

Process strategy

Bernoulli process, unknown parameter $\theta = \text{prob of success}$

Do 7 trials, 6 success, 1 fail \exists random var whose dist. is related to θ , so can we then probabilistically

what can we do?

estimate $\hat{\theta} = \bar{X} \Rightarrow \theta \approx \frac{6}{7}$

$P(\theta > \frac{1}{2})$?

Can't ask this.

θ is not a random var - its number.

this is a good estimator - min var unbiased.

confidence interval?

e.g. find estimators $\hat{\theta}_{sm}$, $\hat{\theta}_{big}$ s.t.

$$P(\hat{\theta}_{sm} < \theta < \hat{\theta}_{big}) \geq .90$$

prob. statement.

after measurement, rough approx: $\sum X_i = n\bar{X}$ is binomial
 either true or false, not a prob statement. w/ mean $n\theta$ and var $n\theta(1-\theta)$

so
$$\frac{n\bar{X} - n\theta}{\sqrt{n\theta(1-\theta)}} \approx Z$$

"

$$\sqrt{n} \frac{\bar{X} - \theta}{\sqrt{\theta(1-\theta)}}$$

so, say $\sqrt{7} \frac{\bar{X} - \theta}{\sqrt{\theta(1-\theta)}} \geq 2$ p = 0.5 that

$$\sqrt{7}(\bar{X} - \theta) \geq \sqrt{\theta(1-\theta)}$$

$$\sqrt{7}\bar{X} \geq \sqrt{\theta(1-\theta)} + \sqrt{7}\theta$$

$$\bar{X} \geq \frac{\sqrt{\theta(1-\theta)}}{7} + \theta$$

graph

Bayesian Approach

Notre in previous case, we knew nothing about process beforehand. no info prior.

Imagine: have a box of white ones,
equally likely to get one of any prob $\theta \in \mathcal{I}$
we get one out.

this reps a population (of flips)
with an unknown parameter θ .
but θ is not as unknown as before.

previously, couldn't say $P(\theta > 1/2) = ?$

but now we can!

knowing $f(x|\theta) = \theta$ and $g(\theta) = \begin{cases} 1 & \theta \in [0,1] \\ 0 & \text{else} \end{cases}$
we can consider a "joint distribution"

$$f(x, \theta) = f(x|\theta)g(\theta)$$

what if we get a success w/ one trial?
get a new (posterior) dist for θ

$$g(\theta | X=1)$$

via conditional parameter dist

$$g(\theta | x) = g(\theta | X=x) = \frac{f(x, \theta)}{f(x)} \leftarrow \text{maximize } \frac{\partial}{\partial \theta}$$

$f(x, \theta) = f(x|\theta)g(\theta)$

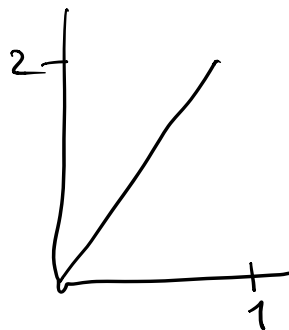
$$f(x) = \int f(x, \theta) d\theta$$
$$= \int f(x|\theta)g(\theta) d\theta$$

For example, in our case:

$$g(\theta | X=1) = \frac{f(x|\theta)g(\theta)}{\int f(x|\theta)g(\theta) d\theta}$$

$$= \begin{cases} \frac{\theta \cdot 1}{\int_0^1 \theta d\theta} & \theta \in [0, 1] \\ 0 & \text{else} \end{cases}$$

$$= \begin{cases} 2\theta & \theta \in [0, 1] \\ 0 & \text{else} \end{cases}$$



Straightforward in principle,
in practice, either need to do numerically
or leverage tons of nice answers.
note: in general, $X \rightsquigarrow X_1, \dots, X_n$!

Nice answer?

Start w/ some dist. $f(x|\theta)$
and an a priori dist $g(\theta)$ for θ .

now, the a priori dist. has some particular form
assume $g(\theta) = g(\theta, \varphi)$ w/ param. φ

φ specified

note, this is a "known"
dist.

nice means given as sample,

$X_1, \dots, X_n = x_1, \dots, x_n$, the a posteriori dist
 $g(\theta | x_1, \dots, x_n)$ is has the form

$g(\theta, \varphi')$
same φ' .

Ex:

if $f(x, \mu)$ normal w/ known σ^2 , unknown μ .

can consider an a priori dist for $g(\mu)$ for μ via a random var $M, N(\mu_0, \sigma_0^2)$

Can then consider $g(\mu | \bar{X} = \bar{x})$

get another normal variable

of type $N\left(\frac{n\bar{x}\sigma_0^2 + \mu_0\sigma^2}{n\sigma_0^2 + \sigma^2}, \frac{\sigma^2\sigma_0^2}{n\sigma_0^2 + \sigma^2}\right)$

Can flip example?

if you keep flipping, the posterior
 $n > 1$

dist. for θ is a Beta dist!

recall this is

$$g(\theta) = \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} x^{\alpha-1} (1-x)^{\beta-1} \quad x \in [0, 1]$$

$$g(\theta | \alpha, \beta)$$

↑
"φ"

for $\alpha=1, \beta=1$ this is constant
 $g(\theta) = 1$.

for $\alpha=2, \beta=1$ get $g(\theta) = 2x$

↑
a posteriori for $\bar{X} = 1, n=1$

and in general, a posteriori for $\bar{X} = \bar{x}$
so $n\bar{x}$ successes
"k"

get

$$\alpha = k + 1$$
$$\beta = n - k + 1$$

↑
"φ"

$$\sqrt{z}(\bar{x} + \theta) \geq \sqrt{\theta(1-\theta)}$$

$$x - \gamma = \sqrt{\gamma(1-\gamma)}$$

$$(x - \gamma)^2 = \gamma(1-\gamma)$$

$$x^2 - 2x\gamma + \gamma^2 = \gamma(1-\gamma)$$

$$= \gamma - \gamma^2$$

$$x^2 - 2x\gamma + 2\gamma^2 - \gamma = 0$$

$$x^2 - 2x\gamma + 2\gamma^2 - \gamma z = 0$$

$$\gamma = 0 \quad z = 1 \quad x = 1$$

project.