

normal linear models

$$\text{Data} = |y_1 \ y_2| \dots |y_n|$$

multilevel / hierarchical
linear mixed effects

model

for each i in $1 \dots n$

$$y_i \sim \underline{N(\mu_i, \sigma^2)}$$

$$\underline{\mu_i = \beta_0 + \beta_1 x_i}$$

$$\underline{\theta_i = f(\underline{x_i})} \quad \underline{\text{nonlinear}}$$

change distribution

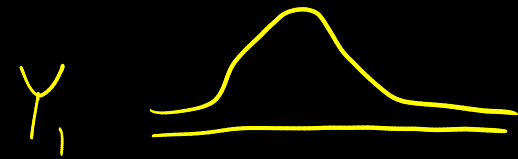
generalized linear models

link function

- classical
- Bayesian

population

sample



data: $\underline{y}_1 \underline{y}_2 \dots \underline{y}_n$ sample

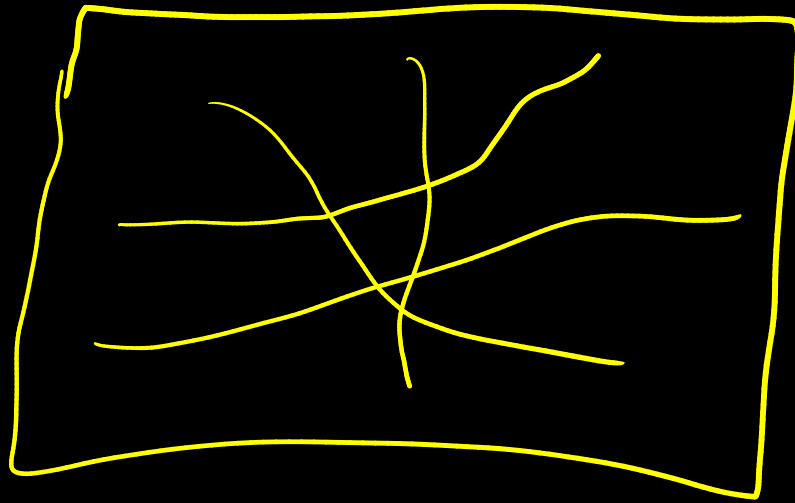
realizations of \underline{n} random variables

$\left[\begin{matrix} Y_1 & Y_2 & Y_3 & \dots & Y_n \end{matrix} \right]$

probability
distribution

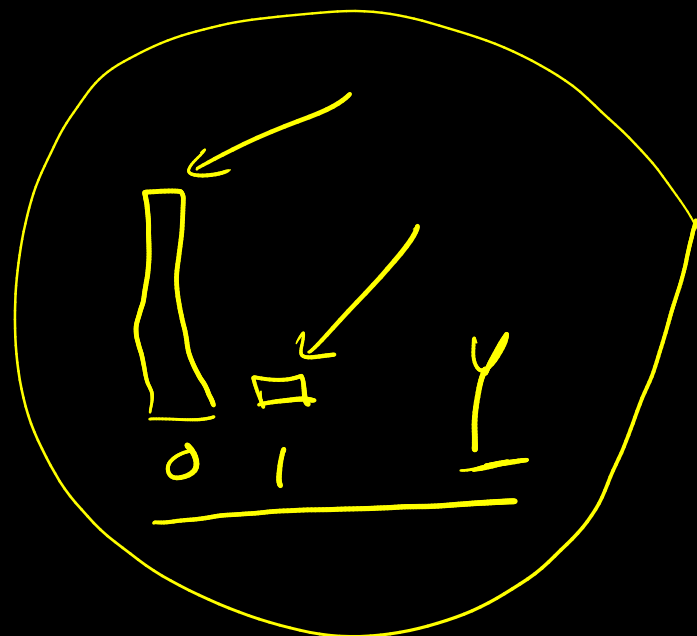
- population \Leftrightarrow probability
- sample
- model

Box:
all models are wrong
but
some are useful



$y_1, y_2, \dots, y_{5000}$

$y_1 = y_2 = y_3 \dots y_n$



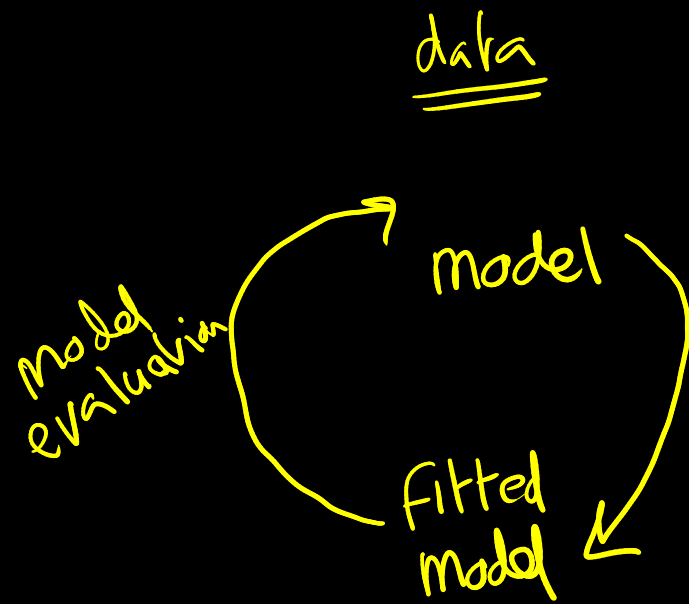
$y_1 \quad y_2 \quad \dots \quad y_n$

age X_{11}

sex X_{12}

ethnicity X_{13}

$Y_1 \quad Y_2 \quad \dots \quad Y_n$



inference

- classical
- Bayesian

$$y_i \sim N(\mu, \sigma^2)$$

\uparrow \uparrow

$y_1 \ y_2 \ \dots \ y_n$ ($n = 250$)

$y_i \in \{T, H\}$

$\in \{0, 1\}$

$m = 139$

model

each $\left[y_i \sim \text{Bernoulli}(\theta) \right]$

$$y_1 \quad y_2 \quad \dots \quad y_n$$

$$Y_1 \quad Y_2 \quad \dots \quad Y_n$$

$$Y_1 = Y_2 = Y_3 \dots Y_n$$

iid

$$(1-\theta)^{\sum_0} \theta^{\sum_1}$$

Data: y_1, y_2, \dots, y_n

model: $y_i \sim \text{bernoulli}(\theta)$ for i in $1 \dots n$

probability of data given θ

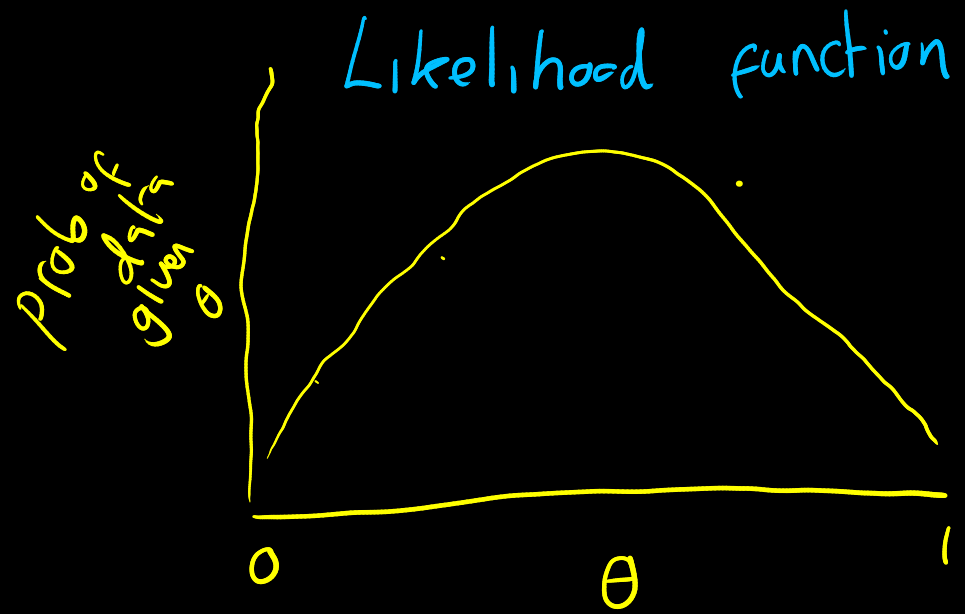
$$m = 139$$

$$n = 250$$

$$p(\text{Data} \mid \theta)$$

$$p(\text{Data} \mid \theta) = \binom{n}{m} \theta^m (1-\theta)^{n-m}$$

$$\frac{n!}{m!(n-m)!}$$



Classical inference

- estimator
- sampling distribution of estimator
- test hypotheses
- confidence intervals

Bayes's theorem

Data : D

unknown : θ

posterior
distribution $p(\theta|D)$

=

$$\frac{\overset{\text{likelihood}}{p(D|\theta)} \overset{\text{prior}}{p(\theta)}}{\int p(D|\theta) p(\theta)}$$

