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† Last Time: Continued...

• $y \in \{0,1\} = \mathcal{Y}$ output space.

= true system (not a model, but it's sometimes called the "true model")

$y = f(z_1, z_2, z_3)$, where z_i

z_1 = has sufficient funds,

z_2 = unforeseen emergency

z_3 = criminal intentions.

- Say $y = z_1(1 - z_2)(1 - z_3)$

- Big Problem: z_1, z_2, z_3 are unobservable.

Also, we can't tell the future.

- So, impossible to get $\{z_1, z_2, z_3\}, t \rightarrow$

Next Best: Try to define and collect information "related" to the true causal inputs:

X_1 = Salary: How to measure? Avg. over yr.

X_2 = Previous loans repayments $\in \{0,1\}$ ^{never defaulted?}

X_3 = Previous crime type $\{ \text{no crime, infraction, misdemeanor, felony} \}$ ^{at least one?}

- Business-as-usual: Use what you have or use what is cheaply

is available.

• Example: Bob's Information:

$\vec{x} = [x_1, x_2, x_3] \in \mathcal{X}$, $\dim(\vec{x}) = 3$ aka \mathcal{P} or \mathcal{D} .

↑
observation,
record,
object, input
"independent
variables"

↑ features, variables (dependent), characteristics, attributes, regressors, covariates, predictors

→ \mathcal{X} is called "covariate space."

$x_1 \in \mathbb{R}$ "continuous variable"

$x_2 \in \{0,1\}$ "binary or dummy variable"

x_3 is a "categorical variable with

4 "levels" = unique possible values.

- How do we deal with this?

Some ideas:

a) Code it numerically, e.g.

$$X_3 \in \{0, 1, 2, 3\}$$

→ but this should only be done if the categorical predictor is "ordinal!"

b) Dummification:

$$X_3 \rightarrow X_{3a}, X_{3b}, X_{3c}, X_{3d}$$

Binary Binary

$$\text{e.g. } X_3 = \text{misdemeanor } \vec{x} = [0 \dots 0 \ 0 \ 1 \ 0]$$

→ Problem: Now we have more ^(dimensional) covariates, i.e. p is now 10.

• So, impossible to get $\{z_1, z_2, z_3\}$ but we do have $\{x_1, x_2, x_3\}$.

Goal: Do the best we can in explaining

y by creating a model f , the approx; i.e. the best functional relationship we can get.

- Does $y = f(x_1, x_2, x_3)$ ever? No...

Instead, $y \approx f(x_1, x_2, x_3) \Rightarrow$

$$y = f(x_1, x_2, x_3) + \epsilon; \text{ where } \epsilon = \epsilon(\vec{z}) + \epsilon(\vec{x}).$$

comes from

- How do we get f ?

→ First note there is no analytical solutions.

e.g. $h(x) = x^2$ Find $\min(h)$.

→ Use an "Empirical Solution" i.e. use data,
⇒ "Learning from Data!"

→ "Supervised learning" Uses historical examples of records & their responses.

It requires 3 ingredients:

(I) $\mathcal{D} = \{ \langle \vec{x}_1, y_1 \rangle, \langle \vec{x}_2, y_2 \rangle, \dots, \langle \vec{x}_n, y_n \rangle \}$,

where \vec{x}_i is Bill's characteristics and y_i is whether or not he paid back the loan.

\vec{x}_2, \dots Jill's \dots , y_2, \dots , etc.

(II) $\vec{x} = \begin{bmatrix} \vec{x}_1 \\ \vec{x}_2 \\ \vdots \\ \vec{x}_n \end{bmatrix} \in \mathbb{R}^{n \times p}$ &

$\vec{y} = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix} \in \mathbb{R}^n$

$\Rightarrow \dim(\vec{x}) = n \cdot p$ & $\dim(\vec{y}) = n$.

goo.gl/i98SV6

[Github.com/gd6](https://github.com/gd6)