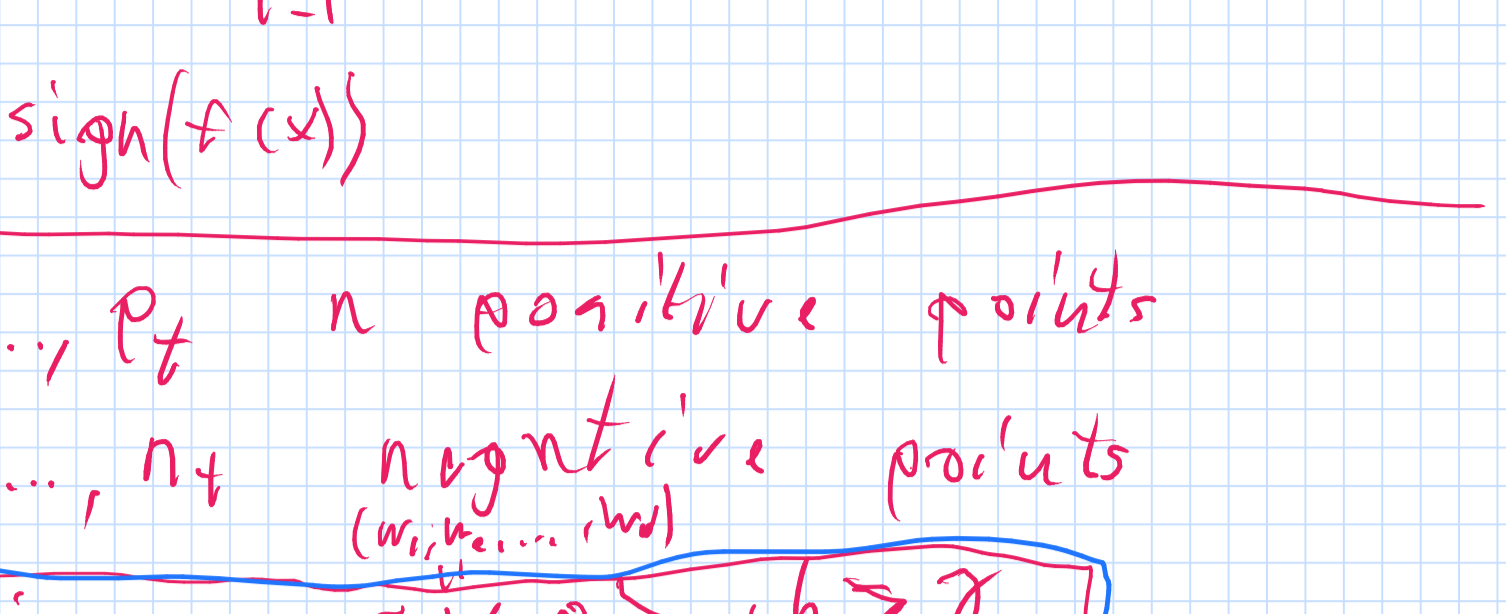


$f: \mathbb{R}^d \rightarrow \mathbb{R}$



Occam's razor

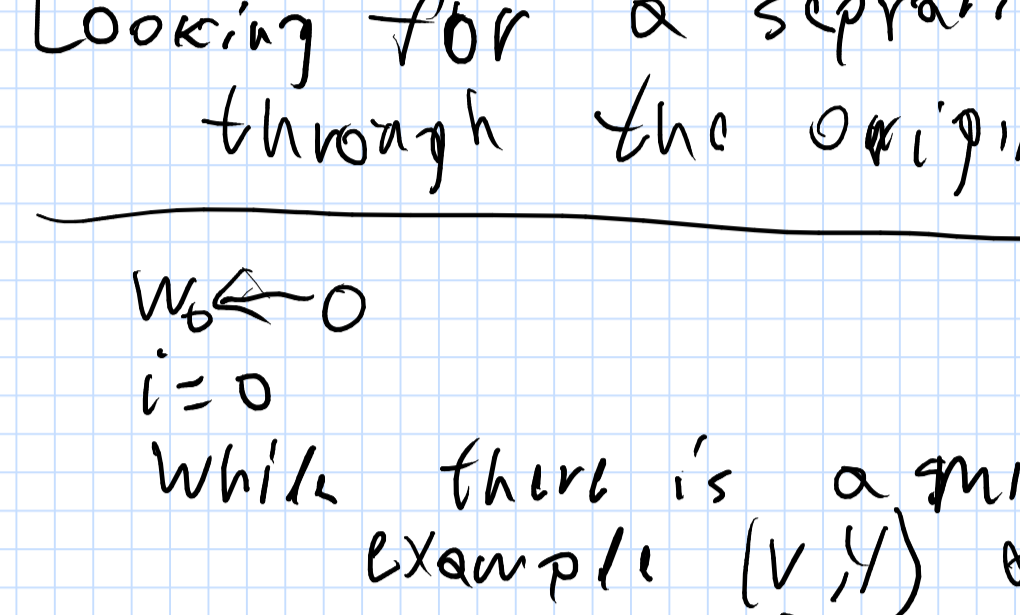
Find the "simplest" function that classifies the samples correctly.



$f(x) = \sum_{i=1}^n w_i \cdot x_i + b$
 $\text{sign}(f(x))$

p_1, \dots, p_t n positive points
 n_1, \dots, n_t negative points
(w_1, \dots, w_t)

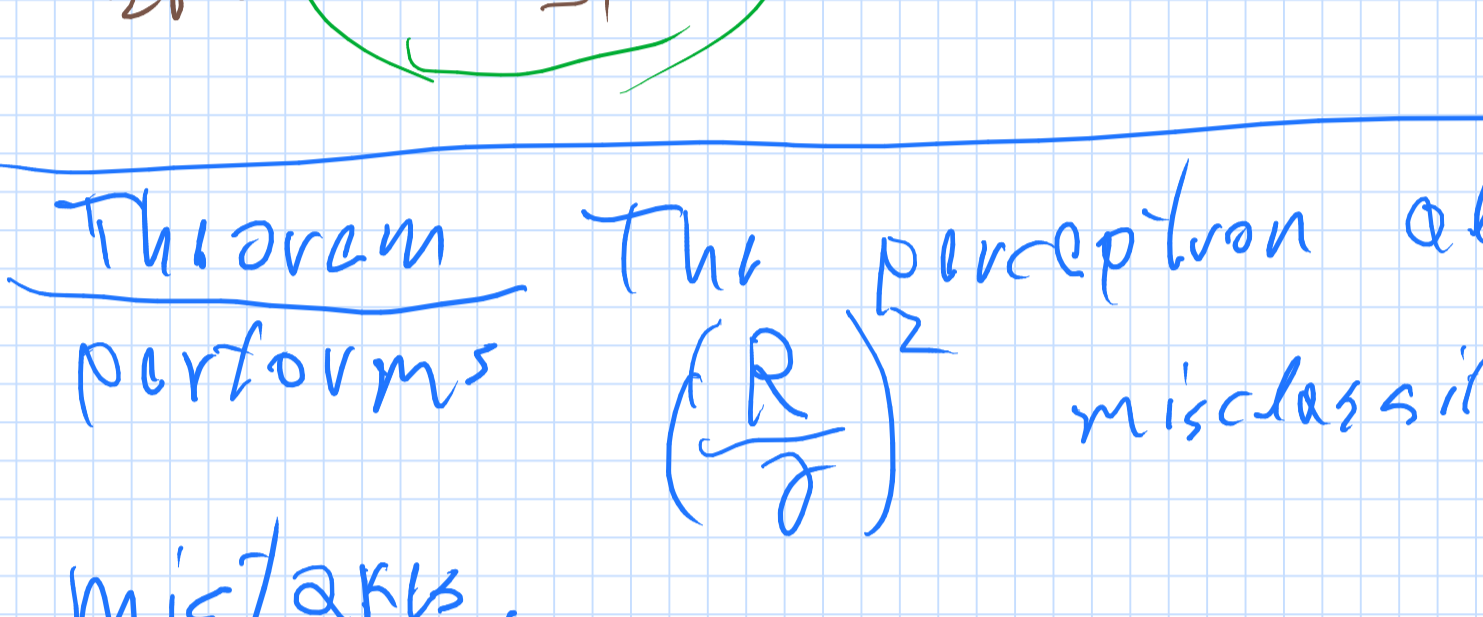
$\forall i \quad \langle w, p_i \rangle + b \geq \gamma$
 $\forall j \quad \langle w, n_j \rangle + b \leq -\gamma$
 γ tiny constant. Linear program



$(v_1, y_1), \dots, (v_n, y_n)$ $v_i \in \mathbb{R}^n$
 $y_i \in \{-1, +1\}$

Looking for a separating hyperplane through the origin.

$w_0 \leftarrow 0$
 $i = 0$
 While there is a misclassified example (v, y) do:
 $\text{sign}(\langle w_{i-1}, v \rangle) \neq y$
 $w_i \leftarrow w_{i-1} + yv$
 return $f(x) = \langle w_i, x \rangle$ as the classifier



Theorem The perceptron alg performs $\left(\frac{R}{\gamma}\right)^2$ misclassifications at most.

No dependency on d !
 No dependency on n !

Proof potential function

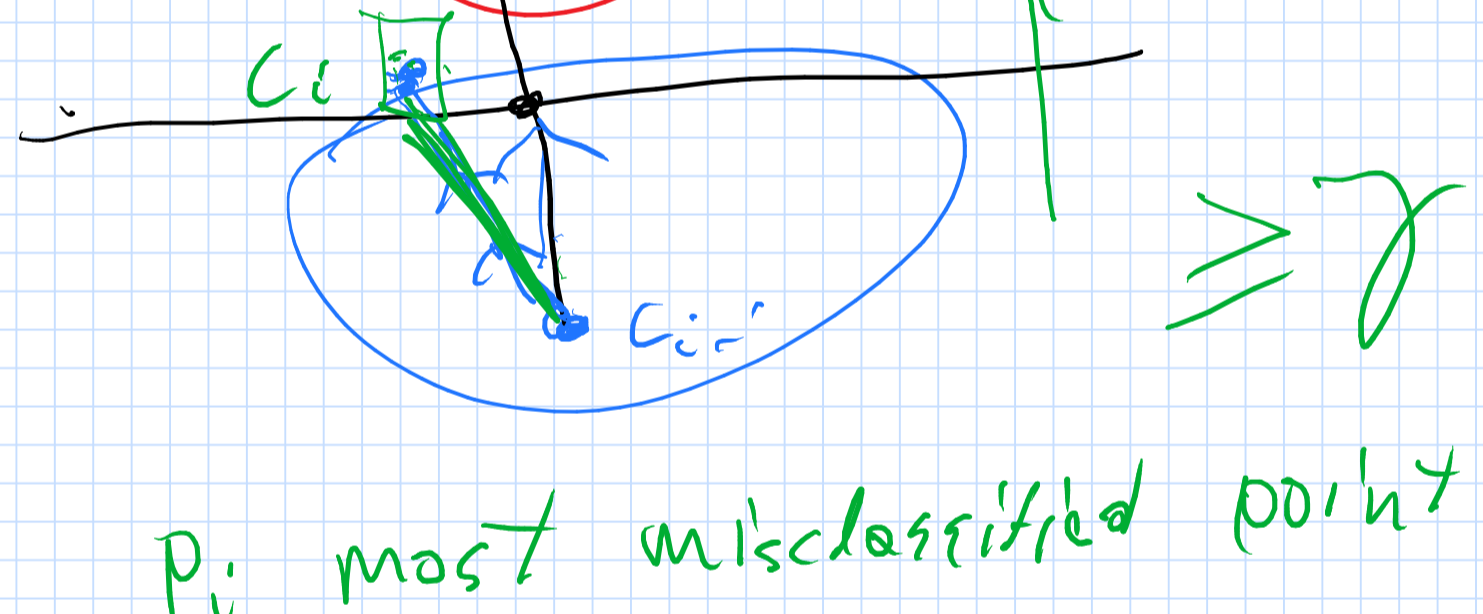
$\alpha_i = \|w_i - \frac{R^2}{\gamma} w_{opt}\|^2$

$\|w_{opt}\| = 1$
 w_{opt} is the normal of the perfect classifier.

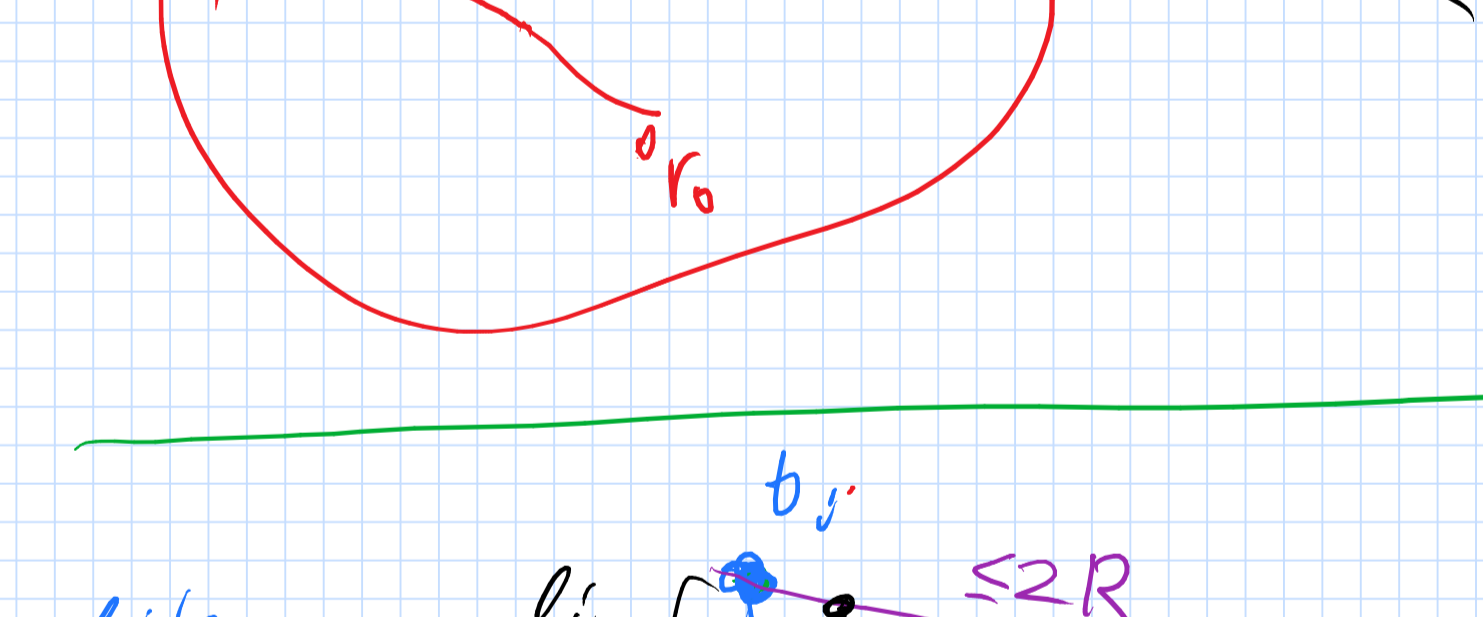
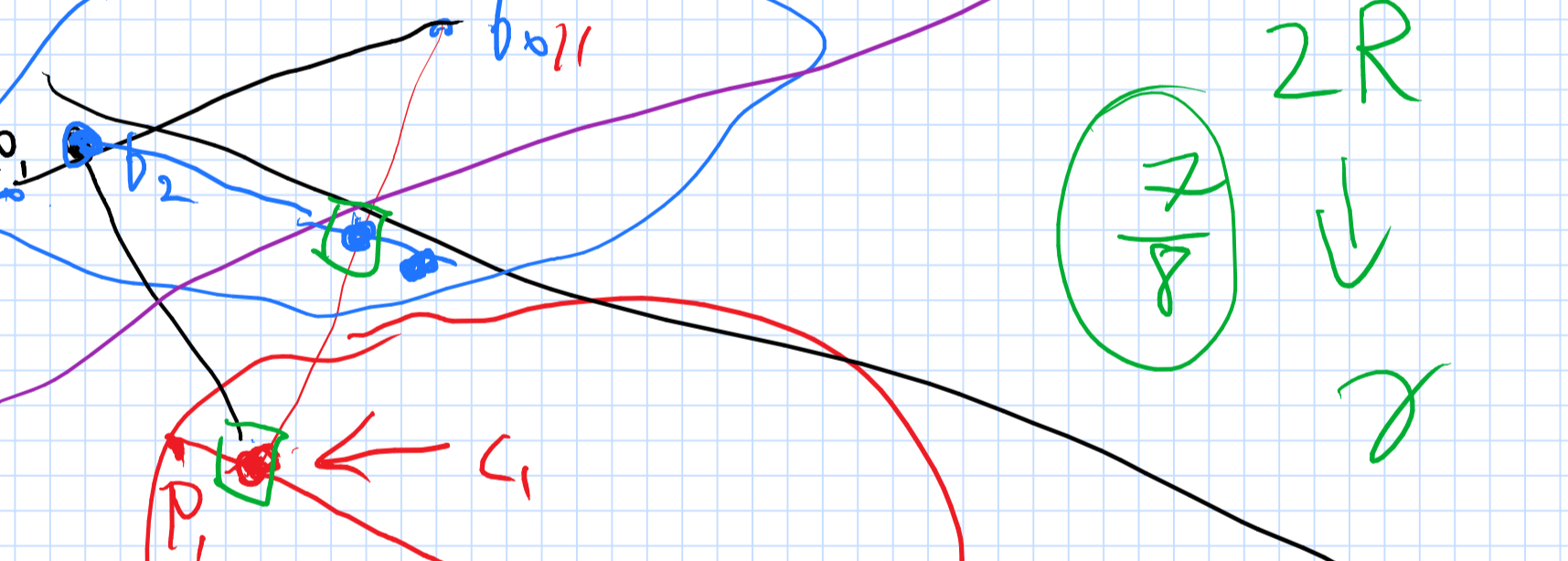
$\alpha_{i+1} \leq \alpha_i - R^2$

$w_0 = 0$
 $\alpha_0 = \|0 - \frac{R^2}{\gamma} w_{opt}\|^2 = \frac{R^4}{\gamma^2}$

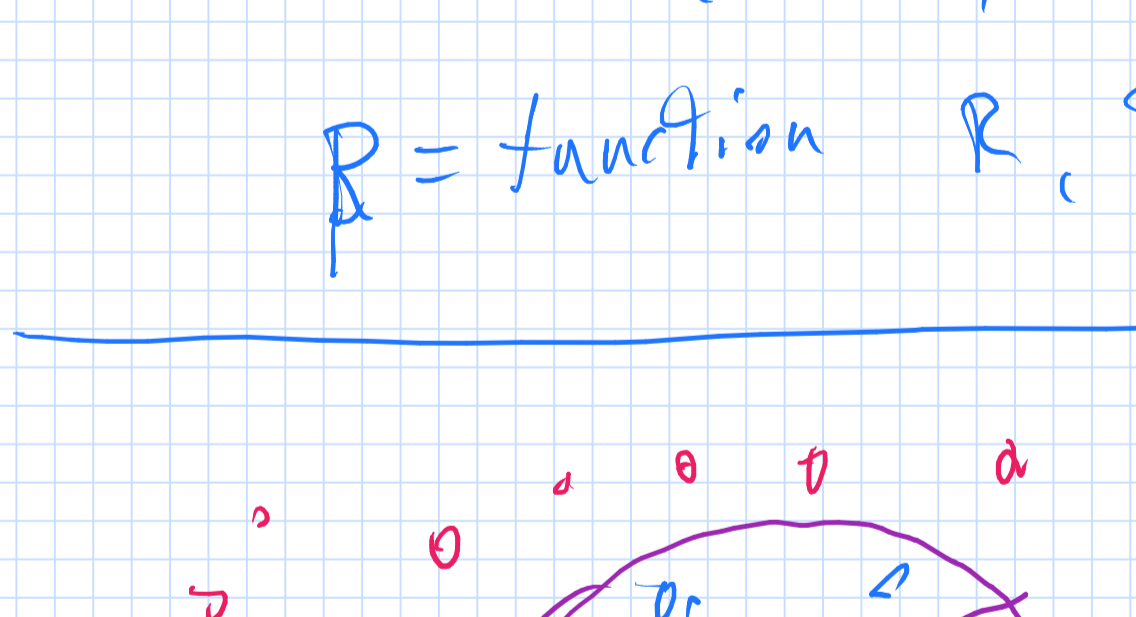
iterations $\leq \frac{R^4/\gamma^2}{R^2} = \frac{R^2}{\gamma^2}$



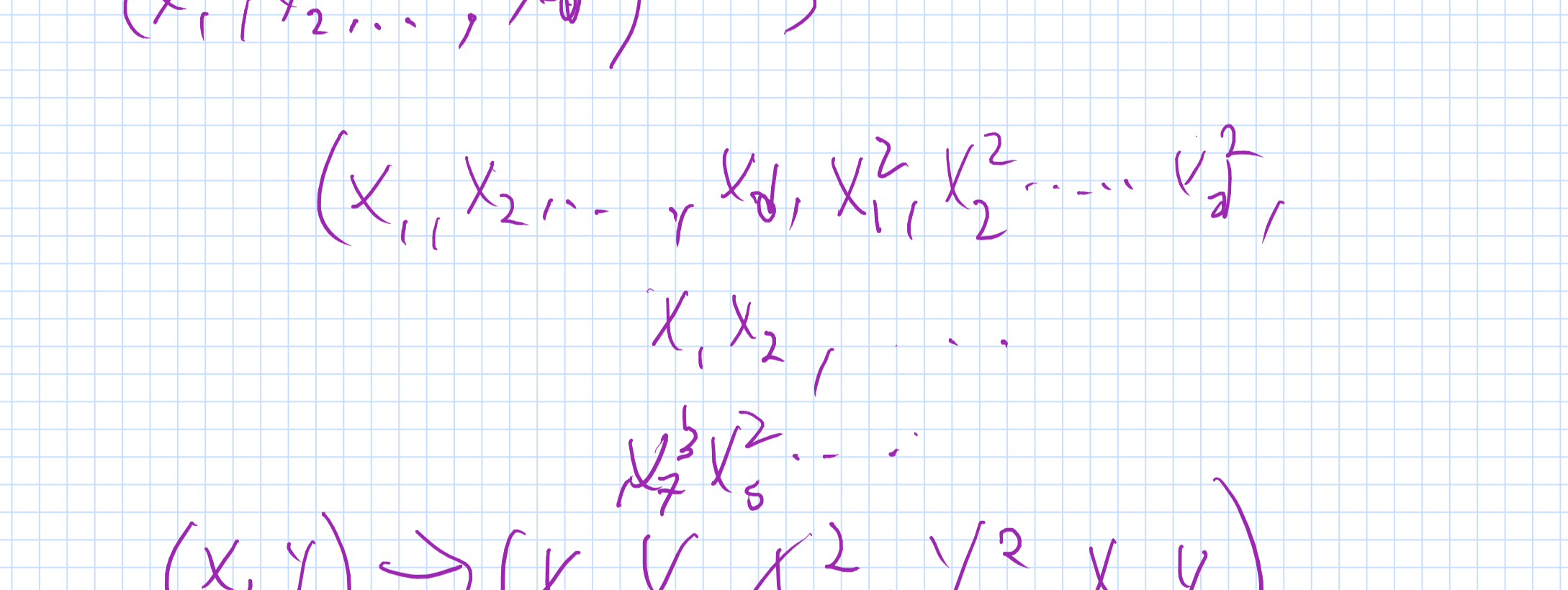
p_i most misclassified point (blue)
 $S_i = \text{segment}(c_i, p_i)$
 $c_i = \text{closest point on } S_i \text{ to the current red point}$



$l_{i+1} \leq d_i(1 - \beta)$
 $\beta = \text{function } R, \gamma$



Linear classifier $\mathbb{R}^5 \Rightarrow$ quadratic polynomial in the plane



$\langle w_i, v \rangle$
 (origin point) $\xrightarrow{\text{dotting}}$ $\langle v, v \rangle$
 $\langle v, v \rangle$ $\xrightarrow{\text{dotting}}$ $\langle w, w \rangle$

