# Probability Updating When Drawing Signals Without Replacement 

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## Conceptual Background

Setup:
You face two decks of cards:one predominantly red, theYou draw a (red) card.What is the updated
probability that you faprobability that you face the
predominantly red deck?predominantly red deck?
What is the updated
What is the updated
probability to draw another
probability
red card?

The standard paradigm in experimental research on probability updating
Drawing with replacement


What is different in the
two settings?
Next to the
"state information effect" ",state information effect „removal effect".

Absolute card numbers matter! Absolute card numbers matter!
The strength of the removal effect The strength of the removal effect
depends on the absolute numbers depends on the absolute numbers


Many real world situations rather resemble a „without replacement" setting:

- High dividend payments point to a financially strong firm BUT the money has left the firm.
- High previous sales point to an attractive product BUT might also indicate saturated product
demand.
- A high current level of oil production suggests generally rich oil deposits BUT might also indicate that the oil field will soon be depleted.
- High past returns of an investment opportunity


## Experimental Design

To investigate individuals' probability updating when signals are drawn without replacement, we conducted a preregistered and incentivized lab experiment with 148 students.



This figure shows the deck size N (horizontal axis) and the number of signals n (vertical axis) of the scenarios considered in the three parts of the experiment. The difference between red and black cards drawn was always 1. Dots with $\mathrm{n}=1$ mark scenarios which were used in Part 1 of the experiment where subjects had to compare outcome probabilities $p$ with a benchmark probability of $55 \%$. The arrows connect the scenarios for which the subjects had to compare outcome probabilities $p$ in Part 2 of the
experiment. In Part 3 of the experiment, subjects had to provide explicit outcome probabilities $p$ for ea scenario indicated by a black dot.

## Results

Part 1:


The table provides changes in Bayesian probabilities, that is, the total effect $\Delta p^{*}$ as well as its two constituting components removal effect $\Delta p_{r}^{*}$ and state information effect $\Delta p_{s}^{*}$. The last three columns and the right-hand diagram show the proportion of subjects who believe in $p<50 \%, p=50 \%$, and $p>50 \%$. The striped background marks the correct answer for each judgment task

## Part 3:

| N | $n$ | $a$ | $b$ | $p^{*}$ | $p_{\text {median }}$ | $p_{\text {median }}-p^{*}$ | $\Delta p^{*}$ | $\Delta p_{r}^{*}$ | $\Delta p_{s}^{*}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 3 | 1 | 1 | 0 | 0.3333 | 0.3500 | 0.0167 | -0.1667 | -0.2500 | 0.0833 |
| 6 | 1 | 1 | 0 | 0.4667 | 0.4500 | -0.0167 | -0.0333 | -0.1000 | 0.0667 |
| 6 | 3 | 2 | 1 | 0.5000 | 0.4000 | -0.1000 | 0.0000 | -0.1667 | 0.1667 |
| 9 | 1 | 1 | 0 | 0.5000 | 0.4800 | -0.0200 | 0.0000 | -0.0625 | 0.0625 |
| 9 | 3 | 2 | 1 | 0.5238 | 0.4600 | -0.0638 | 0.0238 | -0.0833 | 0.1071 |
| 9 | 5 | 3 | 2 | 0.6000 | 0.4400 | -0.1600 | 0.1000 | -0.1250 | 0.2250 |
| 15 | 1 |  | 0 | 0.5238 | 0.4800 | -0.0438 | 0.0238 | -0.0357 | 0.0595 |
| 15 | 3 | 2 | 1 | 0.5385 | 0.4900 | -0.0485 | 0.0385 | -0.0417 | 0.0801 |
| 15 | 5 | 3 | 2 | 0.5636 | 0.4800 | -0.0836 | 0.0636 | -0.0500 | 0.1136 |
| 15 | 9 | 5 | 4 | 0.7143 | 0.4700 | -0.2443 | 0.2143 | -0.0833 | 0.2976 |
| 30 | 1 | 1 | 0 | 0.5402 | 0.4900 | -0.0502 | 0.0402 | -0.0172 | 0.0575 |
| 30 | 5 | 3 | 2 | 0.5569 | 0.4865 | -0.0704 | 0.0569 | -0.0200 | 0.0769 |
| 30 | 9 | 5 | 4 | 0.5844 | 0.4900 | -0.0944 | 0.0844 | -0.0238 | 0.1082 |

This table provides information on each of the 13 judgment tasks for which subjects stated explicit outcome probabilities, that is, deck size $N$, number of cards drawn $n$, number of red cards drawn a, number of black $c$ cat
drawn b , Bayesian outcome probability $p *$, subjects' median outcome probability estimate $p_{\text {median }}$ and the resulting median judgment bias $\mathrm{p}_{\text {median }}-p^{*}$

## Key Insights

Part 1: - A large portion of subjects does not update their beliefs in the correct direction
Subjects are not completely insensitive with respect to differences between the five scenarios.
Subjects' judgment is more in line with the removal effect than with the state information effect

Part 3: - Median outcome probability estimates monotonically increase in N for each level of n .

- Subjects tend to systematically underestimate the outcome probability to observe a red card in another draw.


## What this paper is about:

Individual biases in probabilistic belief updating have been typically examined in experimental settings where signals are drawn with replacement. Motivated by the variety of real world applications, we investigate belief updating in without replacement settings.

In such settings, drawing a specific signal has two opposed effects on the likelihood to obtain another signal of the same type in a further draw:

- First, and equivalent to drawing with replacement, the signal provides information on the state of the world resulting in a higher probability to draw the same signal type again.
- Second, the drawn signal is removed reducing the probability to draw the same signal type again.

We find that subjects severely underinfer with respect to the first effect and mildly underestimate the second effect.

