

Math 3236 Statistical Theory

1/26/23

Beta distribution and

Bayes Updates

(sect 5.8, 7.1 & 7.2)

Coin flip: an experiment with two outcomes: 0, 1

Probability of 1, called p .

Assume that we know nothing at the beginning that is our prior distribution on p is uniform.

$$f(p) = \begin{cases} 1 & 0 \leq p \leq 1 \\ 0 & \text{otherwise} \end{cases} \quad \text{prior}$$

Make one exp. and The result is

\mathcal{I} :

$$f(p | \mathcal{I}) = \frac{P(\mathcal{I} | p) f(p)}{P(\mathcal{I})}$$

$$= 2p$$

posterior

$$f(p | \mathcal{O}) = 2(1-p)$$

————— \mathcal{O} —————

I flip my coin N Times

and I get n Heads (\mathcal{I})

$$f(p | n) = \frac{\binom{N}{n} p^n (1-p)^{N-n}}{\int \binom{N}{n} p^n (1-p)^{N-n}}$$

number of
H

$$f(p|n) = \frac{p^n (1-p)^{N-n}}{\int_0^1 p^n (1-p)^{N-n}}$$

Beta distribution.

Def: Beta function:

$$B(\alpha, \beta) = \int_0^1 x^{\alpha-1} (1-x)^{\beta-1} dx$$

Th:

$$B(\alpha, \beta) = \frac{\Gamma(\alpha) \Gamma(\beta)}{\Gamma(\alpha + \beta)}$$

$$B(n, m) = \frac{(n-1)! (m-1)!}{(n+m-1)!}$$

$$\int_0^1 x^{n-1} (1-x)^{m-1} dx = \frac{(n-1)! (m-1)!}{(n+m-1)!}$$

$$\int x^n (1-x)^m dx =$$

$$\frac{m}{n+1} \int x^{n+1} (1-x)^{m-1} dx \dots$$

$$x^\alpha = \frac{1}{\Gamma(\alpha)} \int_0^\infty t^\alpha e^{-tx} dt$$

$$f(x | \alpha, \beta) = \frac{x^{\alpha-1} (1-x)^{\beta-1}}{B(\alpha, \beta)}$$

$$= \frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha) \Gamma(\beta)} x^{\alpha-1} (1-x)^{\beta-1}$$

If $X \sim \text{Beta}(\alpha, \beta)$ Then

$$E(X) = \frac{\alpha}{\alpha + \beta}$$

$$V(X) = \frac{\alpha \beta}{(\alpha + \beta)^2 (\alpha + \beta + 1)}$$

$B(1, 1) =$ uniform

If we start our experiment with the prior that $P \approx B(1, 1)$

$f(p, n)$ be T_n dist.

with $\alpha = n + 1$ $\beta = N - n + 1$

Suppose now that my prior is

$B(\alpha, \beta)$

after observing n I in N attempts The posterior

is $B(\alpha + n, \beta + N - n)$

Bernoulli and B are
conjugate family

Updating rule for a Bern.
exp. Takes a B dist and
returns a B dist.

Since a B dist. is
identified by the two par.
 α, β , the update can be
written as an updating rule

on α, β

$$\alpha \rightarrow \alpha + n$$

$$\beta \rightarrow \beta + N - n$$

$$E(P) = \frac{p}{2} \quad \frac{\alpha}{\alpha + \beta} = \frac{1}{2}$$

$$\alpha = \beta \implies E(P) = \frac{1}{2}$$

The larger α The more

convinced you are that

The coin is fair.



$$E(P | n) = \frac{\alpha + n}{\alpha + \beta + N}$$

$$\text{if } \alpha = \beta$$

$$= \frac{\alpha + n}{2\alpha + N}$$

if The coin is actually fair

Then, in probability,

$n = \frac{N}{2}$ for N large.

$E(P, n)$ converge in probability to $\frac{1}{2}$ when $N \rightarrow \infty$.

If prob of H is p

Then, in prob,
 $n = Np$

$$\lim_{N \rightarrow \infty} \frac{\alpha + Np}{\alpha + \beta + Np} = p$$

If I estimate q

$\min_q E((P - q)^2 | n)$ is reached
for $q = E(P | n)$.

Ex:

$$\min_m \mathbb{E}((X - m)^2) = \text{Var}(X)$$

$$\arg \min = \mathbb{E}(X)$$

If cost $\mathbb{E}(|P - q|)$

Then The best $q = \text{median of } P/n$

$$B(1, 1) \xrightarrow{0} B(2, 1) \xrightarrow{1} B(2, 2)$$

$$B(1, 1) \xrightarrow[\text{one}]{\text{zero}} B(2, 2)$$

X_i are i.i.d. Bernoulli r.s.

$$\mathbb{E}(X_i) = p$$

$$\hat{p} = \frac{1}{N} \sum_{i=1}^N X_i = \bar{X}$$

$$E(\hat{p}) = p$$

$$\hat{p} \rightarrow p$$

for N large

$$\frac{\sqrt{N}}{\sqrt{p(1-p)}} (\bar{X} - p) \Rightarrow N(0, 1)$$

$$B(Np, N(1-p)) \xrightarrow{N \rightarrow \infty} \text{?}$$