

# Introduction to Probability

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- There are other ways of thinking about probability, but we'll stick with this one

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  - The probability of disjoint (mutually exclusive) sets is equal to their sums

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  - $\pi$  represents the probability of success

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- PMF:

$$\binom{n}{k} p^k (1 - p)^{n-k}$$



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  - Kinda. . .



# Continuous distributions

```
ggplot(data.frame(x = c(-5, 5)), aes(x)) +  
  stat_function(fun = dnorm)
```

