

Materials: <http://huppenkothen.org/bayesian-statistics-tutorial/>



Bayesian Statistics

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**All models are wrong,
but some are useful**

— George Box

Nature is **complex** ...

... and so is our **data collection!**



Quick poll!



Part 1: Probabilities

What snacks will we have for coffee break today?



What snacks will we have for coffee
break today?

$p(\text{🍪})$

$p(\text{🥐})$

Basic rule of probability (1)

$$0 \leq p(\text{🍪}) \leq 1$$

If $p(\text{🍪}) = 0.3$, what is $p(\text{🥐})$?

- a) 0.3
- b) 0.5
- c) 0.01
- d) 0.7

$$p(\text{🍪}) + p(\text{🥐}) = 1$$

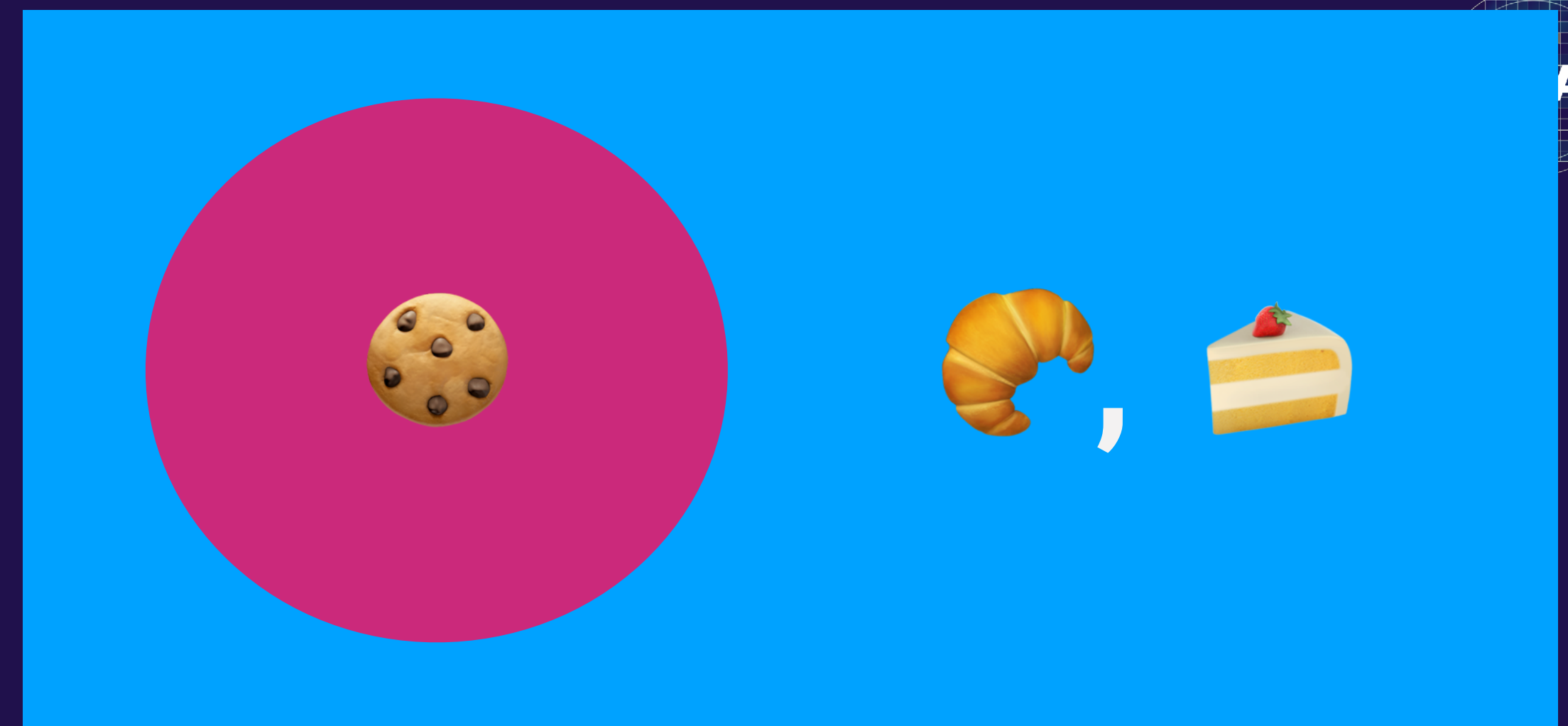
$$\sum_i P_i(x) = 1$$



$$p(\text{🍪}) + p(\text{🥐}) + p(\text{🍰}) = 1$$

$$p(\text{🥐}) + p(\text{🍰}) = p(\text{not 🍪})$$

$$p(\text{🍪}) = 1 - p(\text{not 🍪})$$



“complement”

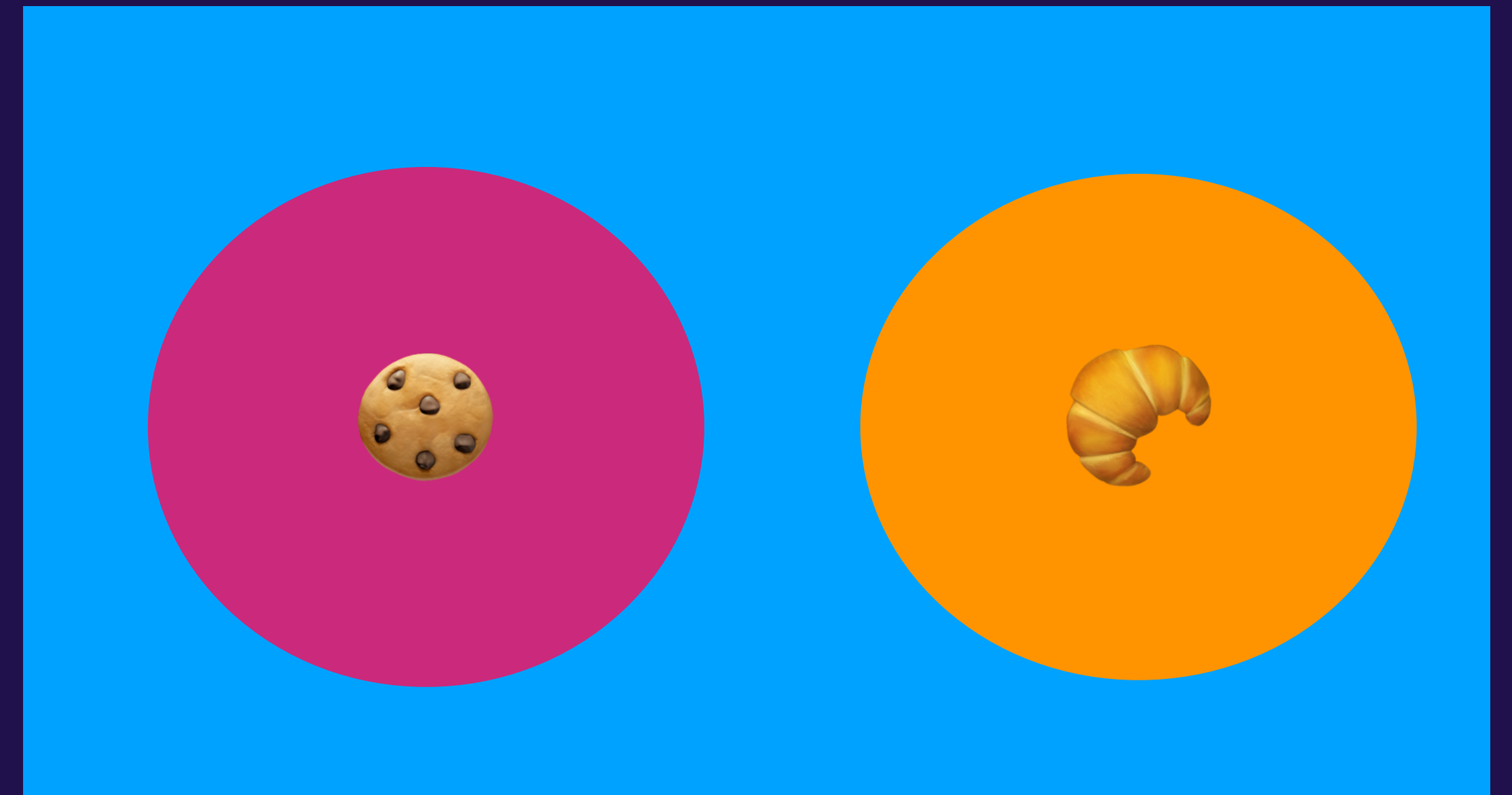
$$p(\text{🍪} \cap \text{🥐}) = \text{"p(🍪 and 🥐)"}"$$

$$p(\text{🍪} \cup \text{🥐}) = \text{"p(🍪 or 🥐)"}"$$

🍪 , 🥐 are independent:

$$p(\text{🍪} \cap \text{🥐}) = 0$$

$$p(\text{🍪} \cup \text{🥐}) = p(\text{🍪}) + p(\text{🥐})$$



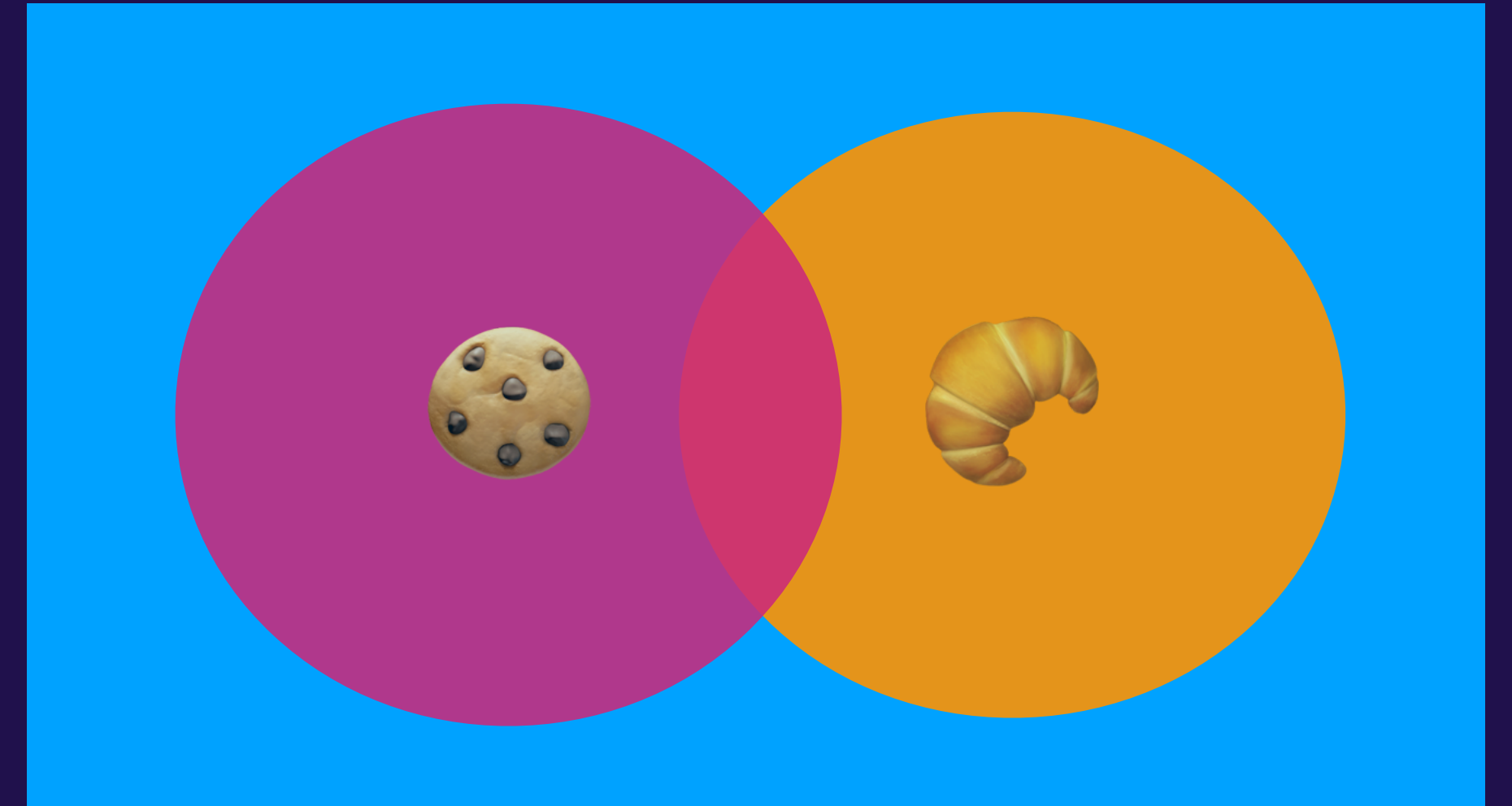
$$p(\text{🍪} \cap \text{🥐}) = \text{"p(🍪 and 🥐)"}'$$

$$p(\text{🍪} \cup \text{🥐}) = \text{"p(🍪 or 🥐)"}'$$

🍪 , 🥐 are **not** independent:

$$p(\text{🍪} \cap \text{🥐}) = x$$

$$p(\text{🍪} \cup \text{🥐}) = p(\text{🍪}) + p(\text{🥐}) - p(\text{🍪} \cap \text{🥐})$$





Let's add another category

Is there a break with snacks in the morning or the afternoon?

$p(\text{🌅})$ = there are snacks in the morning

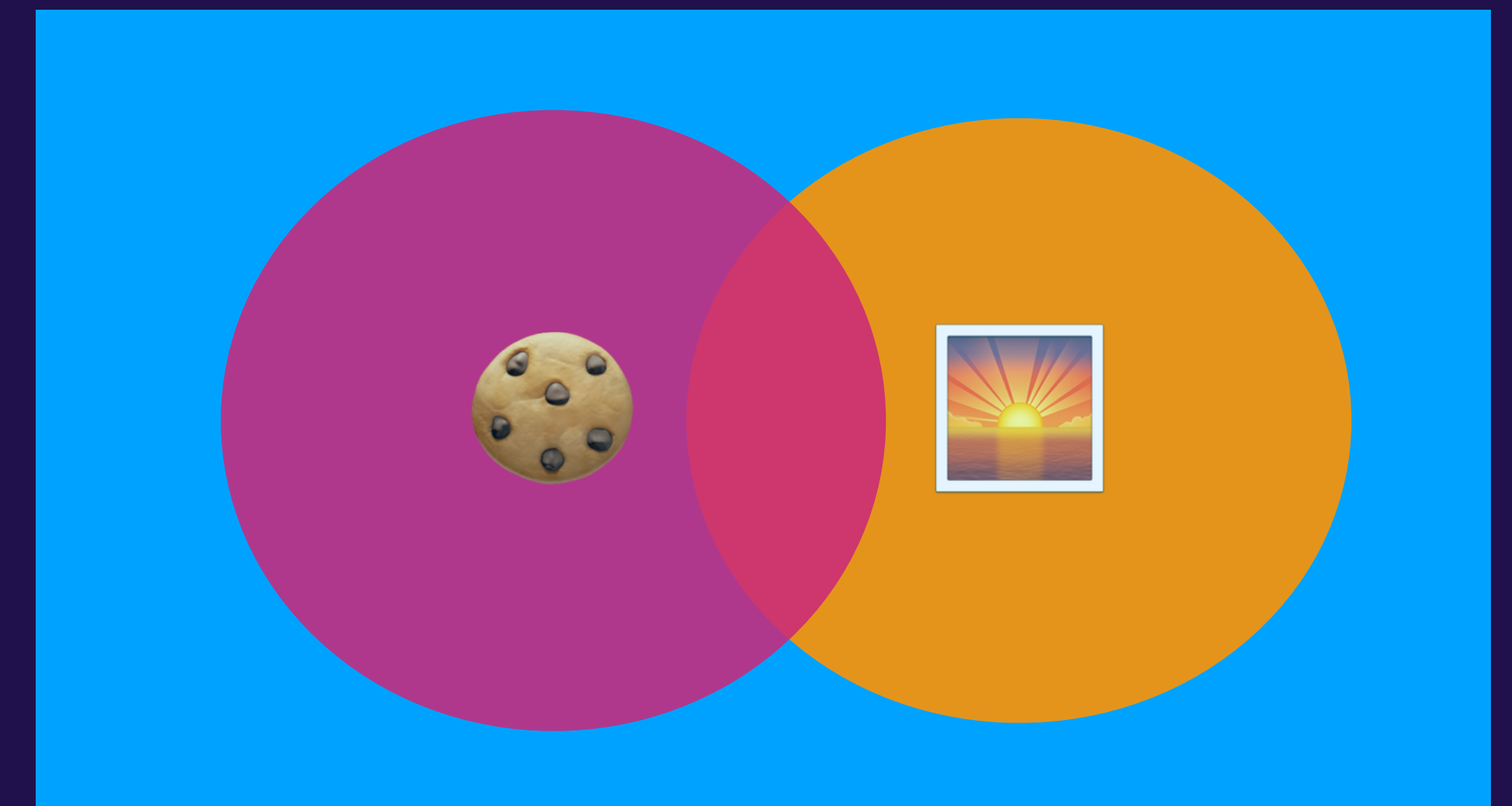
$p(\text{😴})$ = there are snacks in the afternoon

Do we get cookies more often in the morning or in the afternoon?

$p(\text{🍪} \mid \text{🌅})$ = “probability of 🍪 given 🌅”

$p(\text{🍪} \mid \text{🌅}) = p(\text{🍪} \cap \text{🌅}) / p(\text{🌅})$

$p(\text{🍪} \mid \text{🌅}) = p(\text{🍪}, \text{🌅}) / p(\text{🌅})$



Which snacks do we get at which time of the day?

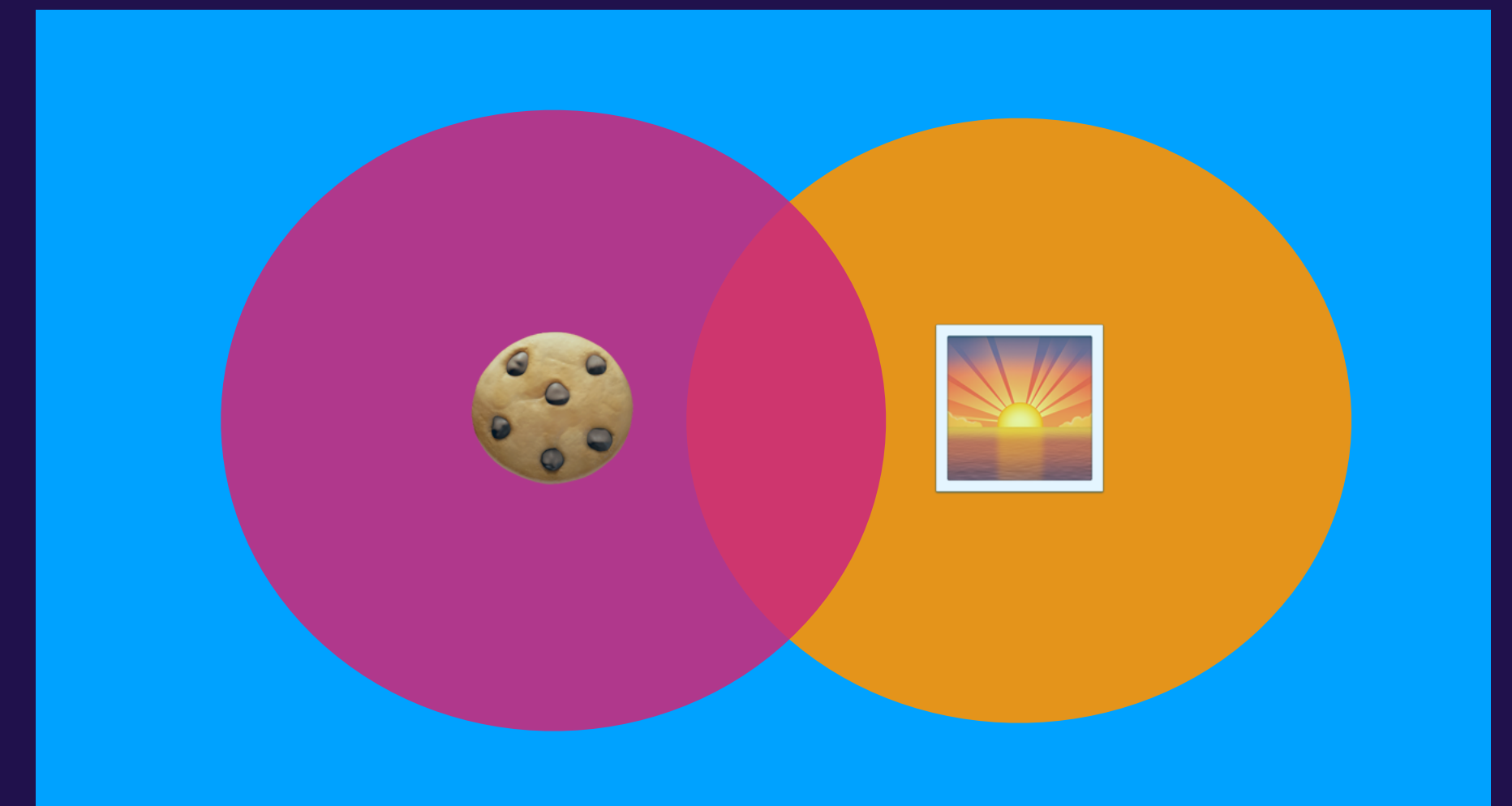
	$p(\text{🍪})$	$p(\text{🥐})$	$p(\text{🍰})$
$p(\text{🌅})$	0.25	0.06	0.29
$p(\text{😴})$	0.25	0.14	0.01

- $p(\text{🍰}) = ?$
- $p(\text{🌅} \mid \text{🥐}) = ?$
- $p(\text{🥐}) = ?$
- $p(\text{🍰} \mid \text{🌅}) = ?$

$$p(\text{🍪} \mid \text{🌅}) = p(\text{🍪}, \text{🌅}) / p(\text{🌅})$$

What is $p(\text{🌅} \mid \text{🍪})$?

$$p(\text{🌅} \mid \text{🍪}) = p(\text{🍪}, \text{🌅}) / p(\text{🍪})$$

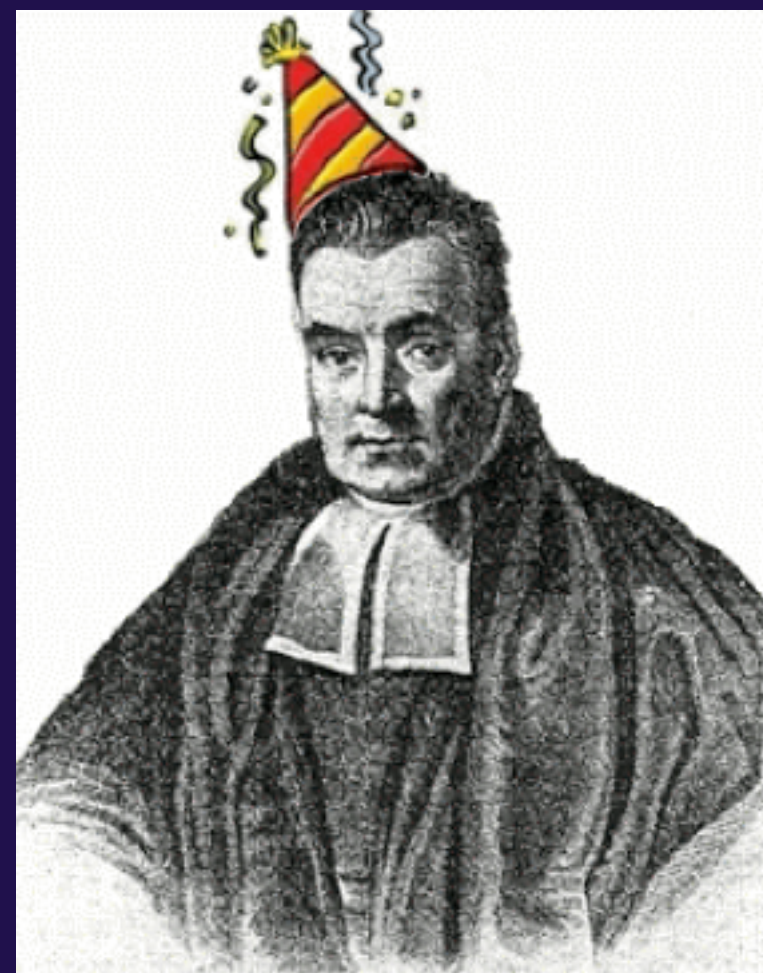


$$p(\text{🍪} \mid \text{🌅}) = p(\text{🍪}, \text{🌅}) / p(\text{🌅})$$

$$p(\text{🌅} \mid \text{🍪}) = p(\text{🍪}, \text{🌅}) / p(\text{🍪})$$

$$p(\text{🌅} \mid \text{🍪}) p(\text{🍪}) = p(\text{🍪} \mid \text{🌅}) p(\text{🌅})$$

Bayes rule



$$P(A | B) = \frac{P(B | A)P(A)}{P(B)}$$

Bayes rule

Important: $P(A | B) \neq P(B | A)$

$$p(\text{☔} | \text{☁}) \neq p(\text{☁} | \text{☔})$$

$$P(\theta | y) = \frac{P(y | \theta)P(\theta)}{P(y)}$$

y = data

θ = model parameters

black hole X-ray spectrum

black body temperature



What is the probability of drawing a **blue** M&M from a bag?

What is the fraction of **blue** M&Ms made at the factory?

likelihood

prior

posterior

$$P(\theta | y) = \frac{P(y | \theta) P(\theta)}{P(y)}$$

marginal likelihood
or evidence

$$P(\theta | y) \propto P(y | \theta) P(\theta)$$

posterior

likelihood

prior

$$P(\theta | y) \propto P(y | \theta) P(\theta)$$

Find the posterior distribution for the percentage of **blue** m&ms made at the factory

What kind of data are m&ms?

1. **nominal**
2. **continuous**

How will you record the data? **Table (all colours)**

How might we model the probability of drawing a **blue** m&m?

likelihood



$$P(\theta | y) \propto P(y | \theta) P(\theta)$$

y = number of successes (blue m&ms)

$n - y$ = number of failures (not-blue m&ms)

θ = probability of drawing a blue m&m

I have drawn the following sequence of blue (b) and not-blue (t) m&ms:

S = b b t b t t t b t b

What's the probability of drawing exactly that sequence?

$$\begin{aligned} p(S) &= \theta\theta(1-\theta)\theta(1-\theta)(1-\theta)(1-\theta)\theta(1-\theta)\theta \\ &= \theta^5(1-\theta)^5 \end{aligned}$$

Binomial distribution

$$p(y | \theta) \propto \theta^y (1 - \theta)^{n-y}$$

$$P(\theta | y) \propto P(y | \theta) P(\theta) \text{ prior}$$

Prior distribution: What do we think we know about the colour distribution of m&ms?



How many **different colours** of M&Ms are there?

What **percentage** of **blue** m&ms are made at the factory?

Do you think every bag of m&ms will have the **same percentage** of **blue** m&ms?

Sketch your prior



prior

$$P(\theta | y) \propto P(y | \theta) P(\theta)$$

beta distribution

$$P(\theta) \propto \theta^{\alpha-1} (1 - \theta)^{\beta-1}$$

“conjugate prior”

$\alpha = ?$

$\beta = ?$

Write a function for the beta distribution and try out different values for alpha and beta

Some values to try:

$$\alpha = \beta$$

$$\alpha = \beta = 1$$

$$\alpha, \beta \text{ very large}$$

$$\alpha, \beta \text{ very small}$$

likelihood

$$p(y) \propto \theta^y (1 - \theta)^{n-y}$$

prior

$$P(\theta) \propto \theta^{\alpha-1} (1 - \theta)^{\beta-1}$$



$$P(\theta | y) \propto P(y | \theta) P(\theta)$$

$$p(\theta) \propto \theta^{y+\alpha-1} (1 - \theta)^{n-y+\beta-1}$$

Gather some data!



take the first 20 m&ms out of your bag ($n=20$)

record the counts for all colours



Use your function for the beta distribution to plot both the **prior** and the **posterior** in the **same plot**

- Is the posterior distribution what you expected?
- Compare the posterior distribution to the prior distribution: Is this the result you expected, given six different colours?
- How sensitive is the posterior to the prior?

How would you expect the posterior to change with **more data**?

Let's **pool all the data** and find out!

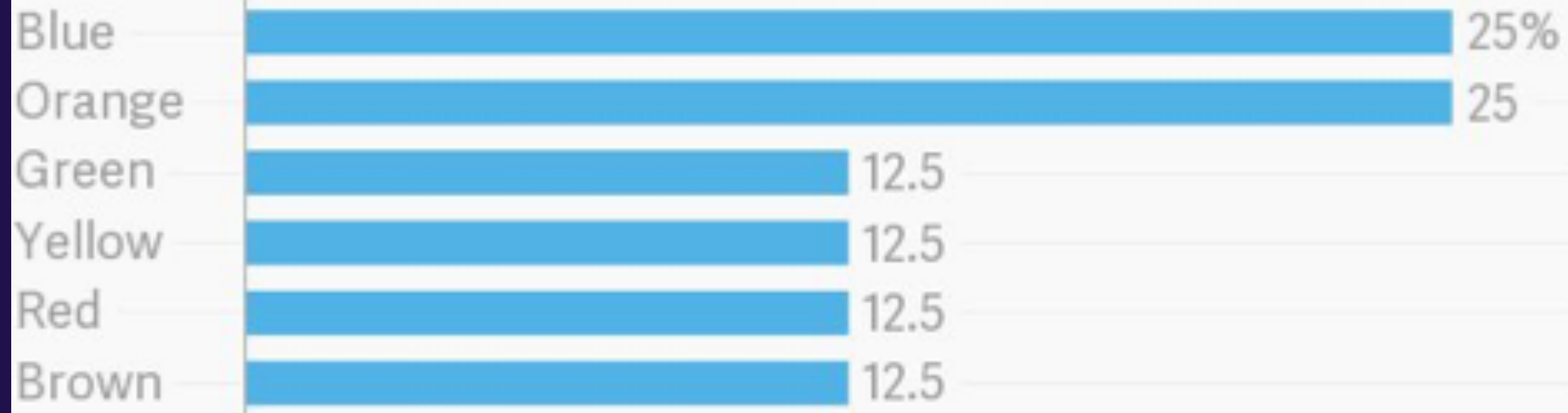
<https://forms.gle/zHy8LSbJsB8Pm9XA9>

Surprise Twist!

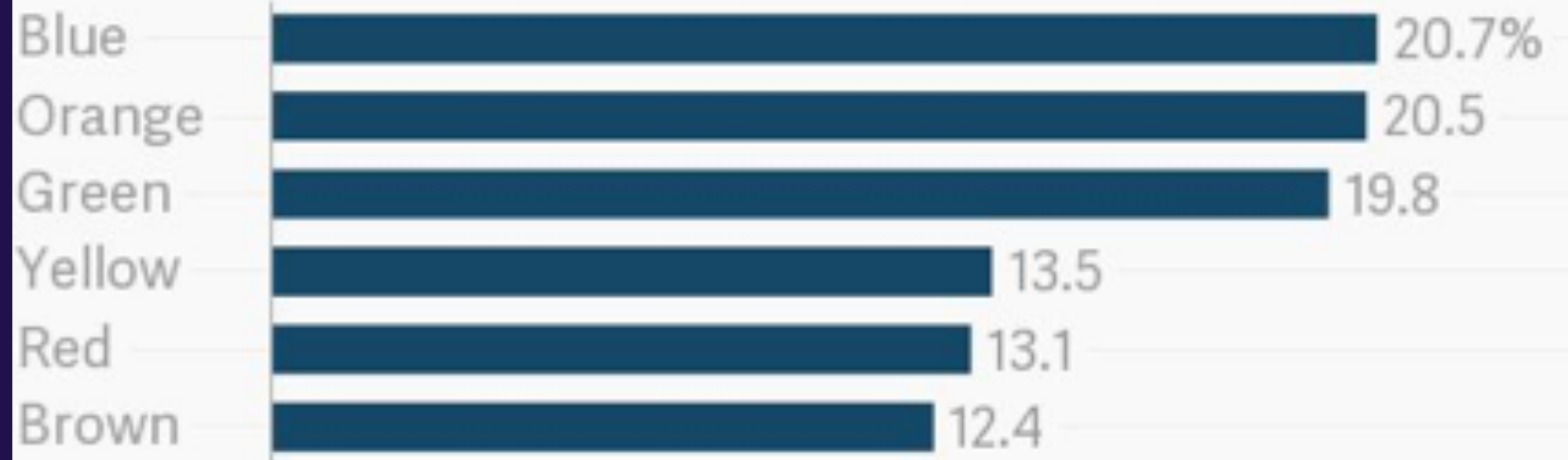
Different
factories make
different
distributions of
m&ms!

M&Ms color distribution, c. 2017

New Jersey factory



Tennessee factory

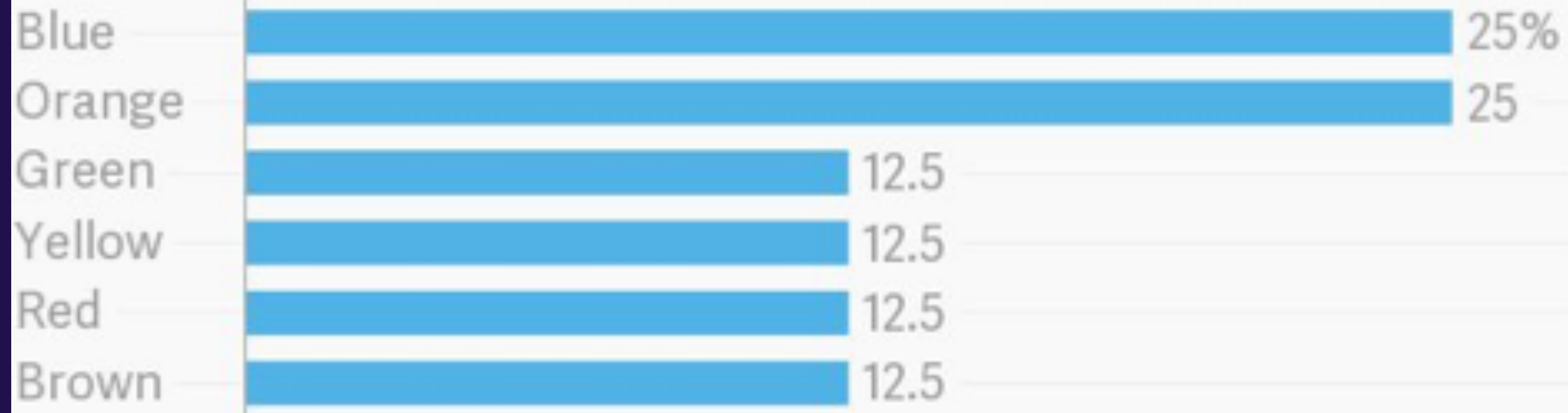


Surprise Twist!

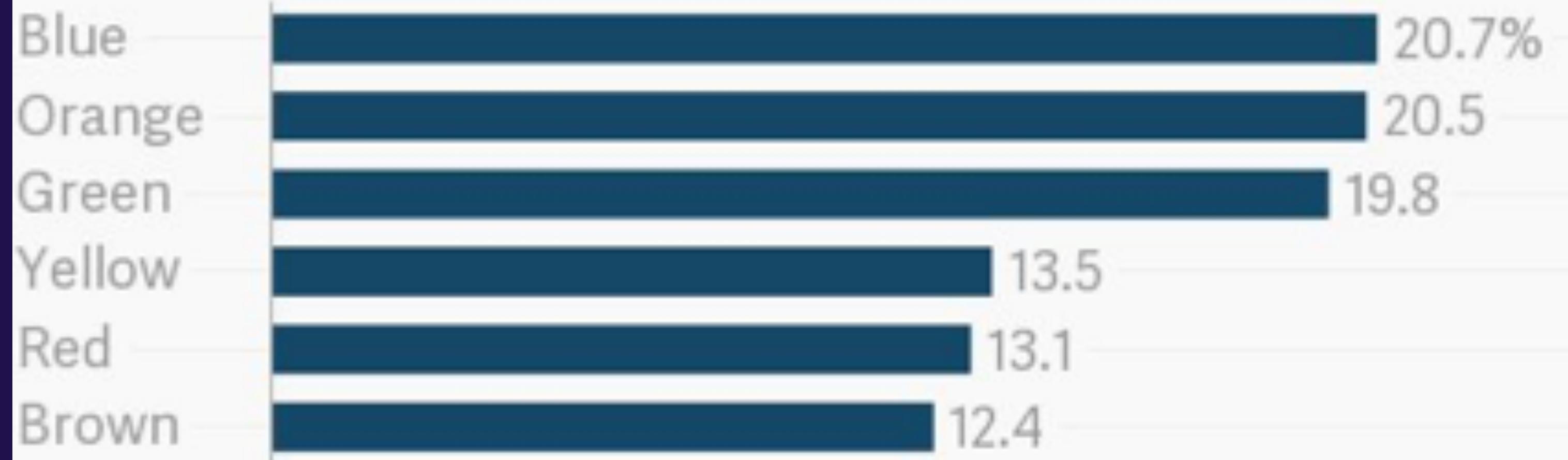
Which factory did your m&ms come from?

M&Ms color distribution, c. 2017

New Jersey factory



Tennessee factory



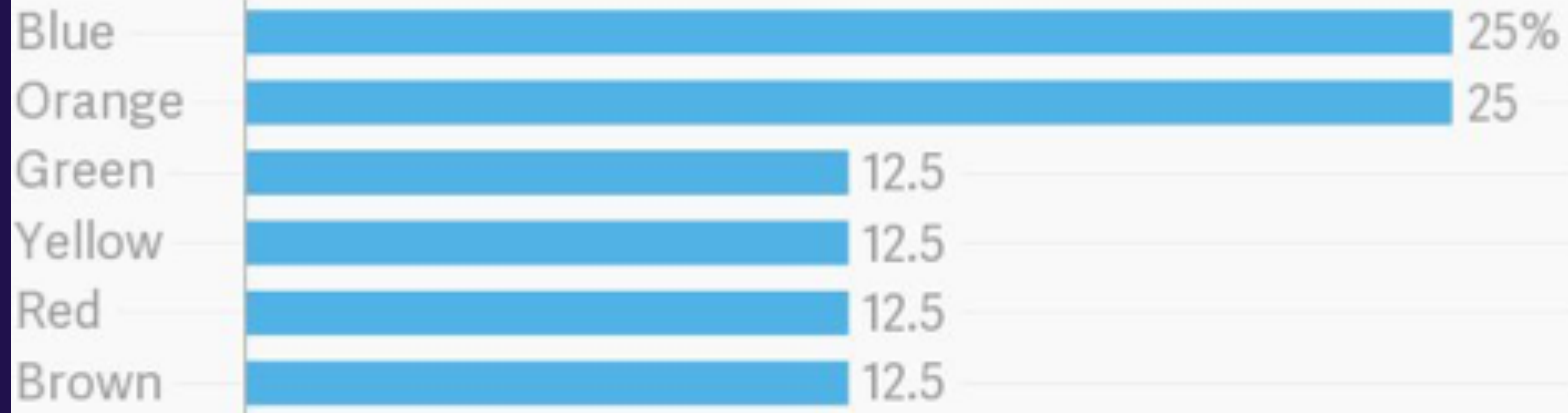
Surprise Twist!

**New Jersey =
HKL**

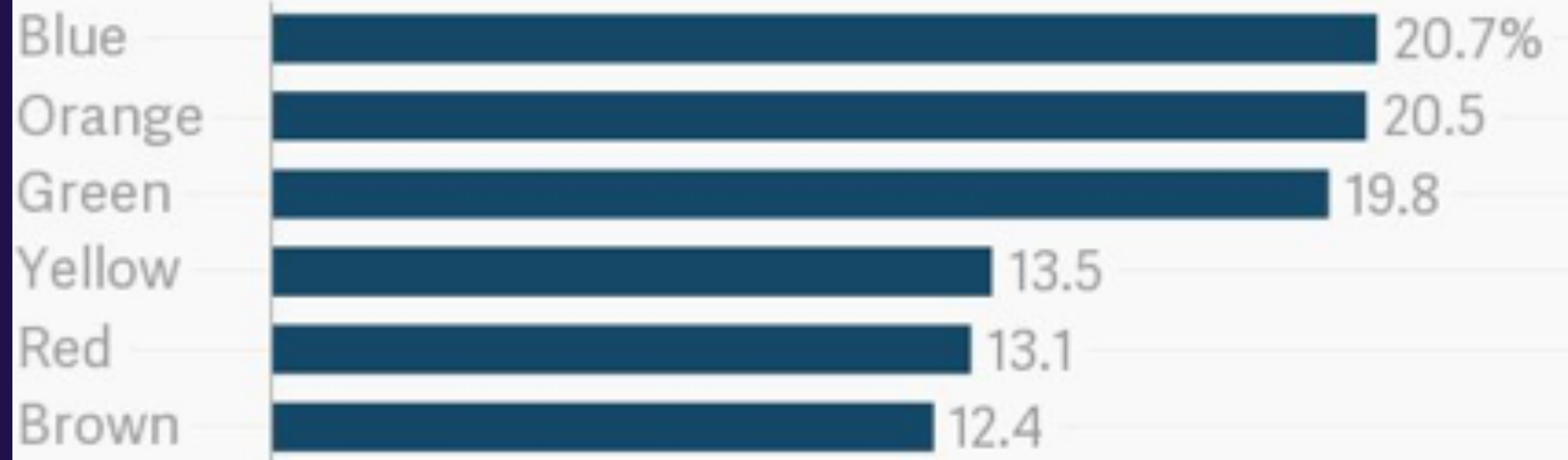
**Tennessee =
CLV**

M&Ms color distribution, c. 2017

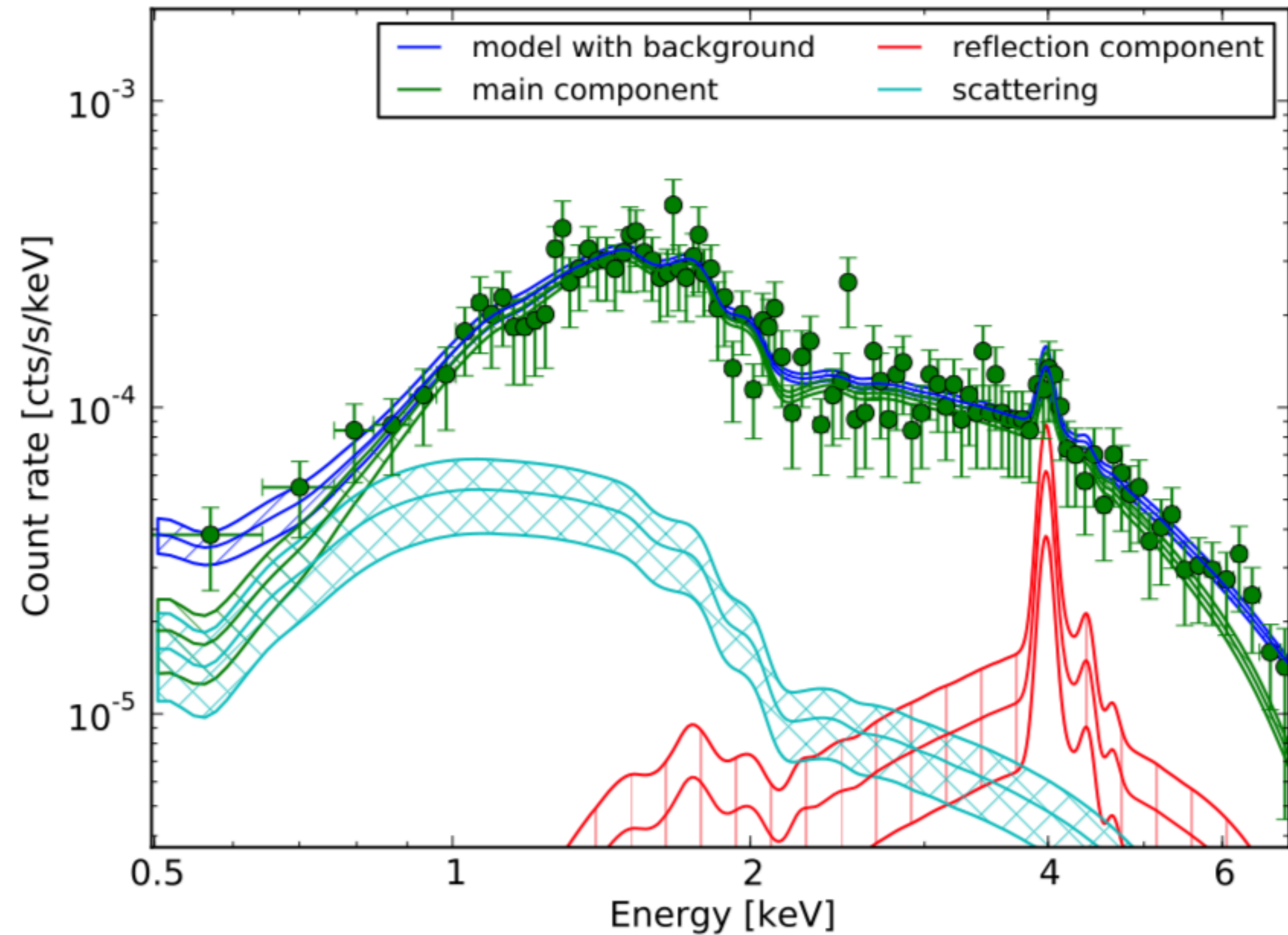
New Jersey factory



Tennessee factory



What about astronomy?



Data:

X-ray spectrum

Model:

**absorption +
scattering + reflection**

Likelihood: Poisson

Ethical considerations in statistics

data and algorithms are social constructs

... are created by humans

... encode biases

**... can be (mis-)used to
serve an agenda**

Based on the **M&Ms exercise**, can you think of ways our statistical procedure could be **mis-used** or be made to be **misleading**?

Opinion: Is science really facing a reproducibility crisis, and do we need it to?



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Article

Figures & SI

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Abstract

Efforts to improve the reproducibility and integrity of science are typically justified by a narrative of crisis, according to which most published results are unreliable due to growing problems with research and publication practices. This article provides an overview of recent evidence suggesting that this narrative is mistaken, and argues that a narrative of epochal changes and empowerment of scientists would be more accurate, inspiring, and compelling.

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Article

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Psychology journal bans *P* values

Test for reliability of results 'too easy to pass', say editors.

Chris Woolston

26 February 2015 | Clarified: 09 March 2015

PDF

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A controversial statistical test has finally met its end, at least in one journal. Earlier this month, the editors of *Basic and Applied Social Psychology* (BASP) announced that the journal would no longer publish papers containing *P* values because the statistics were too often used to support lower-quality research¹.

Most mistakes I see made in statistical procedures are either **mis-applications of methods or **mis-interpretation** of the results**

Know your assumptions!



