## Slaying the Mythical p

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Binomial distribution:

$$P(F_h|p_h, F) = \binom{F}{F_h} p_h^{F_h} (1 - p_h)^{F - F_h}$$
(1)

Statistician:

$$F = 250, F_h = 140, p_h = 0.5, F - F_h = F_t = 110$$
 (2)

$$P(F_h = 140 | p_h = 0.5, N = 250) = {250 \choose 140} 0.5^{140} (1 - 0.5)^{110} \simeq 10^{-21}$$
 (3)

$$P(F_h \ge 140 | p_h = 0.5, N = 250) = \sum_{F_h = 140}^{250} {250 \choose F_h} 0.5^{250} = 0.0332$$
 (4)

 $\mathcal{H}_0$ : Coin is fair  $\mathcal{H}_1$ : Coin is biased

D: Data of 140 heads in 250 spins

 $F_h$ : 140 (frequency of heads)  $F_t$ : 110 (frequency of tails) F: 250 (number of coin spins)

 $p_h$ : Probability of heads Getting model comparison:

$$P(\mathcal{H}_0|D) = \frac{P(D|\mathcal{H}_0)P(\mathcal{H}_0)}{P(D)}$$
 (5)

$$P(\mathcal{H}_1|D) = \frac{P(D|\mathcal{H}_1)P(\mathcal{H}_1)}{P(D)} \tag{6}$$

$$P(\mathcal{H}_0) = P(\mathcal{H}_1) = \frac{1}{2} \tag{7}$$

$$\frac{P(\mathcal{H}_1|D)}{P(\mathcal{H}_0|D)} = \frac{P(D|\mathcal{H}_1)}{P(D|\mathcal{H}_0)} \tag{8}$$

Done getting model comparison.

Getting  $P(D|\mathcal{H}_1)$ :

$$P(D|\mathcal{H}_1) = \int_0^1 \mathrm{d}p_h P(D|p_h, \mathcal{H}_1) P(p_h|\mathcal{H}_1) \tag{9}$$

$$P(D|p_h, \mathcal{H}_1) = p_h^{F_h} (1 - p_h)^{F_t}$$
(10)

$$P(p_h|\mathcal{H}_1) = 1 \tag{11}$$

$$P(D|\mathcal{H}_1) = \int_0^1 \mathrm{d}p_h p_h^{F_h} (1 - p_h)^{F_t} \tag{12}$$

$$B(x,y) = \int_0^1 dt t^{x-1} (1-t)^{y-1} = \frac{\Gamma(x)\Gamma(y)}{\Gamma(x+y)}$$
 (13)

$$P(D|\mathcal{H}_1) = B(F_h + 1, F_t + 1) = \frac{\Gamma(F_h + 1)\Gamma(F_t + 1)}{\Gamma(F_h + F_t + 2)}$$
(14)

$$\Gamma(x) = (x-1)! \tag{15}$$

$$P(D|\mathcal{H}_1) = \frac{(F_h + 1)!(F_t + 1)!}{(F_h + F_t + 1)!} = \frac{F_h!F_t!}{(F + 1)!}$$
(16)

End of getting  $P(D|\mathcal{H}_1)$ . Getting  $P(D|\mathcal{H}_0)$ :

$$p_f = \frac{1}{2} \tag{17}$$

$$P(D|\mathcal{H}_0) = \left(\frac{1}{2}\right)^{F_h} \left(1 - \left(\frac{1}{2}\right)\right)^{F_t} = \left(\frac{1}{2}\right)^F \tag{18}$$

Done getting  $P(D|\mathcal{H}_0)$ .

Computing model comparison:

$$\frac{P(\mathcal{H}_1|D)}{P(\mathcal{H}_0|D)} = \frac{\frac{F_h!F_t!}{(F+1)!}}{\left(\frac{1}{2}\right)^F} = \frac{\frac{140!110!}{251!}}{\left(\frac{1}{2}\right)^{250}} \simeq 0.48$$
 (19)

End computing model comparison.

Getting Beta prior:

$$P(p_h|\mathcal{H}_1,\alpha) = \frac{1}{Z(\alpha)} p_h^{\alpha-1} (1 - p_h)^{\alpha-1}$$
 (20)

$$Z(\alpha) = \frac{\Gamma(\alpha)^2}{\Gamma(2\alpha)} \tag{21}$$

$$P(D|\mathcal{H}_1) = \int_0^1 \mathrm{d}p_h P(D|p_h, \mathcal{H}_1) P(p_h|\mathcal{H}_1)$$
 (22)

$$P(D|\mathcal{H}_1) = \int_0^1 \mathrm{d}p_h p_h^{F_h} (1 - p_h)^{F_t} \frac{1}{Z(\alpha)} p_h^{\alpha - 1} (1 - p_h)^{\alpha - 1}$$
 (23)

$$P(D|\mathcal{H}_1) = \frac{1}{Z(\alpha)} \int_0^1 dp_h p_h^{F_h + \alpha - 1} (1 - p_h)^{F_t + \alpha - 1}$$
 (24)

$$B(F_h + \alpha - 1 + 1, F_t + \alpha - 1 + 1) = \frac{\Gamma(F_h + \alpha)\Gamma(F_t + \alpha)}{\Gamma(F_t + 2\alpha)}$$
(25)

$$P(D|\mathcal{H}_1) = \frac{1}{Z(\alpha)} \frac{\Gamma(F_h + \alpha)\Gamma(F_t + \alpha)}{\Gamma(F + 2\alpha)}$$
(26)

$$P(D|\mathcal{H}_1) = \frac{\Gamma(2\alpha)\Gamma(F_h + \alpha)\Gamma(F_t + \alpha)}{\Gamma(\alpha)^2\Gamma(F + 2\alpha)}$$
(27)

$$\frac{P(\mathcal{H}_1|D)}{P(\mathcal{H}_0|D)} = \frac{\Gamma(2\alpha)\Gamma(F_h + \alpha)\Gamma(F_t + \alpha)2^{250}}{\Gamma(\alpha)^2\Gamma(F + 2\alpha)}$$
(28)

End getting Beta prior.

Getting best possible prior:

$$\frac{P(\mathcal{H}_1|D)}{P(\mathcal{H}_0|D)} = f^{F_h}(1-f)^{F_t}2^F \simeq 6.1 \tag{29}$$