

# Econ 211

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# Neuroeconomics

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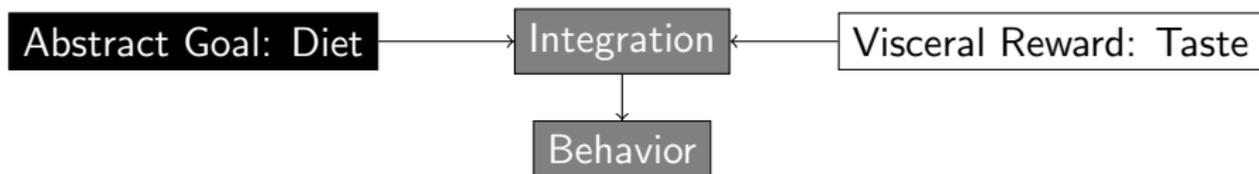
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- ▶ How do we measure these foundations?
  - ▶ Scans like PET, CAT, MRI
  - ▶ Secondary reactions like skin conductance, pulse rate, eye tracking

# 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
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  - ▶ The brain integrate these multiple signals to decide on a final course of action
- ▶ Example: do you want a cookie right now?

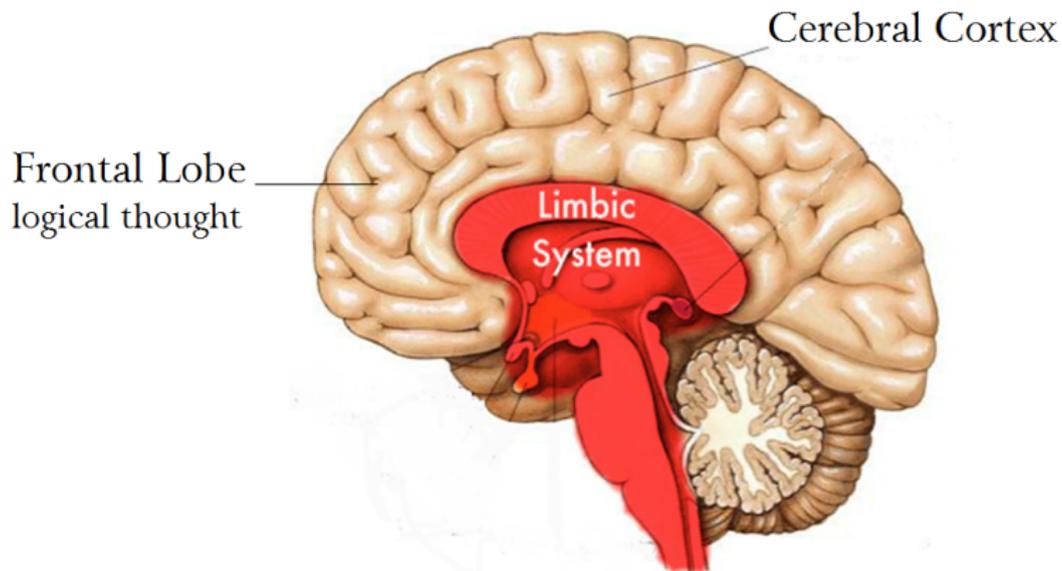


## Connection to System 1 and 2

- ▶ 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	$\frac{1}{2}$	$\frac{1}{2}$	$\frac{1}{2}$

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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
  - ▶ Low cognitive load: 2 digit number
- ▶ Results:

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- ▶ Any alternate explanations?
  - ▶ 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
- ▶ Task: Subjects make binary decisions between a smaller sooner reward and a larger later reward
  - ▶ Sooner period: delay  $d = 0, 2, \text{ or } 4$  weeks
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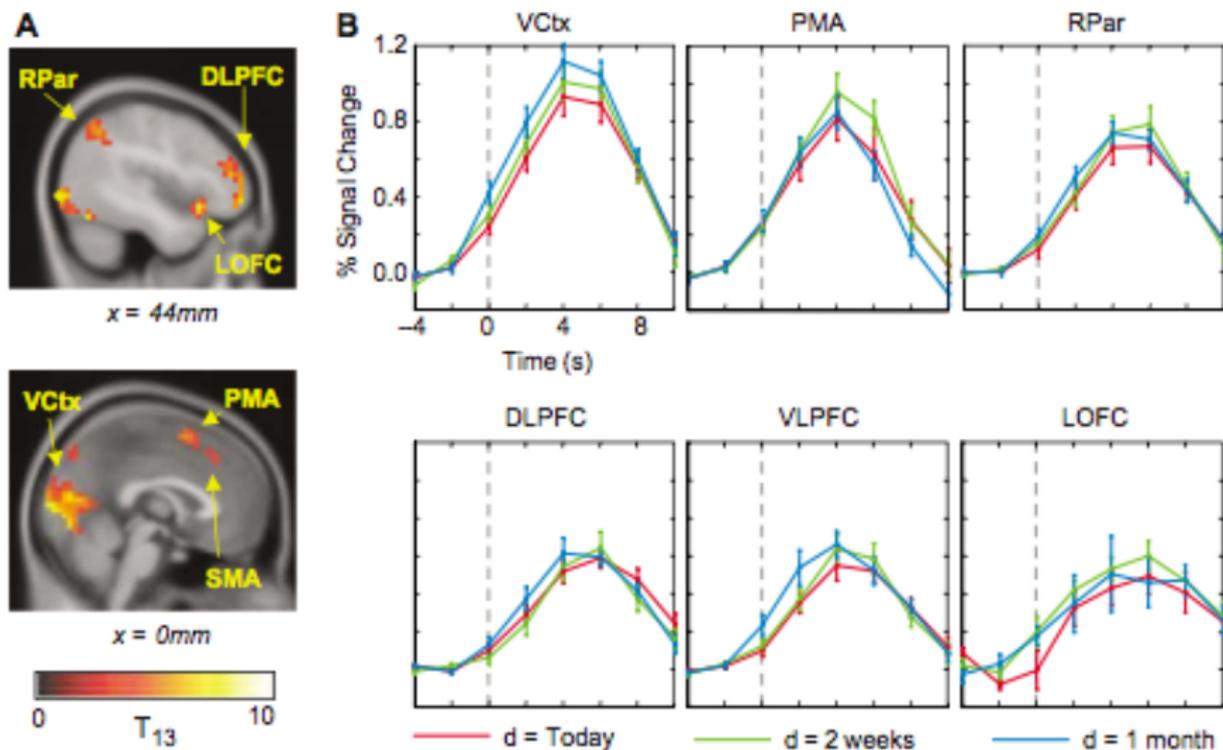
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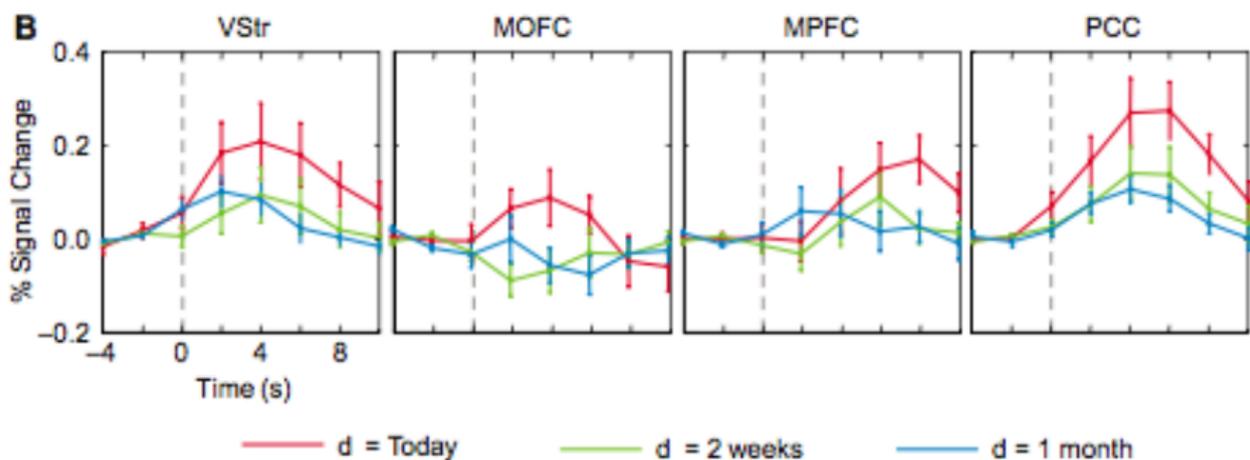
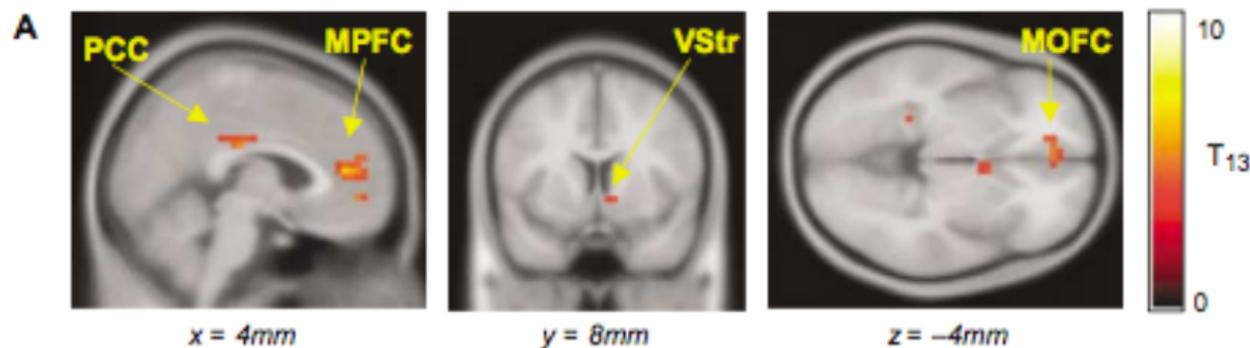
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- ▶ Predictions of which tasks brain areas will send signal?
  - ▶ PFC: Send signal for every task (the  $\delta$  part of the  $\beta - \delta$  model)
  - ▶ Limbic system: Send signal only for tasks with  $d = 0$  (the  $\beta$  part)

# $\delta$ Areas Activate for All Options



# $\beta$ Areas Activate Only for Options with Immediate Rewards



# Behavioral Economics and The Internet

# Motivation

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  - ▶ 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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  - ▶ 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 know as the *echo chamber effect*

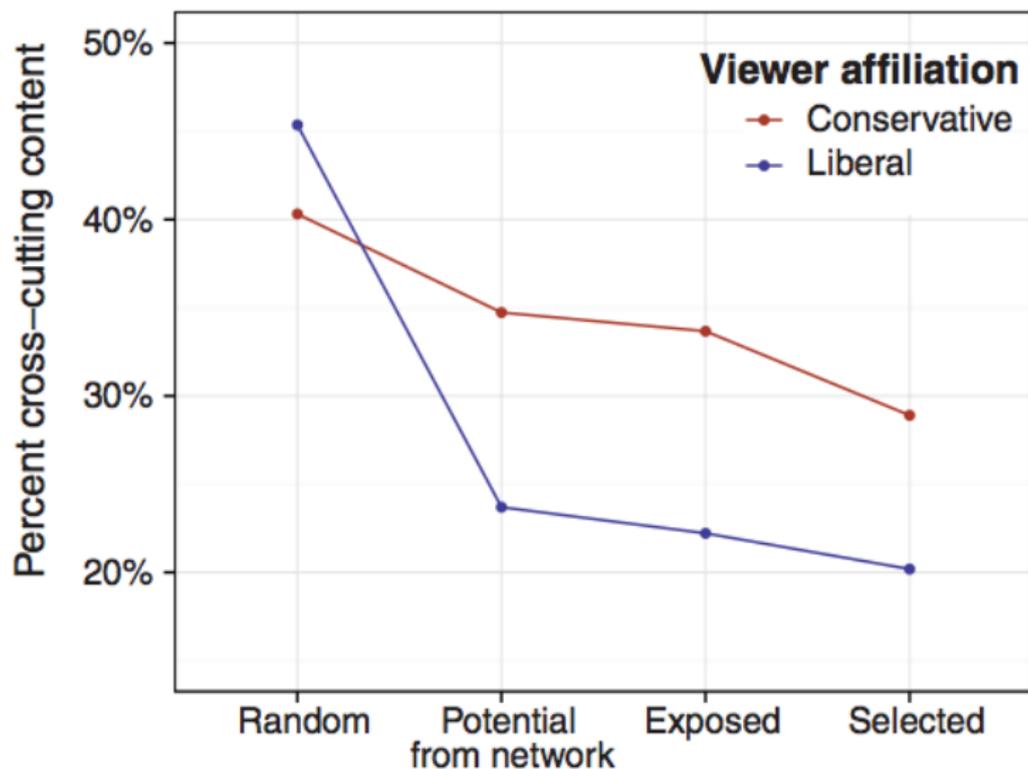
# Facebook Echo Chamber Study

- ▶ 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
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- ▶ Linked stories were classified either “cross-cutting” or “ideologically consistent” with each person’s self-reported political affiliation
- ▶ 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

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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

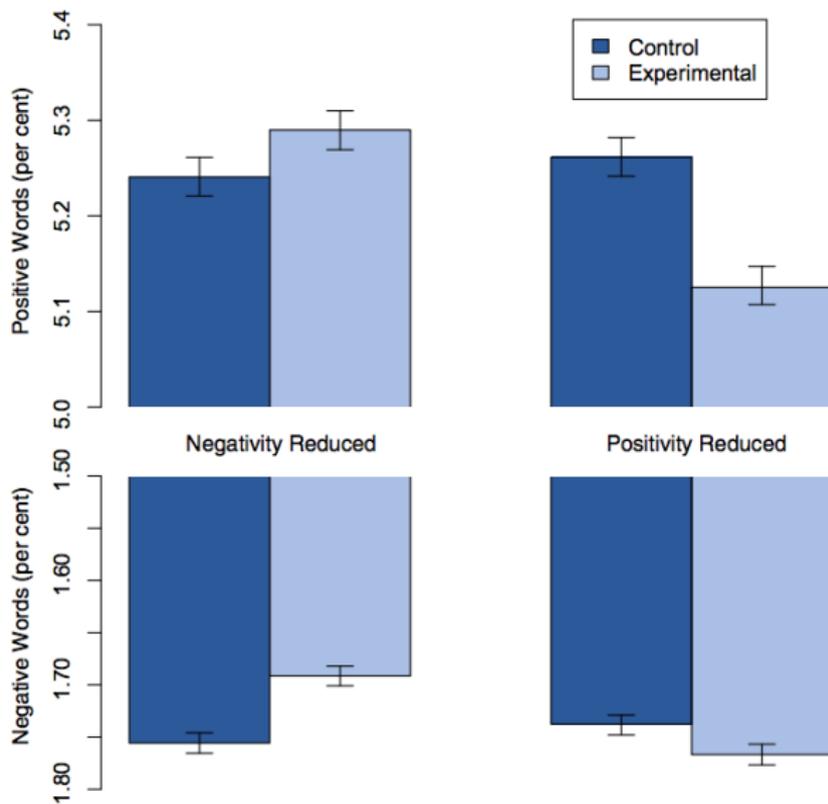
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- ▶ Experimental design:
  - ▶ Facebook posts categorized as either positive or negative
    - ▶ 22.4% negative, 46.8% positive
  - ▶ Treatment 1: Omit a percentage of all positive posts by friends that would otherwise show up on Newsfeed
  - ▶ Treatment 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