

210 = even
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maximum a posteriori - max after the fact
L most probable θ after you see the data.

$$\hat{\theta}_{\text{MAP}} := \arg \max_{\theta \in \{0,1\}} \{P(\theta|x)\}$$

↑ AFTER draw

This is one of 3 Bayesian Point Estimates.

$$X = \text{iid Bern}(\theta), n=3, x=(0,1,1) \quad \hat{\theta} = \{.5, .75\}$$

$$x \in \mathcal{X} = \{0,1\} \times \{0,1\} \times \{0,1\} \quad 2^3 = 8 \text{ data sets}$$

The arg of the
derives
is the
probability

	\mathcal{X}							
$\theta = .75$	(1,1,1)	(1,1,0)	(1,0,1)	(0,1,1)	(1,1,0)	(0,1,0)	(0,1,1)	$\hat{\theta}$
$\theta = .5$	(1,1,1)	(1,1,0)	(1,1,1)	(0,1,1)	(1,0,0)			

.5
↑
by log

$$P(X=(1,1,1) | \theta=.75) = (.75)^3 = .422 \quad 42\%$$

$$P(X=(1,1,0) | \theta=.75) = (.75)^2 (.25) = .141 \quad 14\%$$

Since X is iid Bern(θ) $P(X=(1,1,0) | \theta=.75) = P(X=(1,0,1) | \theta=.75)$

$$P(X=(1,0,0) | \theta=.75) = .047 \quad 4\%$$

$$P(X=(0,0,0) | \theta=.75) = .25^3 = .016 \quad 1.6\%$$

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$$P(X=(1,1,1) | \theta=.5) = .125$$

∴ All will be the same.

hence the bottom half should be 8 even boxes

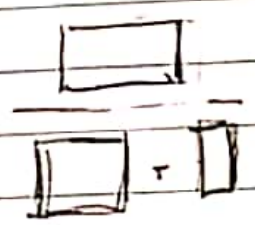
$$P(X=(1,0,0)) = P(X=(1,0,0), \theta=.75) + P(X=(1,0,0), \theta=.5)$$

$$.047 + .125$$

$$P(\theta=.5 | X=(1,0,0)) = \frac{P(\theta=.5 \cap X=(1,0,0))}{P(1,0,0)}$$

$$= .125$$

$$.047 + .125$$



For discrete parameter spaces

$$\sum_{\theta \in \Theta} P(\theta) = 1 \quad \sum_{\theta \in \Theta} P(\theta | x) = 1$$

$$\sum_{\theta \in \Theta} P(x | \theta) = \text{could be anything}$$

(If $P(\theta)$ is constant
Laplace's idea)

$$P(\theta | x) = \frac{P(x, \theta)}{P(x)} \propto P(x | \theta) \propto P(x | \theta) P(\theta) \propto P(x | \theta)$$

By principle of difference

$$\hat{\theta}_{MLE} = \underset{\theta \in (\pm 1)_0}{\text{arg max}} \{ P(\theta | x) \} = \underset{\theta \in (\pm 1)_0}{\text{arg max}} \{ P(x | \theta) P(\theta) \} = \underset{\theta \in (\pm 1)_0}{\text{arg max}} \{ P(x | \theta) \}$$

only Equal, if
the MLE is in
the parameter set you
specify, $\hookrightarrow (\pm 1)_0$

if?
 $\hat{\theta}_{MLE}$

$$\text{Let } (\pm 1)_0 = \{ .1, .25, .5, .75, .9 \}, (\text{Laplace Prior})$$

$$\text{let } x = (1, 1, 0)$$

$$P(x | \theta = .1) = .009$$

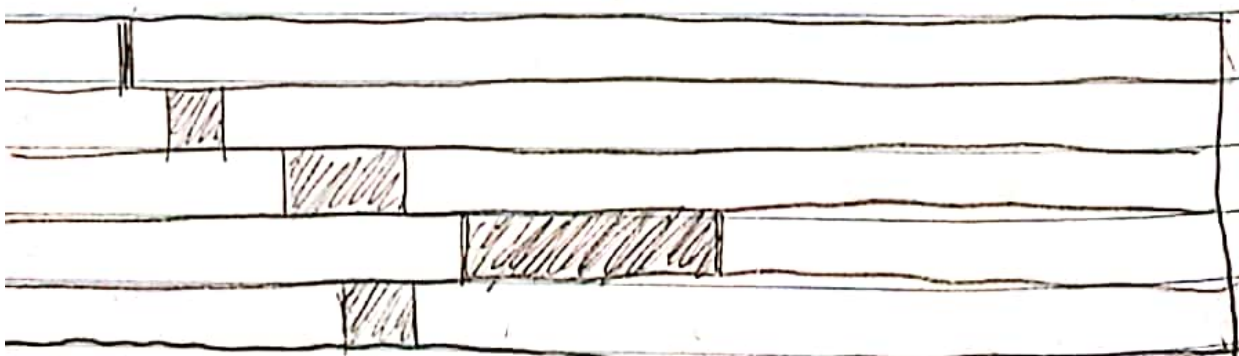
$$P(x | \theta = .25) = .047$$

$$P(x | \theta = .5) = .125$$

$$P(x | \theta = .75) = .141$$

$$P(x | \theta = .9) = .081$$

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$$\hat{\theta}_{MAP} = .75$$

yes? area

$$P(\theta = .75 | x = (1, 1, 0)) = \frac{P(x | \theta = .75) P(\theta = .75)}{P(x)}$$

+

+

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MLE is not in this set Θ_0 . So what do we do?

Let's examine Laplace's prior under many different parameter spaces (approaching the full space).

from to give θ every " (0,1)"

$$\Theta_{0,3} = \left\{ \frac{1}{4}, \frac{1}{2}, \frac{3}{4} \right\} \Rightarrow P(\theta) = U(\Theta_{0,3}) = \left\{ \frac{1}{3} \forall \theta \right\}$$

from prob by Laplace prior results (0,1)"

$$\Theta_{0,9} = \left\{ \frac{1}{10}, \frac{2}{10}, \dots, \frac{9}{10} \right\} \Rightarrow P(\theta) = U(\Theta_{0,9}) = \left\{ \frac{1}{9} \forall \theta \right\}$$

from prob by Laplace prior results (0,1)"

$$\Theta_{0,n} = \left\{ \frac{1}{n+1}, \frac{2}{n+1}, \dots, \frac{n}{n+1} \right\}$$

$$= P(\theta) = U(\Theta) = \left\{ \frac{1}{n} \forall \theta \right\}$$

from full parameter set for Bern

$$(0,1) = \lim_{n \rightarrow \infty} \Theta_{0,n} \Rightarrow P(\theta) = 0 \forall \theta \text{ not a PMF !}$$

from full parameter set for Bern

$$\lim_{n \rightarrow \infty} F_n(\theta) = F(\theta) = \theta \Rightarrow P(\theta) = F'(\theta) = 1$$

$$\Rightarrow P(\theta) = U(0,1)$$

ie cts

$\tilde{F} = \text{ind Bern}(\theta) \quad x = (1, 1, 0) \quad P(\theta) = U(0,1)$

$$P(\theta|x) = \frac{P(x|\theta) P(\theta)}{P(x)} = \frac{P(x|\theta)}{P(x)} = \frac{P(x|\theta)}{\int P(x, \theta) d\theta}$$

(H)

$$= \frac{P(x|\theta)}{\int P(x|\theta) P(\theta) d\theta} = \frac{\theta^2(1-\theta)}{\int_0^1 \theta^2(1-\theta) d\theta} = \frac{\theta^2(1-\theta)}{\left[\frac{\theta^3}{3} - \frac{\theta^4}{4} \right]_0^1} = \frac{\theta^2(1-\theta)}{\frac{1}{12}} \rightarrow$$

(H)

mean, median, mode

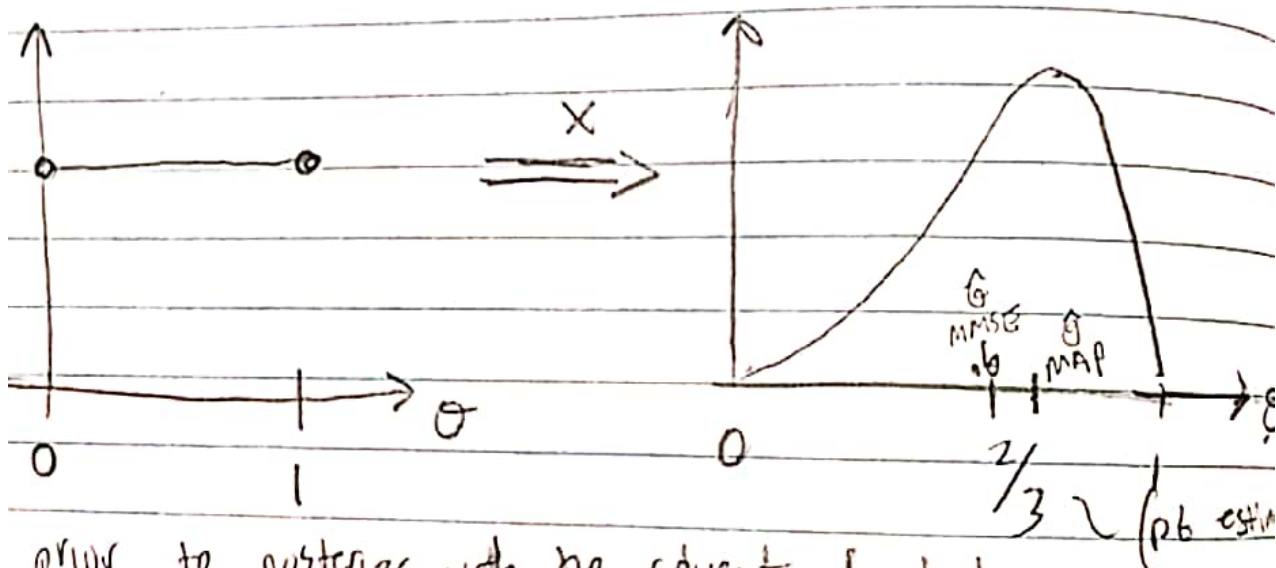
$$= 12 \theta^2 (1-\theta) = P(\theta | x)$$

$$\hat{\theta}_{MAP} = \operatorname{argmax} \{ 12 \theta^2 (1-\theta) \} = \operatorname{argmax} \{ \theta^2 (1-\theta) \}$$

(take derivative and set to 0)

$$= \operatorname{argmax} \{ \mathcal{L}(\theta; x) \} = \frac{2}{3}$$

$$P(\theta) \xrightarrow{x} P(\theta | x)$$



prior to posterior with the advent of data

$$P(\theta > .5 | x) = \int_{.5}^1 12 \theta^2 (1-\theta) d\theta = .688$$

the coin is tilted
heads

We talked about the MAP Bayesian point estimate. Are there other measures of "best guess there" if you have posterior distribution $P(\theta | x)$?

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$$\hat{\theta}_{\text{MMSE}} := E[\theta | x] = \arg \min_{\theta \in \Theta} \left\{ E[(\theta - \hat{\theta})^2 | x] \right\}$$

The minimum mean squared error Bayesian point estimate is the posterior mean (expectation). In our case:

$$\hat{\theta}_{\text{MMSE}} = \int_0^1 \theta P(\theta | x) d\theta = \int_0^1 \theta 12 \theta^2 (1-\theta) d\theta = \frac{12}{20} = \frac{3}{5}$$

(this is the default estimator)