

Econ 211

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Neuroeconomics

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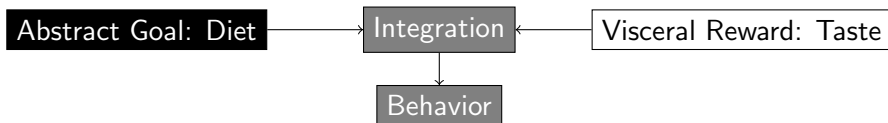
- ▶ *Neuroeconomics* is the study of economic decision-making through its biological foundations in the brain
- ▶ What are these biological foundations?
 - ▶ Neural mechanisms like neurons, chemical pathways, brain systems
 - ▶ Genetics
- ▶ How do we measure these foundations?
 - ▶ Scans like PET, CAT, MRI
 - ▶ Secondary reactions like skin conductance, pulse rate, eye tracking

Multiple Systems Hypothesis

- ▶ One possible neuroeconomic way to study behavior is the *multiple systems model*
- ▶ The model:
 - ▶ Brain is built up from many independent systems
 - ▶ Each system has a physical locus in the brain, and is specialized for a certain task or activity
 - ▶ Given a stimulus, each system produces a (potentially different) response
 - ▶ The brain integrate these multiple signals to decide on a final course of action

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- ▶ Example: do you want a cookie right now?

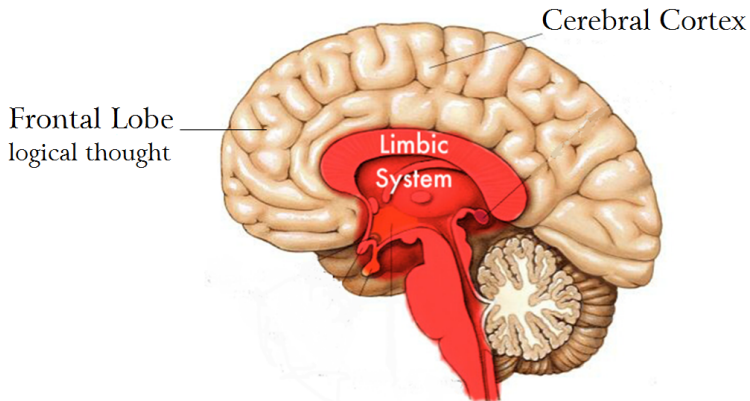


Connection to System 1 and 2

- ▶ The multiple systems model sounds a lot like Kahneman's System 1 and System 2
- ▶ However, system 1 and system 2 is just *one example* of a multiple systems hypothesis
- ▶ Other examples:
 - ▶ Freud's id, ego, and superego
 - ▶ Prefrontal cortex vs Mesolimbic dopamine system
 - ▶ Deliberative vs impulsive
 - ▶ Patient vs myopic
- ▶ Note that there can be more than two systems interacting in general

An Over-Simplified Model of the Brain

- ▶ Prefrontal cortex (PFC): the center higher reasoning, logic, self control
- ▶ Limbic system: releases dopamine in response to rewards like food and sex



Relation to Time Preferences and Self-Control

- ▶ Hypothesis: the PFC is patient but the limbic system is impatient
- ▶ Preferences are derived from adding up the outputs of the two systems
- ▶ For example, consider how the two systems evaluate the prospect of getting a small reward each period:

Period	1	2	3	4
PFC contribution	1	1	1	1
Limbic contribution	1	0	0	0
Average signal	1	$\frac{1}{2}$	$\frac{1}{2}$	$\frac{1}{2}$

- ▶ What does average signal look like?

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- ▶ What does average signal look like? Present-biased model with $\beta = \frac{1}{2}$ and $\delta = 1$

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- ▶ How could we easily implement this?
 - ▶ If we tax or distract the PFC, people should look more impatient
 - ▶ Alternatively, we can directly look at the signal strength with brain scans

Cognitive Load

- ▶ Shiv and Fedorikhin (1999) ask people to remember a number
- ▶ While holding the number in their head, they are asked if they want cake or fruit
- ▶ Two treatments:
 - ▶ High cognitive load: 7 digit number
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 - ▶ Could be that remembering longer numbers just makes you hungrier

Discount Rates

- ▶ Hinson, Jameson, and Whitney (2003) seek to measure time preferences directly using price list methodology we saw earlier in course
- ▶ Subjects choose between smaller, sooner reward and later, larger reward
- ▶ Vary the cognitive load in a similar way:
 - ▶ Control: no cognitive load
 - ▶ Treatment: hold a 5-digit number in memory

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 - ▶ Treatment: 49.8%

Measuring Brain Activity Directly

- ▶ McClure, Laibson, Loewenstein, and Cohen (2004) take a more direct approach
- ▶ Attempt to measure the signal coming from each of the two systems
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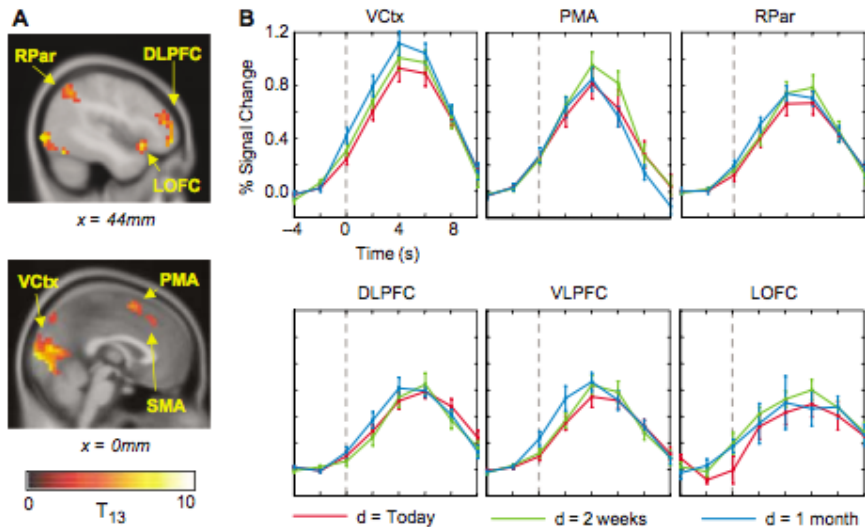
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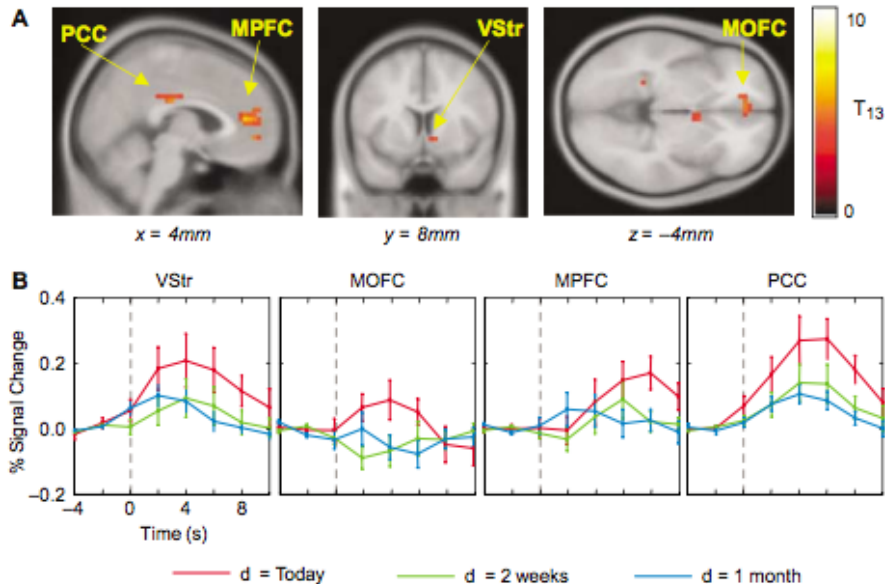
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- ▶ Predictions of which tasks brain areas will send signal?
 - ▶ PFC: Send signal for every task (the δ part of the $\beta - \delta$ model)
 - ▶ Limbic system: Send signal only for tasks with $d = 0$ (the β part)

δ Areas Activate for All Options



β Areas Activate Only for Options with Immediate Rewards



Behavioral Economics and The Internet

Motivation

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- ▶ Much more data being collected for empirical studies
 - ▶ 6,000 tweets per second
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 - ▶ Terabytes of publicly available financial data every day
- ▶ Also many more platforms for running experiments
 - ▶ Social media companies running experiments essentially constantly
 - ▶ Lower barrier to entry for researchers though Amazon Mechanical Turk

Is All This Useful?

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Is All This Useful?

- ▶ Question: does the internet make people better-informed?
 - ▶ Maybe yes:
 - ▶ Information is easier to obtain and verify
 - ▶ More likely to have conversations with people very different from yourself
 - ▶ Maybe not:
 - ▶ People may choose to surround themselves with connections and information sources that fit with their preferences
 - ▶ This is known as the *echo chamber effect*

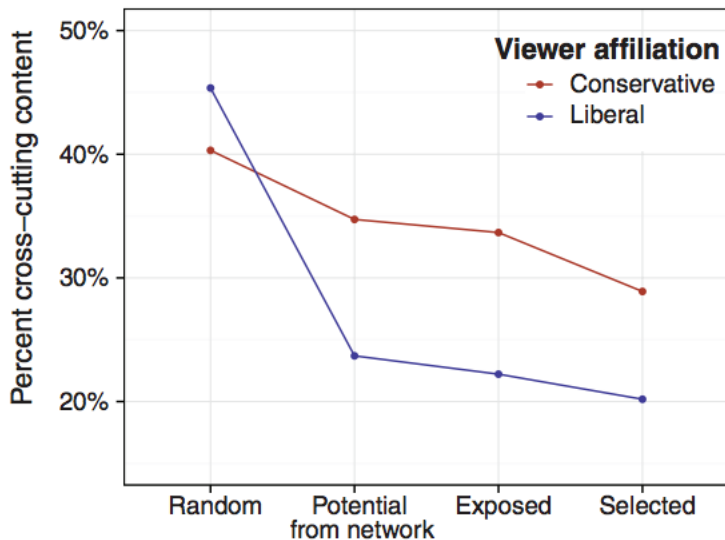
Facebook Echo Chamber Study

- ▶ Bakshay, Messing, Adamic (2015) address this issue using data from Facebook posts
- ▶ Observed approx. 10 million people on Facebook (no experimental variation)
- ▶ Linked stories were classified either “cross-cutting” or “ideologically consistent” with each person’s self-reported political affiliation
- ▶ What determines which content people read?

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- ▶ What determines which content people read?
 1. Your network of friends
 2. How Facebook shows you your friends’ content (Newsfeed)
 3. What content you choose to click on
- ▶ Baseline: how much cross-cutting content you would see if you were show random Facebook posts

Results from Adamic et al



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Viewer affiliation	Random → Potential	Potential → Exposed	Exposed → Selected
Liberal	-0.626	-0.080	-0.063
Conservative	-0.212	-0.046	-0.172

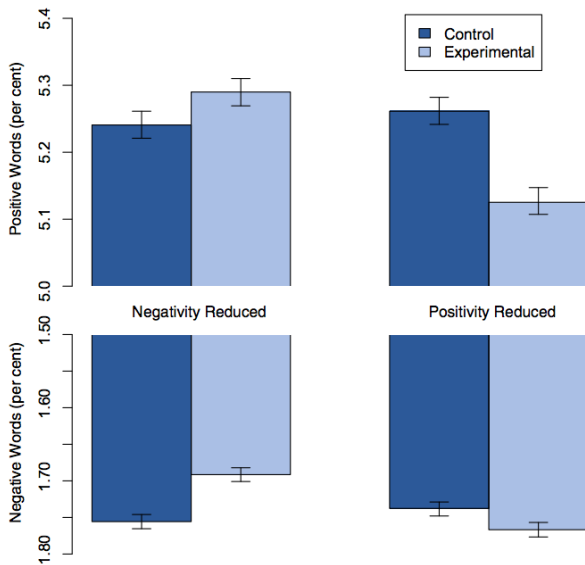
News Feed Experiment

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- ▶ Kramer, Guillory, and Hancock (2014) run experiment to determine how much of an effect news feed content has on user's emotions
- ▶ Experimental design:
 - ▶ Facebook posts categorized as either positive or negative
 - ▶ 22.4% negative, 46.8% positive
 - ▶ Treatment 1: Omit a percentage of all positive posts by friends that would otherwise show up on Newsfeed
 - ▶ Treatment 2: Omit a percentage of all negative posts by friends that would otherwise show up on Newsfeed
 - ▶ Controls: Omit a percentage of all posts
- ▶ Outcome variable: Positive/negative content of subjects' posts
- ▶ $N = 689,003$ people

Kramer et al Results



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- ▶ Results show emotional “contagion”
 - ▶ Omitting positive posts in feed lead to a 0.1% decrease in positive posts by subjects and a 0.04% increase in negative posts
 - ▶ Omitting negative posts in feed lead to a 0.07% decrease in negative posts by subjects and a 0.06% increase in positive posts
 - ▶ Results are statistically significant (due to large sample) but effect size is small

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- ▶ Responses to these objections?
 - ▶ Note that Facebook gathered consent through terms of use agreement
 - ▶ No claim that the baseline algorithm is good or bad for mental health
 - ▶ One could argue that Facebook has an obligation to test their algorithm

Methodology: Amazon Mechanical Turk

- ▶ Most researchers do not have access to Facebook data (and certainly not able to manipulate their software)
- ▶ However, other tools do exist to reach lots of people online
- ▶ One such tool: Amazon Mechanical Turk
 - ▶ Online labor platform of English-speaking workers
 - ▶ Employers posts small tasks with an associated wage rate
 - ▶ Tasks can include experiments (either explicitly or implicitly)
 - ▶ Much cheaper and faster than running lab or field experiment
- ▶ Another tool: Harvard Digital Lab for the Social Sciences