

$$\text{Posterior} \downarrow \quad \text{Likelihood} \downarrow \quad \text{Prior} \downarrow \\ P(H|E) = \frac{P(E|H) \cdot P(H)}{P(E)} \quad \text{Evidence}$$

Beta distribution.

$$P(\theta|a,b) = \frac{\theta^{a-1}(1-\theta)^{b-1}}{\text{Beta}(a,b)} \propto \theta^{a-1}(1-\theta)^{b-1} \quad \theta \in [0,1]$$

Beta(a,b) ← normalization term.

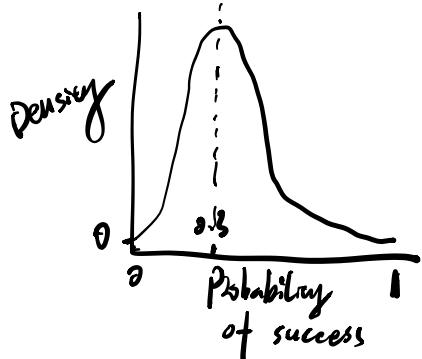
we use the Beta distribution to specify prior knowledge about parameter θ .

$$E(\theta) = \frac{a}{a+b}$$

Suppose our conversion rate = 0.3, we can choose any arbitrary number for a and b such that $E(\theta) = \frac{a}{a+b} = 0.3$

Beta(3, 7)

ex: ($a=3, b=7$) ($a=6, b=14$)



Smoothing: bias and Variance $P = \frac{s+\alpha}{s+\alpha+\beta}$

Smoothing is similar to regularization, where we induce a small bias on our estimate in order to reduce the variance.

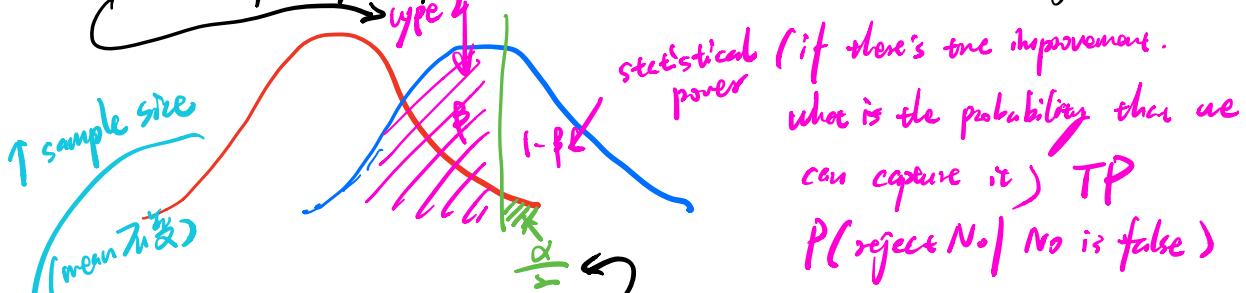
A/B Test

Frequent Metrics: Click thru rate, conversion rate.

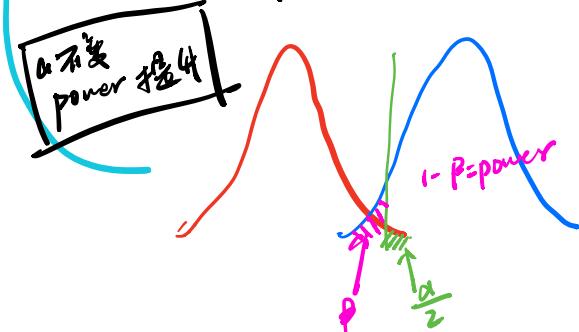
Renewal rate, Average retention.

若无2个问题

实验确实有提升，实验结果不能证明提升。 (False Negative)



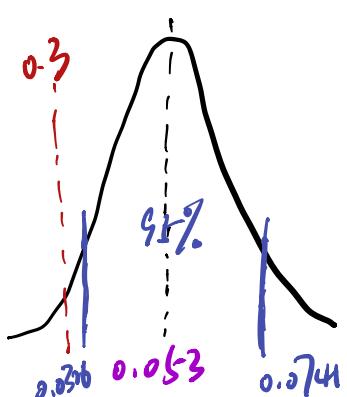
实验并没有提升，但是实验结果证明有提升。 (False Positive)



Sample size increases
↓
se decreases
power increases (which means that
80% of the time we will capture the
improvement if there's actually
a improvement.)

power = 0.8

Test Result:



sample
N_a: 2000
N_b: 1943

conversion
 $\bar{x}_a: 200$
 $\bar{x}_b: 298$

CI: 95%

Practical significance: 0.03 (expected improvement)

$$P_a = 0.1, P_b = 0.153$$

$$d = 0.153 - 0.1 = 0.053$$

$$P_{pool} = (298 + 200) / (2000 + 1943) = 0.1263$$

$P_{pool} = \frac{1}{N_1 + N_2}$
 $SE = \sqrt{P_{pool}(1-P_{pool})\left(\frac{1}{N_1} + \frac{1}{N_2}\right)} = 0.0106$
 $M = Z \times SE = 1.96 \times 0.0106 = 0.0207$
 $CI: (d - M, d + M) = (0.0326, 0.0741)$

our improvement could
 be lower than our expected
 improvement. So
 we are sure that
 the improvement is actually
 desired.

If lower bound < 0.1, increase sample size
 (run longer to collect more samples)
 law of large numbers: the distribution will converge
 to the true mean.

