

A Mouse-Tracking Bandit Experiment on Meaningful Learning in Weighted Voting¹

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Suppose that you were a subject.

choice stage (30 sec.); no choice within 30 seconds → zero point.
Each number in red can be viewed by a mouse-click.

Round 1 out of 60

remaining time 16

Choose one of two committees. The quota is 14 in both committees.
The committee you chose allocates 120 points among four members.
Click on Choice 1 button or Choice 2 button within 30 seconds.

Choice 1	member	YOU	Player 2	Player 3	Player 4
	votes	V1	V2	V3	V4

Choice 2	member	YOU	Player 2	Player 3	Player 4
	votes	W1	W2	W3	W4

Figure: Choice stage.

feedback stage 1 (30 sec.); OK button → feedback stage 2. Each number in red can be viewed by a mouse-click.

Round 1 out of 60

remaining time 23

Click on OK button within 30 seconds after seeing the following information. You chose the following committee.

Choice 1	member	YOU	Player 2	Player 3	Player 4
	votes	X1	X2	X3	X4

The committee decided to allocate 120 points in this time as follows.

member	YOU	Player 2	Player 3	Player 4
Points	p1	p2	p3	p4

OK

Figure: Feedback stage 1.

feedback stage 2 (30 sec.); OK button → next round. Each number in red can be viewed by a mouse-click.

Round 1 out of 60

remaining time 8

Click on OK button within 10 seconds after seeing the following information.

You obtained **yyy** points in this round.

You have obtained **zzz** points in total so far.



Figure: Feedback stage 2.

- ▶ Answer 60 binary choice problems (60 rounds).
When you finish 60 rounds, you are asked to wait until the others finish.
- ▶ 1 point = 1 JPY (about 115 JPY/ USD in 2017),
added to the payment for participation (1000 JPY)

mouse-tracking to capture what information subjects view on their monitor

- ▶ When measuring gaze with **eye trackers**, subjects often unconsciously move their eye-sight to places other than the measurement targets. (a kind of **noise**)

1. Introduction: Research Question

- ▶ **Question:** **What information** did subjects view, when they could (not) “meaningfully learn” the **latent feature** of weighted voting in a “2-armed bandit experiment”?
 - ▶ Felsenthal and Machover (1998): It would be difficult (for people) to see the **underlying relationship between the actual voting powers and the nominal voting weights**.
- ▶ When people generalize what they have learned in a situation to a similar but different one, this higher order concept of learning is called **meaningful learning** (Rick and Weber, 2010).
- ▶ **2-armed bandit** experiment: Subjects choose one of **two** weighted voting games repeatedly. Payoffs are determined stochastically according to a payoff-generating function that is **hidden** from subjects.

1. Introduction: Previous Research

- ▶ Guerci et al. (2017, TD) could **not observe meaningful learning** by subjects who received immediate feedback information on their **current payoffs**.
 - ▶ **Withholding feedback information induced meaningful learning** (introspective thinking). Did the subjects get **confused by the feedback information**?
 - ▶ Ogawa et al. (2021, mimeo.) reconfirmed a similar result at four different experimental sites.
- ▶ ref. Guerci et al. (2014, SCW): subjects played 4-person weighted voting games
 - ▶ not focused on their learning but compared frequencies of their “mistakes” observed in bargaining protocols in voting. (single approval or multiple approval...)
 - ▶ Esposito et al. (2012, mimeo.): subjects chose a weighted voting game and then they played the game they chose.
→ subjects’ learning was affected by others’ voting behavior.

1. Introduction: Main Result

We measured how often or long each subject views information necessary for his or her choice without duplication of viewing time.

The feedback Information about **cumulative payoffs promoted meaningful learning**, whereas **the information on current payoffs did not, or even hindered it**.

- ▶ The **latter part** of this result may explain the reason why meaningful learning was not observed by Guerci et al. (2017), because in their experiment **only the current payoffs** were shown to the subjects.

search behavior: win-stay-lose-shift (WSLS), random choice of runs
→ outcomes: **current payoff** and **failure in meaningful learning**

2. Experimental Design

- ▶ Subjects choose one of two weighted voting games.
 - ▶ $[q; v_1, v_2, v_3, v_4]$; q is the quota, v_i is the voting weight.
 - ▶ Each subject acts as **Member 1**, who has v_1 .
 - ▶ clearly informed that the other three members are **all fictitious**.
- ▶ one binary choice problem (Problem) for the first **40 rounds**, and a similar but different one in the following **20 rounds**;
 - ▶ This information was not provided to subjects.
- ▶ four sequences: $A \rightarrow B, B \rightarrow A, C \rightarrow D, D \rightarrow C$
- ▶ payoff-generating function: based on **Deagan-Packel index**
e.g., Problem A, Choice 1 \rightarrow **40 points** with Prob=2/3,
0 point with Prob=1/3

Table: Binary choice problems and expected payoffs for Member 1

Problem	Choice 1	(expected payoff)	Choice 2	(expected payoff)
A	[14; 5, 3, 7, 7]	$(120 \times \frac{2}{3} \times \frac{1}{3})$	[14; 5, 4, 6, 7]	$(120 \times \frac{3}{4} \times \frac{1}{3})$
B	[6; 1, 2, 3, 4]	$(120 \times \frac{1}{3} \times \frac{1}{3})$	[6; 1, 1, 4, 4]	$(120 \times \frac{2}{3} \times \frac{1}{3})$
C	[14; 3, 5, 6, 8]	$(120 \times \frac{2}{3} \times \frac{1}{3})$	[14; 3, 6, 6, 7]	$(120 \times \frac{3}{4} \times \frac{1}{3})$
D	[9; 1, 3, 5, 6]	$(120 \times \frac{1}{3} \times \frac{1}{3})$	[9; 1, 2, 6, 6]	$(120 \times \frac{2}{3} \times \frac{1}{3})$

choice stage (30 sec.); no choice within 30 seconds → zero point.
Each number in red can be viewed by a mouse-click.

Round 1 out of 60

remaining time 16

Choose one of two committees. The quota is 14 in both committees.
The committee you chose allocates 120 points among four members.
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feedback stage 1 (30 sec.); OK button → feedback stage 2. Each number in red can be viewed by a mouse-click.

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Click on OK button within 30 seconds after seeing the following information. You chose the following committee.

Choice 1	member	YOU	Player 2	Player 3	Player 4
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The committee decided to allocate 120 points in this time as follows.

member	YOU	Player 2	Player 3	Player 4
Points	p1	p2	p3	p4

OK

Figure: Feedback stage 1.

feedback stage 2 (30 sec.); OK button → next round. Each number in red can be viewed by a mouse-click.

Round 1 out of 60

remaining time 8

Click on OK button within 10 seconds after seeing the following information.

You obtained **yyy** points in this round.

You have obtained **zzz** points in total so far.



Figure: Feedback stage 2.

- ▶ Subjects answer 60 binary choice problems.
→ a message that asks them to wait until the others finish
(→ Questionnaire and Raven test)
- ▶ 1 point = 1 JPY, added to the payment for participation
(1000 JPY)

3-1. Analysis: Session Details

- ▶ December 17 in 2014 to January 17 in 2017
- ▶ at the University of Tsukuba
- ▶ undergraduate students who did not know voting indices
- ▶ 40 subjects for each sequence (160 subjects in total)
- ▶ the average amount paid as a reward was 2507 JPY

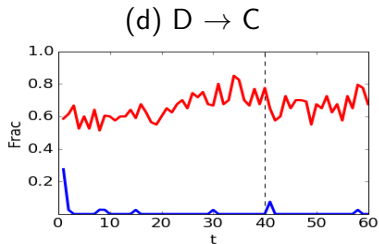
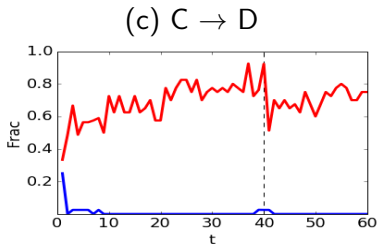
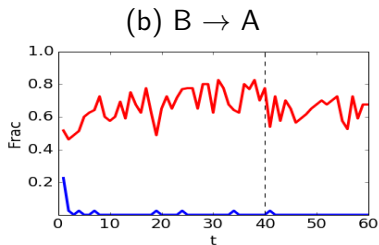
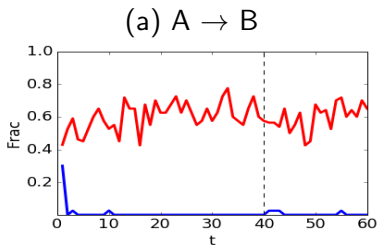


Figure: RED: fraction of subjects who chose the correct answer among those who made choices before the time limit. BLUE: fraction of those who failed to make their choice before the time limit.

- ▶ FR_k^i : the relative frequency of rounds in which subject i chose the correct answer within the k -th block of 5 consecutive rounds.
- ▶ The change in the relative frequencies that subject i chose the correct answer between the l -th block and the m -th block is defined as

$$\Delta FR_{l,m}^i = FR_l^i - FR_m^i,$$

where $l > m$.

- ▶ $\Delta FR_{l,m} (FR_l, FR_m)$: a vector the i -th component of which is $\Delta FR_{l,m}^i (FR_l^i, FR_m^i)$.
 - ▶ When the elements of FR_l are, on average, significantly larger than those of FR_m , we write this as $\Delta FR_{l,m} > 0$.

3-2. Results: Aggregate Data

Definitions: For each binary choice problem, we consider that subjects learned the “correct” answers if $\Delta FR_{8,1} > 0$ was statistically confirmed and that those who learned in the binary choice problem meaningfully learned the “underlying structure” of weighted voting games if $\Delta FR_{9,1} > 0$ in the same binary choice problem was statistically confirmed.

Result 1: For all binary choice problems, the subjects learned the correct answer. For all binary choice problems except Problem B, the subjects meaningfully learned the underlying structure of weighted voting.

	Problem A	Problem B	Problem C	Problem D
$\Delta FR_{2,1}$	0.0415	0.0086	0.0396	0.6943
$\Delta FR_{8,1}$	0.0143	< 0.0001	< 0.0001	< 0.0001
$\Delta FR_{9,1}$	0.0147	0.2498	< 0.0001	0.0230

Table: P-values for the two-sided signed-rank test (normalized). The null hypotheses are $\Delta FR_{2,1} = 0$, $\Delta FR_{8,1} = 0$, and $\Delta FR_{9,1} = 0$, respectively

- ▶ An auxiliary experiment was conducted at Kansai University: **Cumulative payoffs** were shown to the subjects' monitors as well as their current payoffs, and similar results to Results 1 and 2 were observed. → The **use of mouse trackers** would **not** affect subjects' choice in the aggregate level.

3-3. Results: Individual Data

- ▶ **two groups** of subjects $G_0^{91}(s)$ and $G_1^{91}(s)$ for binary choice problem $s = A, B, C, D$: $\Delta FR_{9,1}^i = FR_9^i - FR_1^i > 0$ or not
- ▶ **canonical discriminant analysis**: the **first 40 rounds** to identify important **mouse-tracked** variables that had a large effect on the classification of the two groups.
 - ▶ Box's M test on variance-covariance matrices \rightarrow **Mahalanobis distance** was applied to identify the discriminant function

variable	
<i>A_count</i> :	numbers of views of A; $A = v_r, w_r, x_r, p_r$ ($r = 1, 2, 3, 4$), <i>yyy</i> , <i>zzz</i>
<i>A_time</i> :	cumulative time for viewing A
<i>decision_time</i> :	time spent for the final decision in the choice stage
<i>A_decision_time</i> :	relative length of time spent for viewing A up to the final decision
<i>main2_ok</i> :	time spent in feedback stage 1
<i>main3_ok</i> :	time spent in feedback stage 2
<i>no_info</i> :	dummy variable that takes a value of 1 if the individual subject views v_1, \dots, v_4 or w_1, \dots, w_4 even once; otherwise, 0
<i>judgment</i> :	dummy variable that takes a value of 1 if $\Delta FR_{8,1}^i > 0$; otherwise, 0
<i>judgment2</i> :	dummy variable that takes a value of 1 if $\Delta FR_{9,1}^i > 0$; otherwise, 0

Table: Major variables.

Definition: For each sequence of binary choice problems, we considered that subject i learned the “correct” answers if $\Delta FR_{8,1}^i > 0$ and that those who learned in a binary choice problem meaningfully learned the “underlying structure” of weighted voting games if $\Delta FR_{9,1}^i > 0$ in another binary choice problem in the sequence.

- ▶ the numbers of subjects who succeeded in learning (meaningful learning) were 20, 30, 32, and 27 (14, 13, 19, 17), respectively, under the following definitions.
- ▶ For every binary choice problem $s = A, B, C, D$, we could reject the null hypothesis of no difference between $G_0^{81}(s)$ and $G_1^{81}(s)$ (between $G_0^{91}(s)$ and $G_1^{91}(s)$) for independent variables, because Wilks' lambda < 0.001 .

A	zzz_count 0.678	<i>p1_time</i> 0.627	<i>v4decision_time</i> -0.486	yyy_count -0.484	<i>p3_time</i> -0.378
B	<i>decision_time</i> -0.827	<i>x1_decision_time</i> -0.552	<i>x1_count</i> 0.521	zzz_count 0.506	yyy_count 0.241
C	<i>x1_count</i> -1.359	<i>x3_count</i> 1.040	yyy_time -0.549	<i>w4_time</i> 0.543	<i>v3_time</i> 0.473
D	<i>p4_time</i> 0.749	zzz_count 0.479	<i>p1_time</i> 0.315	<i>no_info</i> -0.298	yyy_count -0.142

Table: Standardized coefficients in the canonical discriminant functions: the four largest absolute values observed in the first 40 rounds in the discriminant function that separated well $G_0^{91}(s)$ and $G_1^{91}(s)$ for binary choice problem $s = A, B, C, D$.

Result 2:

The information on subjects' **cumulative payoffs** promoted meaningful learning, whereas the information on their own **current payoffs** did not, or even hindered it.

- ▶ The **latter part** of this result may explain the reason why meaningful learning was not observed by Guerci et al. (2017), because in their experiment **only the current payoffs** were shown to the subjects.
 - ▶ Watanabe (2022): reconfirmation of meaningful learning without any feedback information (correct answers were reversed in the first 40 rounds and the following 20 rounds).

Subjects' Search Behavior

In what follows, we examine the null hypotheses in 20 rounds.

For each individual subject i , let a_i denote the number of rounds in which $yyy_count - zzz_count \geq 0$ and denote b_i the number of rounds in which $yyy_time - zzz_time \geq 0$.

- ▶ We say that in 20 consecutive rounds, subject i paid **more attention to the cumulative payoffs** than to the current payoffs, if $a_i \leq 10$ and $b_i \leq 10$ in those rounds.
- ▶ If either $a_i > 10$ or $b_i > 10$, we say that subject i paid **more attention to the current payoffs**.

Define *current payoff/runruns* as the number of subjects who paid more attention to the current payoffs among those who chose the runs randomly and *meaningful/randruns* as the number of subjects who succeeded in meaningful learning among those who chose the runs randomly.

In the next table, the *values in the parentheses* are the numbers of subjects who paid *more attention to the cumulative payoffs* among subjects who chose the runs of options randomly and the numbers of those who *failed in meaningful learning* among those who chose the runs randomly.

- ▶ Under the null hypothesis in the (Wald-Wolfowitz) runs test, the number of runs of options chosen by a subject is a random variable.
 - ▶ Computer scientists consider the optimal length of runs of options in which decision-makers should continue to choose the same options for accumulating the feedback information.

Result 3:

It was plausible that more subjects paid attention to cumulative payoffs but failed in meaningful learning when they chose the runs of options randomly.

rounds	current payoff/randruns			meaningful/randruns		
	1-20	21-40	41-60	1-20	21-40	41-60
A → B	11 (17)	10 (13)	6 (18)	9 (19)	8 (15)	7 (17)
B → A	11 (19)	7 (9)	8 (14)	13 (17)	14 (14)	8 (14)
C → D	12 (8)	6 (14)	7 (15)	10 (10)	9 (11)	12 (10)
D → C	18 (12)	10 (18)	10 (15)	11 (19)	9 (19)	6 (19)

Table: Frequencies of observations: current payoff, meaningful learning, and random choice of runs. The rejection of the null hypothesis at the 5% significance level in the one-sided binomial test is indicated in boldfaced value.

Rounds 21-40 are more important than rounds 1-20 due to the definition of learning.

4. Final Remarks

Note: We could not observe that individual subjects exactly took Win-stay-lose-shift strategy (Nowak and Sigmund, 1993).

Summary:

- ▶ Information on subjects' **cumulative payoffs** promoted subjects' deep understanding (**meaningful learning**) the underlying structure of weighted voting.
- ▶ Even if subjects paid more attention to their cumulative payoff, they would fail in meaningful learning when they **chose the runs of options randomly**.
- ▶ future research
 - ▶ **length of memory** → case-based decision theory (Gilboa and Schmeidler (1995, 2001))
 - ▶ **response time** in the process of learning (ref. Rubinstein, 2013)