

Bernoulli distribution

The Bernoulli distribution for a single binary random variable X with state $x \in \{0,1\}$.

It is governed by a single continuous parameter $\mu \in [0,1]$ that represents the probability of $X = 1$.

The distribution $Ber(\mu)$ is defined as :

$$P(x|\mu) = \mu^x(1 - \mu)^{1-x}, x \in \{0,1\},$$

$$E[x] = \mu,$$

$$V[x] = \mu(1 - \mu),$$

Where $E[x]$ and $V[x]$ are the mean and variance of binary random variable.

Exponential Family

$$f(x|\theta) = h(x)\exp(\eta(\theta)T(x) - A(\theta))$$

Now , take the Bernoulli distribution series:

$$P(x|\mu) = \mu^x(1 - \mu)^{1-x}$$

Note:

$$f(x|\mu) = \begin{cases} \mu, & \text{if } x = 1 \\ 1 - \mu, & \text{if } x = 0 \end{cases}$$

i. e. μ when $x = 1$ i. e. success and $1 - \mu$, if $x = 0$ i. e. failures and $P(x|\mu)$ where P stands for probability.

Now let us expand the Bernoulli's distribution,

$$\begin{aligned}
 P(x|\mu) &= \mu^x(1 - \mu)^{1-x} \\
 &\Rightarrow \exp[\log(\mu^x(1 - \mu)^{1-x})] \\
 &\Rightarrow \exp[x\log(\mu) + (1 - x)\log(1 - \mu)] \\
 &\Rightarrow \exp[x\log(\mu) + \log(1 - \mu) - x\log(1 - \mu)] \\
 &\Rightarrow \exp[x\log(\mu) - x\log(1 - \mu) + \log(1 - \mu)] \\
 &\Rightarrow \exp\left[x\log\left(\frac{\mu}{1 - \mu}\right) + \log(1 - \mu)\right] - eqn(i)
 \end{aligned}$$

Note , the above eqn(i) can be interpreted in different ways:

i.e. if $\mu = P$, then:

$$\exp\left[x\log\left(\frac{p}{1 - p}\right) + \log(1 - p)\right]$$

And during that time Bernoulli's distribution becomes:

$$P(x|p) = p^x(1 - p)^{1-x}$$

And,

$$f(x|p) = \begin{cases} p, & \text{if } x = 1 \\ 1 - p, & \text{if } x = 0 \end{cases}$$

Therefore, it depends on notable variable we use in the equation.

Now we can write the above equation as:

$$\exp \left[x \log \left(\frac{\mu}{1 - \mu} \right) - (-\log(1 - \mu)) \right] - eqn(ii)$$

Now,

$$h(x) = 1,$$

$$\theta = \log \left(\frac{\mu}{1 - \mu} \right),$$

$$\theta(x) = 1 \times x = x,$$

$$A(\theta) = -\log(1 - \mu)$$

The above interpretation is based on exponential family,

$$f(x|\theta) = h(x) \exp(\eta(\theta)T(x) - A(\theta))$$

i.e.

$$h(x) = 1,$$

$$\eta(\theta) = \log\left(\frac{\mu}{1-\mu}\right),$$

$$T(x) = 1 \times x = x,$$

$$A(\theta) = -\log(1 - \mu)$$

Now, note the relationship between μ and θ are invertible:

$$\mu = \frac{1}{1 + \exp^{-\theta}} - eqn(i)$$

And,

$$\exp^{-\theta} = \frac{1 - \mu}{\mu} - eqn(a)$$

Or,

$$\mu = \frac{1}{1 + \frac{1}{\exp^{\theta}}}$$

$$\Rightarrow \mu = \frac{1}{\frac{\exp^{\theta} + 1}{\exp^{\theta}}}$$

$$\Rightarrow \mu = \frac{\exp^{\theta}}{\exp^{\theta} + 1} - eqn(ii)$$

Similarly:

$$\exp^{\theta} = \frac{\mu}{1 - \mu} - eqn(b)$$

From the above equation(i) we get :

$$\begin{aligned}\mu &= \frac{1}{1 + \exp^{-\theta}} \\ \Rightarrow \mu(1 + \exp^{-\theta}) &= 1 \\ \Rightarrow (1 + \exp^{-\theta}) &= \frac{1}{\mu} \\ \Rightarrow (\exp^{-\theta}) &= \frac{1}{\mu} - 1 \\ \Rightarrow (\exp^{-\theta}) &= \frac{1 - \mu}{\mu} \\ \Rightarrow \mu(\exp^{-\theta}) &= 1 - \mu\end{aligned}$$

we know ,

$$\mu = \frac{1}{1 + \exp^{-\theta}}$$

Therefore,

$$\begin{aligned}\Rightarrow \frac{1}{1 + \exp^{-\theta}} \times (\exp^{-\theta}) &= 1 - \mu \\ \Rightarrow \frac{\exp^{-\theta}}{1 + \exp^{-\theta}} &= 1 - \mu\end{aligned}$$

$$\Rightarrow \frac{\frac{1}{\exp^{\theta}}}{1 + \frac{1}{\exp^{\theta}}} = 1 - \mu$$

$$\Rightarrow \frac{\frac{1}{\exp^{\theta}}}{\frac{\exp^{\theta} + 1}{\exp^{\theta}}} = 1 - \mu$$

$$\Rightarrow \frac{1}{\exp^{\theta}} \times \frac{\exp^{\theta}}{\exp^{\theta} + 1} = 1 - \mu$$

$$\Rightarrow \frac{1}{\exp^{\theta} + 1} = 1 - \mu$$

Therefore we got,

$$\frac{1}{\exp^{\theta} + 1} = 1 - \mu$$

Putting it in:

$$\exp \left[x \log \left(\frac{\mu}{1 - \mu} \right) - (-\log(1 - \mu)) \right]$$

we get,

$$\begin{aligned}
& \Rightarrow \exp \left[x \log \left(\frac{\mu}{1-\mu} \right) - \left(- \left(\log \left(\frac{1}{\exp^{\theta} + 1} \right) \right) \right) \right] \\
& \Rightarrow \exp \left[x \log \left(\frac{\mu}{1-\mu} \right) \right. \\
& \quad \left. - \left(- \left(\log(1) - \log(\exp^{\theta} + 1) \right) \right) \right] \\
& \Rightarrow \exp \left[x \log \left(\frac{\mu}{1-\mu} \right) - \left(- (0 - \log(\exp^{\theta} + 1)) \right) \right] \\
& \Rightarrow \exp \left[x \log \left(\frac{\mu}{1-\mu} \right) - \left(- (-\log(\exp^{\theta} + 1)) \right) \right] \\
& \Rightarrow \exp \left[x \log \left(\frac{\mu}{1-\mu} \right) - \left(+\log(\exp^{\theta} + 1) \right) \right]
\end{aligned}$$

In Canonical Form

$$f(x|\theta) = h(x) \exp(\eta(\theta)T(x) - A(\theta))$$

$$h(x) = 1,$$

$$\eta(\theta) = \log \left(\frac{\mu}{1-\mu} \right),$$

$$T(x) = x,$$

$$A(\theta) = \log(\exp^{\theta} + 1)$$

Mean

$$\begin{aligned} E(T(x)) &= E[x] \\ &= \frac{\partial}{\partial \theta} A(\theta) \\ &= \frac{\partial}{\partial \theta} \times \log(\exp^{\theta} + 1) \\ &= \frac{1}{1 + \exp^{\theta}} \times \frac{\partial}{\partial \theta} (1 + \exp^{\theta}) \\ &= \frac{1}{1 + \exp^{\theta}} \times (0 + \frac{\partial}{\partial \theta} \times \exp^{\theta}) \end{aligned}$$

Note, '**exp**' represents as '**e**' and '**exp**' stands for '**exponent**'.

Re-writing the equation:

$$\begin{aligned} &= \frac{1}{1 + e^{\theta}} \times (0 + \frac{\partial}{\partial \theta} \times e^{\theta}) \\ &= \frac{1}{1 + e^{\theta}} \times (0 + e^{\theta}) \\ &= \frac{1}{1 + e^{\theta}} \times (e^{\theta}) \\ &= \frac{e^{\theta}}{1 + e^{\theta}} \end{aligned}$$

We know,

$$\theta = \log\left(\frac{\mu}{1-\mu}\right),$$

Or,

$$\eta(\theta) = \log\left(\frac{\mu}{1-\mu}\right)$$

Hence putting the value of θ in the equation:

$$= \frac{e^{\left(\log\left(\frac{\mu}{1-\mu}\right)\right)}}{1 + e^{\left(\log\left(\frac{\mu}{1-\mu}\right)\right)}}$$

We can also represent it as:

$$= \frac{e^{\left(\log_e\left(\frac{\mu}{1-\mu}\right)\right)}}{1 + e^{\left(\log_e\left(\frac{\mu}{1-\mu}\right)\right)}}$$

we know, $a^{\log_a b} = b$

$$= \frac{\frac{\mu}{1-\mu}}{1 + \frac{\mu}{1-\mu}}$$

$$= \frac{\frac{\mu}{1-\mu}}{\frac{1-\mu+\mu}{1-\mu}}$$

$$= \frac{\frac{\mu}{1-\mu}}{\frac{1}{1-\mu}}$$

$$= \frac{\mu}{1-\mu} \times \frac{1-\mu}{1}$$

$$= \mu$$

Hence, Mean = $E[x] = \mu$

Variance

$$Var(T(x)) = Var[x]$$

We know variance = σ^2

$$Var[x] = \frac{\partial^2}{\partial \theta^2} \times A(\theta)$$

$$= \frac{\partial^2}{\partial \theta^2} \times \left(\log(\exp^\theta + 1) \right)$$

$$= \frac{\partial}{\partial \theta} \times \frac{\partial}{\partial \theta} \times (\log(\exp^\theta + 1))$$

We know $\frac{\partial}{\partial \theta} \times (\log(\exp^\theta + 1)) = \frac{e^\theta}{1 + e^\theta}$

as we got it while doing calculation of mean, therefore:

$$= \frac{\partial}{\partial \theta} \times \left(\frac{e^\theta}{1 + e^\theta} \right)$$

We know the division calculation of differential

calculus: $\frac{\partial}{\partial x} \times \frac{f(x)}{g(x)} = \frac{g(x) \times \frac{\partial}{\partial x} f(x) - f(x) \times \frac{\partial}{\partial x} g(x)}{(g(x))^2}$

$$= \frac{(1 + e^\theta) \times \frac{\partial}{\partial \theta} \times e^\theta - \left(e^\theta \times \frac{\partial}{\partial \theta} \times (1 + e^\theta) \right)}{(1 + e^\theta)^2}$$

$$= \frac{(1 + e^\theta) \times e^\theta - \left(e^\theta \times \left(0 + \frac{\partial}{\partial \theta} \times e^\theta \right) \right)}{(1 + e^\theta)^2}$$

$$= \frac{e^\theta + e^{2\theta} - \left(e^\theta \times (0 + e^\theta) \right)}{(1 + e^\theta)^2}$$

$$= \frac{e^{\theta} + e^{2\theta} - e^{2\theta}}{(1 + e^{\theta})^2}$$

$$= \frac{e^{\theta}}{(1 + e^{\theta})^2}$$

We know,

$$\theta = \log\left(\frac{\mu}{1 - \mu}\right),$$

Therefore,

$$= \frac{e^{\log\left(\frac{\mu}{1 - \mu}\right)}}{\left(1 + e^{\log\left(\frac{\mu}{1 - \mu}\right)}\right)^2}$$

We can write it as:

$$= \frac{e^{\log_e\left(\frac{\mu}{1 - \mu}\right)}}{\left(1 + e^{\log_e\left(\frac{\mu}{1 - \mu}\right)}\right)^2}$$

$$\text{we know, } a^{\log_a b} = b$$

$$= \frac{\frac{\mu}{1-\mu}}{\left(1 + \frac{\mu}{1-\mu}\right)^2}$$

$$= \frac{\frac{\mu}{1-\mu}}{\left(\frac{1-\mu+\mu}{1-\mu}\right)^2}$$

$$= \frac{\frac{\mu}{1-\mu}}{\left(\frac{1}{1-\mu}\right)^2}$$

$$= \frac{\mu}{1-\mu} \times (1-\mu)^2$$

$$= \mu \times (1-\mu)$$

Therefore, **Variance** = $Var[x] = \mu \times (1-\mu)$

Now if we represent μ as p

Then,

$$\text{Mean} = E[x] = \mathbf{p}$$

$$\text{Var}[x] = \mathbf{p} \times (\mathbf{1} - \mathbf{p})$$