

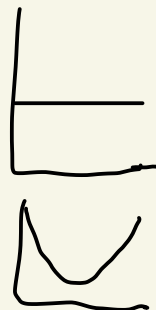
$$\text{Beta}(a, b) = x^{a-1} (1-x)^{b-1} \quad 0 < x < 1$$

- Flexible family of distributions on $(0, 1)$
- Often used as a prior in CG1
- Conjugate Prior to Binomial & many connections to interdisciplinary fields

with $(a=1, b=1)$, beta =

with $(a=2, b=2)$

beta =



Conjugate Prior for Binomial

$X|p$ is $\text{Bin}(n, p) \rightarrow$ if we know p , the probability is binomial

When p is not known, p is $\text{Beta}(a, b) \leftarrow$ prior (not based on data)

Get evidence, update using bayes rule \rightarrow Find Posterior $p|x$

$$f(p|x=k) = \frac{P(x=k|p) f(p)}{p(x=k)} = \frac{\binom{n}{k} p^k (1-p)^{n-k} \cdot (p^{a-1} (1-p)^{b-1})}{p(x=k)}$$

Does not depend on the data (no p)

given p , what is probability, $x=k$?

density function Beta

Ignore constants by using proportions: $f(p|x=k) \propto p^{k+a-1} (1-p)^{n-p+b-1}$

$$p(p|x) \sim \text{Beta}(a+x, b+n-x) \quad (x=k-1)$$

Updating parameters stays as beta function

Intuition: Let our old data have $\text{Beta}(a, b)$

Then with x new successes and $n-x$ new failures our Beta is updated to $\text{Beta}(a+x, b+n-x)$

(Stays in the Beta family) \rightarrow very convenient

Gamma Distribution and Poisson Process

Given a sequence $0, 1, 2, x$: what could x be?

x can be $3, e, \frac{1}{e}, \pi, 720!$, whatever. Any sequence is ok.

Real Question: How do we extend a sequence of numbers?
Or, what is pi factorial?

$$\text{Stirling's formula} \rightarrow n! \approx \sqrt{2\pi n} \left(\frac{n}{e}\right)^n$$

Gamma Function

- Closely connected to gamma function
- Has a gamma distribution

$$\Gamma(a) = \int_0^{\infty} x^a e^{-x} \frac{dx}{x} \quad \text{for } a > 0$$

Behavior of Γ :

At 0, $\int \frac{dx}{x}$ is $\ln|x|$ which is $-\infty$ as $x \rightarrow 0$

At 1, still $\int \frac{dx}{x^2}$ is $-x^{-1}$ which is 0

At large a , it is just $\int e^{-x}$ because e^{-x} dominates x^a

$$\Gamma(n) = (n-1)! \quad \text{for } n \text{ a positive integer}$$

$$\Gamma(x+1) = x \Gamma(x)$$

$$\text{What is } \Gamma\left(\frac{1}{2}\right)? \rightarrow \sqrt{\pi}$$

$$\Gamma\left(\frac{3}{2}\right) = \frac{1}{2} \Gamma\left(\frac{1}{2}\right) = \frac{\sqrt{\pi}}{2}$$

Generating a PDF

$$1 = \int_0^{\infty} \frac{1}{\Gamma(a)} x^a e^{-x} \frac{dx}{x} \quad \leftarrow \text{Gamma}(a, 1) \text{ PDF}$$

Gamma distribution also related to exponential distribution

If x is randomly distributed with $\text{Beta}(a, b)$

$$\text{PDF} = f(x) = \frac{1}{\text{Beta}(a, b)} x^{a-1} (1-x)^{b-1} = \frac{\Gamma(a+b)}{\Gamma(a) \Gamma(b)} x^{a-1} (1-x)^{b-1}$$

$$\text{Beta}(a, b) = \int_0^1 x^{a-1} (1-x)^{b-1} dx$$

$$\text{Beta}(a, b) = \frac{\Gamma(a) \Gamma(b)}{\Gamma(a+b)}$$

Beta distribution is ideal model proportions/probability

$$E(x) = \frac{a}{a+b}, \quad E(x^2) = \frac{(a)(a+1)}{(a+b)(a+b+1)}$$

$$\text{Var}(x) = \frac{ab}{(a+b)^2(a+b+1)}$$

k -th moment

$$E(x^k) = \int_0^1 x^k p(x) dx = \int_0^1 x^k \frac{\Gamma(a+b)}{\Gamma(a) \Gamma(b)} x^{a-1} (1-x)^{b-1} dx$$

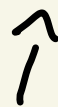
$$= \frac{\Gamma(a+b)}{\Gamma(a) \Gamma(b)} \cdot \int_0^1 x^{a+k-1} (1-x)^{b-1} dx$$

$$= \frac{\Gamma(a+b)}{\Gamma(a) \Gamma(b)} \cdot \frac{\Gamma(a+b) \Gamma(b)}{\Gamma(a+k+b)} \int_0^1 \frac{\Gamma(a+k+b)}{\Gamma(a+k) \Gamma(b)} x^{a+k-1} (1-x)^{b-1} dx$$

$$= \frac{\Gamma(a+b) \Gamma(a+k)}{\Gamma(a) \Gamma(a+k+b)} \cdot \left| \begin{array}{l} \text{PDF, integral of PDF is always 1} \end{array} \right.$$

$\frac{a}{a+b}$

for
1st moment



$$\hookrightarrow \frac{(a+b-1)! \cdot (a+k-1)!}{(a-1)! \cdot (a+b+k-1)!} = \text{for } k=1, \frac{(a+b-1)! \cdot a!}{(a-1)! \cdot (a+b)!}$$

Moments

What's a moment? \rightarrow Has to do with Expectation

$E(x)$ is first moment

$E(x^2)$ is second moment

$E(x^3)$ is third moment

$E(x^k)$ is k th moment

k th central moment is $E((x - E(x))^k)$

$$M_x(t) = E(e^{tx}) = \int_{-\infty}^{\infty} e^{tx} \underbrace{f(x)}_{\uparrow \text{PDF}} dx$$

Expectation of values may exist
for MGF to exist

$$E(x^n) = \frac{d^n}{dt^n} M_x(t) \Big|_{t=0}$$

n th order moment

Why is MGF defined as $E(e^{tx})$?

$$e^x = 1 + x + \frac{x^2}{2!} + \frac{x^3}{3!} \dots$$

$$e^{tx} = 1 + tx + \frac{t^2 x^2}{2!} + \frac{t^3 x^3}{3!} \dots$$

$$E(e^{tx}) = 1 + t E(x) + \frac{t^2 E(x^2)}{2!} + \dots$$

$$\left. \frac{d}{dt} E(e^{tx}) \right|_{t=0} = 0 + E(x) + t E(x^2) + \dots \Big|_{t=0}$$

Since t is evaluated at 0, $\frac{d}{dt} E(e^{tx}) = 0 + E(x) + 0 + \dots$

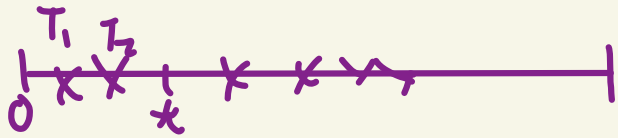
$$\left. \frac{d^2}{dt^2} E(e^{tx}) \right|_{t=0} = 0 + 0 + E(x^2) + 0 \dots$$

Therefore $E(e^{tx})$ gives moment generating function

and deriving MGF gives the k th order moment

Poisson process)

Gamma-expo connection



$N_x = \# \text{ emails up to } x \sim \text{Pois}(\lambda x)$

Assume arrivals in disjoint intervals are independent

We draw x everytime we get an email

Let T_1 be time of first email

$$P(T_1 > x) = P(N_x = 0) = e^{-\lambda x}$$

T_1 is exponential, T_2 is also exponential

fish?

why?

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what is it?

Poisson Distribution:

Suppose we are counting # of occurrences in a given unit of time, distance, area/volume

Example: # of car accidents in a day

The # is a RV that may or may not follow the Poisson Distribution

Suppose Events are independent

The probability does not change through time

Then X , the # of events, has the Poisson distribution

$$P(X=x) = \frac{\lambda^x e^{-\lambda}}{x!} \quad \left. \vphantom{\frac{\lambda^x e^{-\lambda}}{x!}} \right\} \begin{array}{l} \text{Probability mass} \\ \text{function} \end{array}$$

or $P(x)$

$\mu = \lambda$ (mean of the Poisson dist)

$\sigma^2 = \lambda$ (Standard deviation squared = variance)

One nanogram of Plutonium has an average of 2.3 decays per second, and # of decays follows a Poisson distribution

Let X be # decays in 2 seconds find $P(X=3)$

So: $\lambda = 2 \cdot 2.3 = 4.6$

$$P(X=3) = \frac{\lambda^x e^{-\lambda}}{x!} = \frac{4.6^3 e^{-4.6}}{3!} = 0.163$$

"Truncated" Poisson

Recall Poisson $(X=x) \approx P(x) = \frac{\lambda^x \cdot e^{-\lambda}}{x!}$ where $\lambda(\text{lambda}) = EV(f)$

Say we had a party of n people. Find the probability that there is a group of 3 people with the same birthday. Assume a Poisson distribution.

By Poisson: The first moment $EV = \lambda$, $Var = \lambda$

Lambda is Expected value of getting a group of 3 people w/same day

$$\binom{n}{3} \text{ people} \cdot \frac{1}{1} \cdot \frac{1}{365} \cdot \frac{1}{365} \approx \binom{n}{3} \cdot \frac{1}{365^2} = \lambda$$

Probability: We want to find $P(X=x | x > 0) = \frac{P(x > 0 | X=x) \cdot P(X=x)}{P(x > 0)}$

$$\frac{1 \cdot P(X=x)}{P(x > 0)} \quad \begin{array}{l} \swarrow \text{Poisson} \\ \nwarrow \text{change to comp} \end{array}$$

$$\frac{P(X=x)}{1 - P(X=0)} = \frac{\frac{\lambda^x e^{-\lambda}}{x!}}{1 - \frac{\lambda^0 e^{-\lambda}}{0!}}$$

$$= \frac{\frac{\lambda^x e^{-\lambda}}{x!}}{1 - e^{-\lambda}}$$

$$= \frac{\lambda^x e^{-\lambda}}{x!(1 - e^{-\lambda})} = \frac{\lambda^x}{x!(e^{\lambda} - 1)}$$

Poisson Max Likelihood Estimator (MLE)

Suppose $X_i = \#$ of textbooks a person buys

$$X_1 \dots X_{50}, \sum X_i = 150$$

λ is mean, $X_i \sim \text{Poisson}(\lambda)$

$f(X_i | \lambda)$ is the PDF for a given X_i

$$\text{Recall } \text{Poisson}(x) = \frac{\lambda^x \cdot e^{-\lambda}}{x!}$$

so $f(X_i | \lambda) \rightarrow$ pdf of X_i given λ

$$\frac{\lambda^{X_i} \cdot e^{-\lambda}}{X_i!}$$

