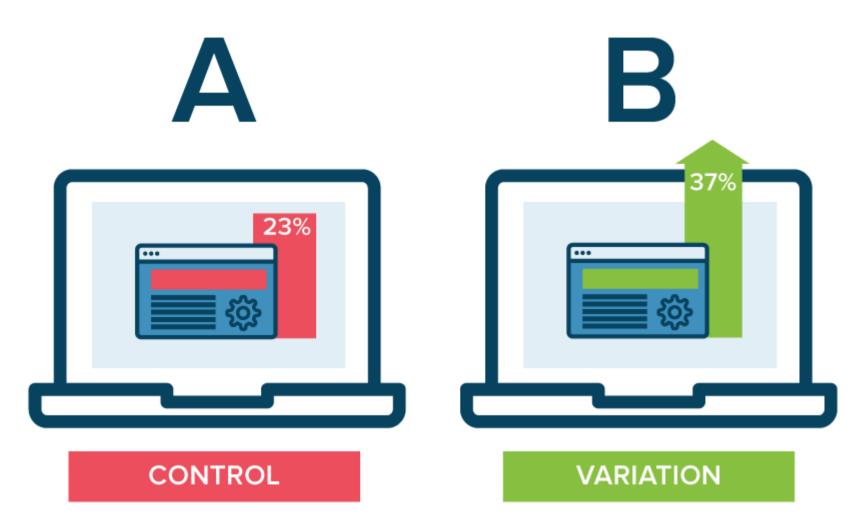
Bayesian A/B Testing



Bayes' Law for A/B testing

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 - What are some problems with t-tests for large amounts of data like web traffic?

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- We can do a t-test for A/B testing things like website design (e.g. red vs blue theme)
 - What are some problems with t-tests for large amounts of data like web traffic?
 - Everything looks significant because of the amount of data. t-tests were made for small data
 - Also assumes t-distributions (similar to normal distributions, but with slightly fatter tails)
 - t-stat inversely proportional to number of samples
 - Check out t.test.example.R in worldclass

$$t = \frac{\overline{x_1} - \overline{x_2}}{\sqrt{\frac{{S_1}^2}{N_1} + \frac{{S_2}^2}{N_2}}}$$

http://rpsychologist.com/d3/NHST/

https://www.optimizely.com/optimization-glossary/ab-testing

- gamma = (n 1)!
- Mean = alpha / (alpha + beta)
- In R, dbeta(alpha, Beta)

$$f(x; \alpha, \beta) = \frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha)\Gamma(\beta)} x^{\alpha - 1} (1 - x)^{\beta - 1}$$
$$= \frac{1}{B(\alpha, \beta)} x^{\alpha - 1} (1 - x)^{\beta - 1}$$

- Beta distribution good for A/B testing (binomial – 0 or 1 results)
- distribution of probability after S successes and F failures is given by Beta(S+1, F+1)
 - So assuming a 50/50 chance with 2 trials, we have Beta(2, 2). A completely blank prior would be Beta(1, 1), which would mean no trials

Parameters:
$$\alpha > 0$$
 and $\beta > 0$

$$f(x) = \frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha)\Gamma(\alpha)} x^{\alpha - 1} (1 - x)^{\beta - 1} \text{ for } 0 \le x \le 1$$

$$\mu = \frac{\alpha}{\alpha + \beta}$$

$$\sigma^2 = \frac{\alpha\beta}{(\alpha + \beta)^2 (\alpha + \beta + 1)}$$

$$x_{\text{mode}} = \frac{\alpha - 1}{\alpha + \beta - 2}$$

- We can add the alpha/beta parameters from a prior and from data to get a posterior distribution
 - e.g. say we expect 30% conversion rate from a webpage for whatever reason, we could use beta(4, 8) as our prior distribution
 - Say we then find that 50 out of 450 people sign up, our evidence is showing a distribution of beta(51, 451)
 - We add the alpha/beta to get our posterior distribution, beta(55, 459). Then this becomes our prior the next time we re-evaluate the results

- If we have one version of a webpage (A) that gets 50/450 to sign up, and B gets 37/425, which one is better and by how much?
- Add the data to our prior again, then randomly sample the two Beta distributions, divide the results of A by the results of B to get a distribution of how much better A is than B
- See bayes.ab.test.example.R file in worldclass