## Supervised Learning: Learning from a randomly sampled batch of labeled data

A a function from datasets  
to predictors  

$$Ex: A(D) = \hat{f}$$
 where  $\hat{f}: \chi = y$   
 $\hat{f}(x) = \begin{cases} y_i & \text{if } x = x_i \text{ for some} \\ i \in \{1, ..., n\} \end{cases}$ 

Otherwise

1: (x x y) → {f |f: x > y}

Ex (of features and labels/targets):

\$\times \times \times^3 \times of rooms, \times of floors, age of a house

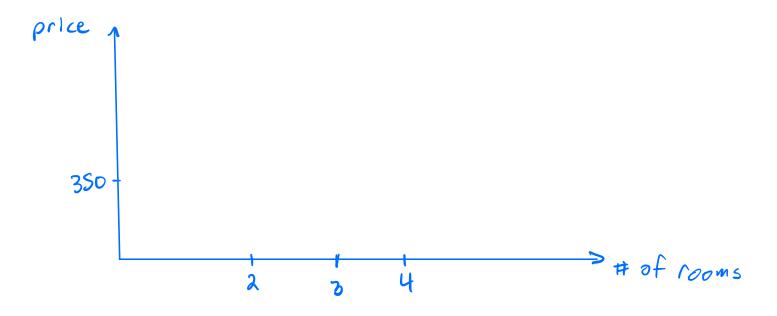
Y; \in \times \times price

\times^2 \times amount of chemical 1, amount of chemical 2

Y; \in \times 0, \times \

What is a feature and what is a label is a design choice. Usually a feature is info that is easy to gather. And the label is hard, which is why you want to predict it

·		



## Regression: YEY represent something with a notion of order

(Usually y is Ror some interval)

Ex: house prices, stack prices, energy consumption, weather prediction

We use:

$$L(f(\bar{X}),Y) = |f(\bar{X})-Y| \text{ absolute loss}$$

$$L(f(\bar{X}),Y) = (f(\bar{X})-Y)^2 \text{ squared}_{loss}$$

01

Classification: if YEY represents something without order

(Vsually 7 is finite)

Ex(of y): type wines, type of image, type of email, type of disease

Ex. f(x) is a predictor that takes as input the amount of a chemical in a wine and outputs the type of wine

Suppose you got multiple wines, what would a good f be?

B months a.

400 amount of a chemical

for I we use:

$$\mathcal{L}(f(\vec{x}),Y) = \begin{cases} 0 & \text{if } f(\vec{x}) = Y \\ 1 & \text{otherwise} \end{cases}$$

$$E_{X}$$
: L(f) if we use 0-110ss  $y = \{A, B, C, D\}$ 

$$L(f) = \mathbb{E}[l(f(\bar{x}), Y)] = \int_{\mathcal{X}} \mathbb{E}[l(f(\bar{x}), y) p(x, y) dx$$

$$= \int_{\mathcal{X}} \left( \sum_{y \in \mathcal{Y}} \mathcal{L}(f(\bar{x}), y) p(y|x) \right) p(x) dx$$

Ex: Let 7 be all linear functions

ERM picks the line that best fits the data

price

# of rooms