In a regression task, the outputs are

- Continuous.
- Binary.
- Oiscrete.

Latent features are

- Correlated features.
- Non observable features.
- Non correlated features.

Let \mathcal{H} be a set of hypotheses. Let $\mathcal{Z} = X \times Y$ be the joint space over the feature space X and label space Y. A loss function ℓ is defined as:

- $0 \quad \ell: \mathcal{X} \times X \to \mathbb{R}^+$

Which risk is defined by $\mathbb{E}_{z \sim S} \ell(h, z)$?

- The true risk.
- The empirical risk.
- The generalization risk.

Let z = (x, y) be a labeled data and h a learned classifier. Let $\ell_{0,1}(h, z)$ be the (0/1)-loss.

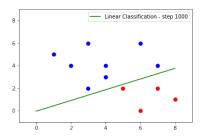
- $\ell_{0,1}(h,z)$ is convex.
- **6** $\ell_{0,1}(h,z)$ is differentiable everywhere.
- $\ell_{0,1}(h,z) = 1$ if $yh(x) \le 0$ and 0 otherwise.

Let $\theta = (\theta_1, \theta_2)$ be a set of parameters. The gradient of the function $f_{\theta}(x) = 4\theta_1^2 x + 6y\theta_2\theta_1$ with respect to θ_1 and θ_2 is:

- **6** $(8\theta_1x + 6y\theta_2, 6y\theta_1)$.
- \bullet 6 $y\theta_1$.

Which ℓ_p -norm is non convex?

- \bullet $\ell_{0.5}$ -norm
- \bullet ℓ_1 -norm
- \circ ℓ_2 -norm



Let S be a set of training examples plotted in the figure above.

What can you conclude from the figure?

- The true risk equals 0.
- The empirical risk equals 0.
- Both the true and empirical risks equal 0.

A differentiable function *F*:

- lacktriangle decreases fastest if one goes in the direction of the negative gradient of F.
- lacktriangle decreases fastest if one goes in the direction of the positive gradient of F.
- lacktriangle increases fastest if one goes in the direction of the negative gradient of F.