

02_Exercise3_MaxL_

April 22, 2018

GIVEN: Samples 0,1,0,0,1,0 from a binomial distribution which has the form: $P(x=0)=(1-\mu)$, $P(x=1)=\mu$

REQUESTED: What is the maximum likelihood estimate of μ .Hint: you can use SymPy to compute the derivities symbolically

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In [2]: import sympy as sp
import numpy as np
import matplotlib.pyplot as plt
sp.init_printing(use_latex=True)
```

Maximum likelihood estimation:

$$L(p \mid y, N) = \binom{N}{y} \cdot p^y \cdot (1-p)^{N-y}$$

where,

p = Probability of success N = Number of trials y = Number of successes in N trials

Since $\binom{N}{y}$ is a constant, this has been ignored.

Taking log on both the sides,

$$\ln L(p \mid y, N) = y \cdot \ln(p) + (N - y) \ln(1 - p)$$

$$\frac{\partial [\ln L(p \mid y, N)]}{\partial p} = \frac{y}{p} - \frac{N - y}{1 - p} = 0$$

$$\frac{y}{p} = \frac{(N - y)}{(1 - p)}$$

$$y(1 - p) = p(N - y)$$

$$y - yp = pN - yp$$

$$pN = y - yp + yp$$

$$p = \frac{y}{N}$$

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In [3]: samples = [0,1,0,0,1,0]
success = sum(samples)
failures = len(samples)-success
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In [4]: mu = sp.Symbol('mu')
eq = mu**success * (1-mu) **failures
eq_prime = sp.diff(eq,mu)
print eq_prime
sp.solve(eq_prime)
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-4*mu**2*(-mu + 1)**3 + 2*mu*(-mu + 1)**4
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Out [4]:
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$$\left[0, \frac{1}{3}, 1\right]$$

Reference : http://www.montana.edu/rotella/documents/502/binom_like.pdf