

Gov 51: Bayes Rule

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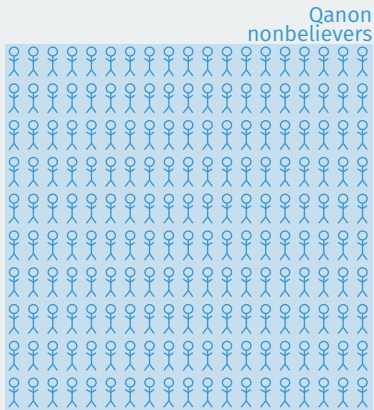
- Common response: probably believes in QAnon since believers tend to be Republicans.



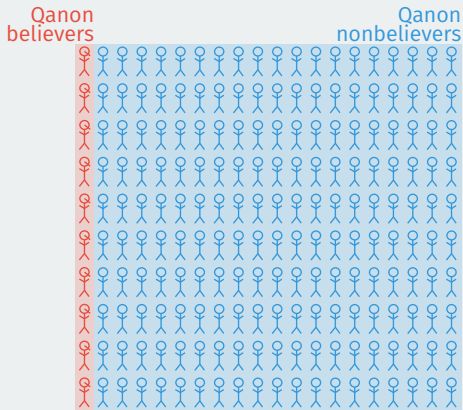
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- Common response: probably believes in QAnon since believers tend to be Republicans.
- **Base rate fallacy:** ignores how uncommon QAnon believers are!

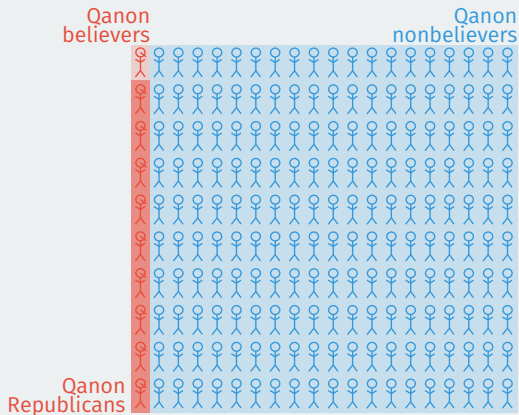
Visualizing QAnon support



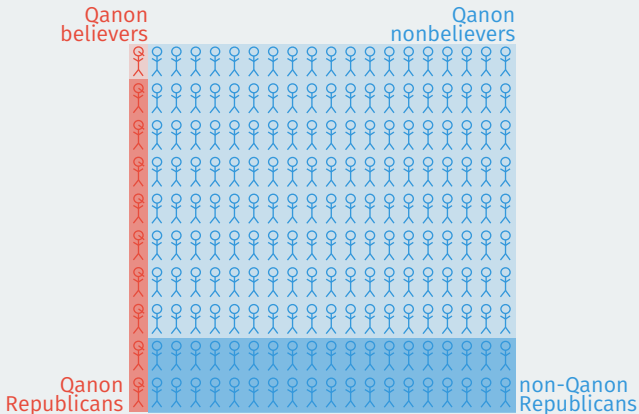
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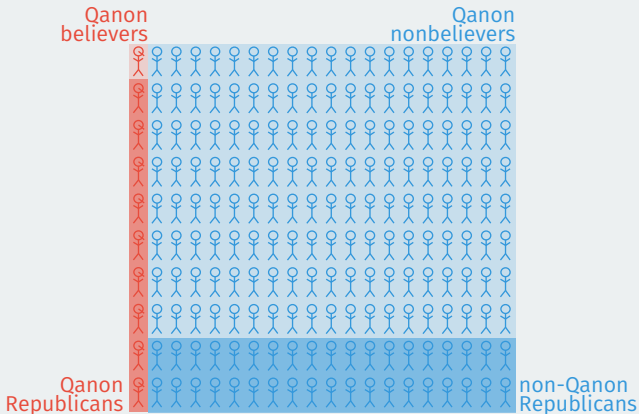
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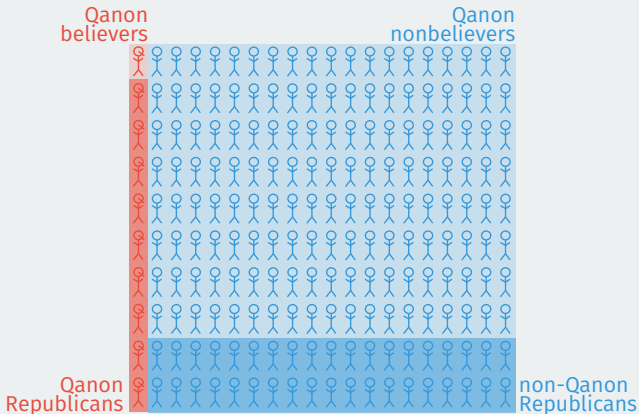


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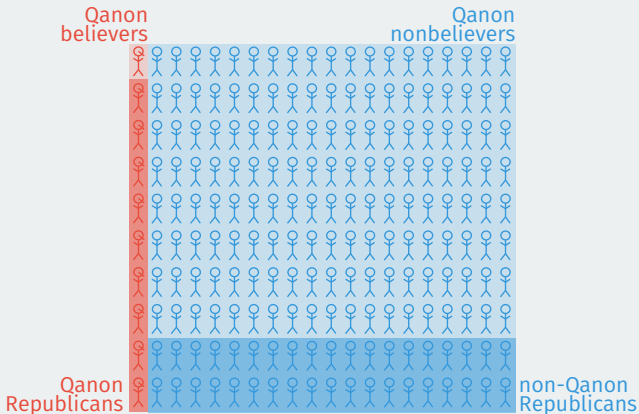
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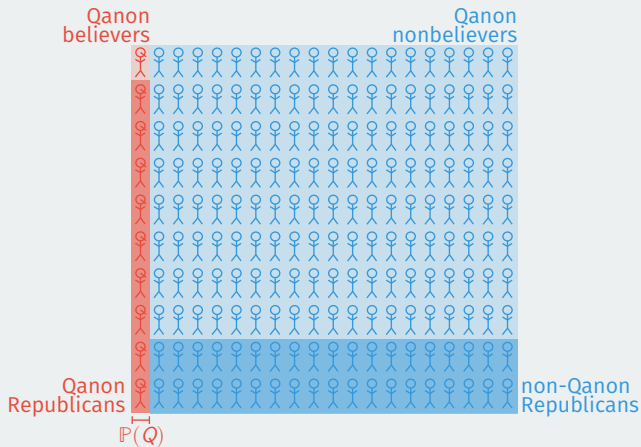
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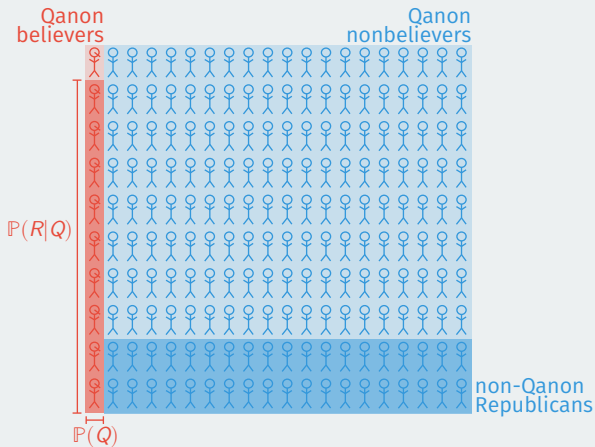
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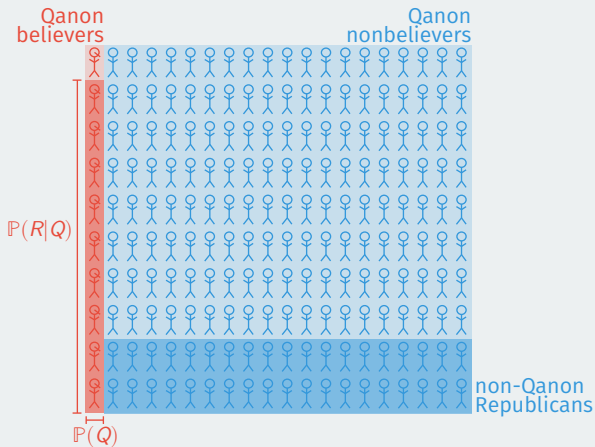
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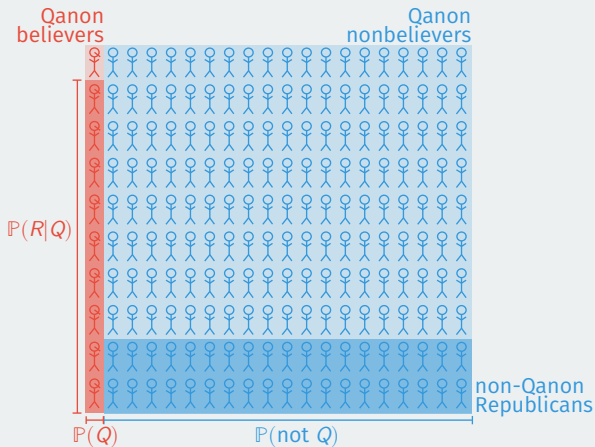
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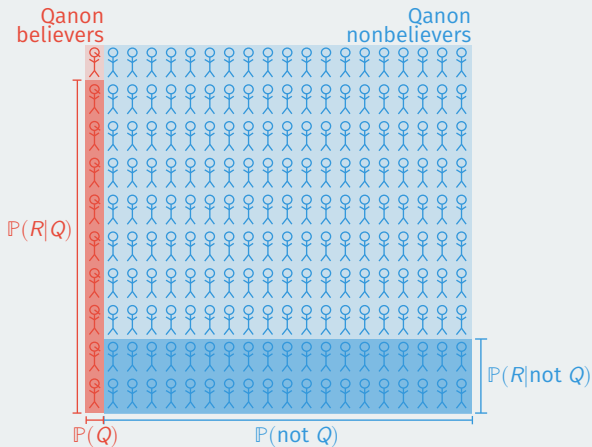
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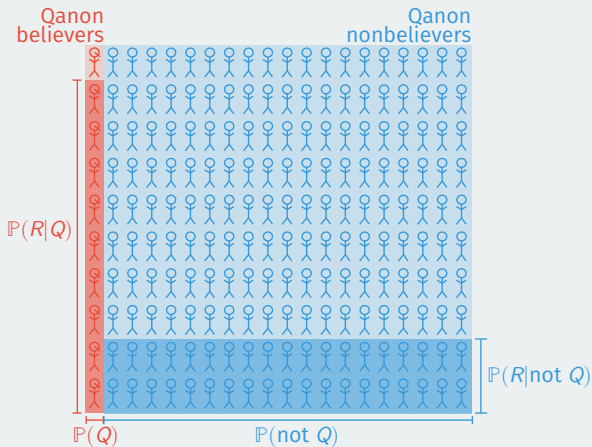
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 - How does the evidence change the chance of the hypothesis being true?

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 - $\mathbb{P}(PT \mid C) = 0.7$: true positive rate
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 - $\mathbb{P}(C) = 0.001$ rough prevalence of active COVID cases.

Applying Bayes' rule to COVID tests

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$$\mathbb{P}(C | PT) = \frac{\mathbb{P}(PT | C)\mathbb{P}(C)}{\mathbb{P}(PT)} = \frac{0.7 \times 0.001}{0.051} \approx 0.014$$