



The neurobiology of simple choice

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Caltech

Goals of neuroeconomics

- Characterize the computational processes used by the brain to make different types of choices
- Understand how does the neurobiology implements and constraints those computations
- Characterize the computational and neurobiological differences underlying decision maker heterogeneity
Ex:
 - > addicts vs non-addicts
 - > healthy eaters vs. Big Mac lovers

Simple economic choice



?



Why study simple choice?

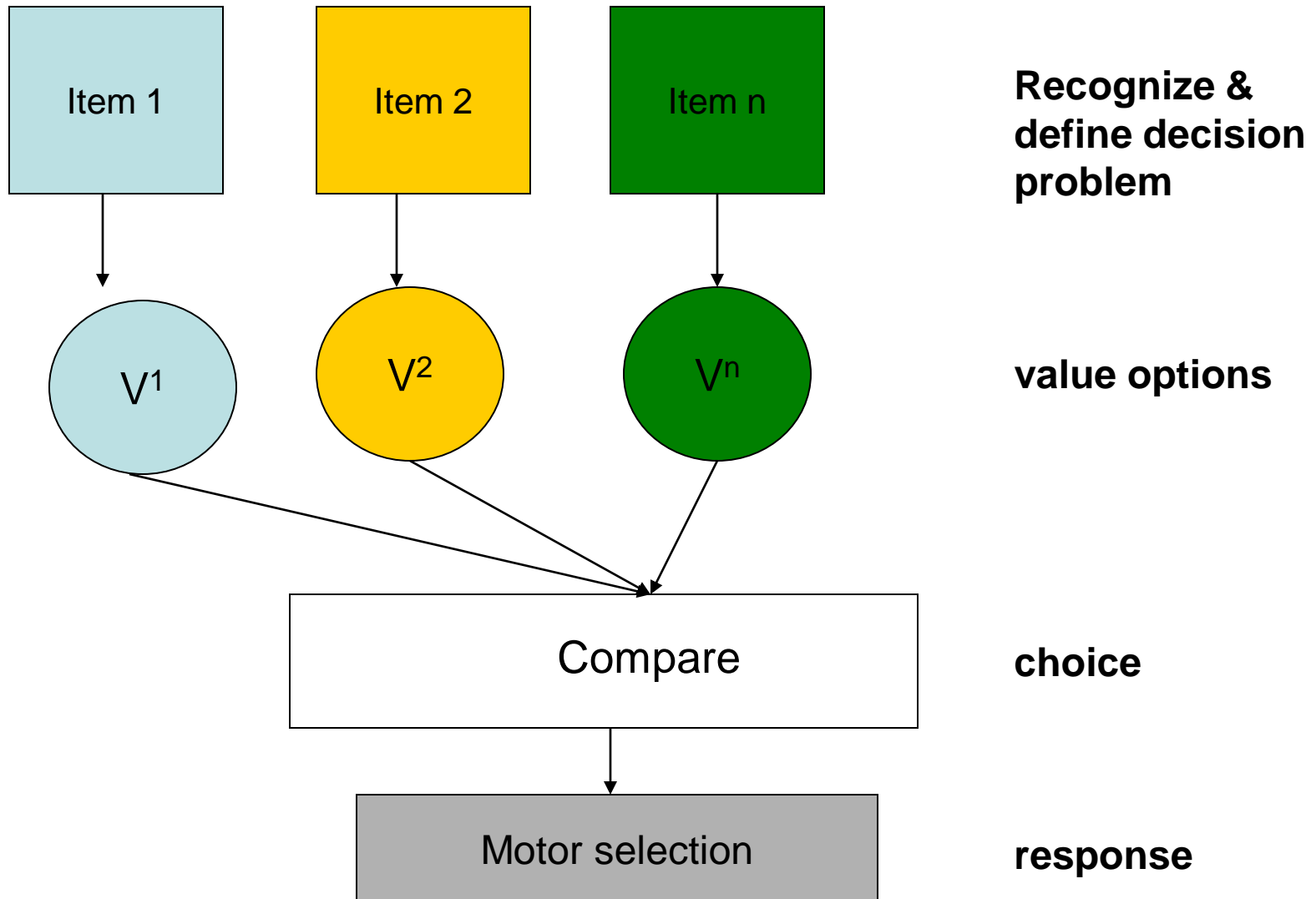
- Simplest setting to study the neurobiology of human DM
- Foundation for more complex choice situations
- Insights about limitations and unexpected features of DM circuitry already be present here



I

A simple but useful framework

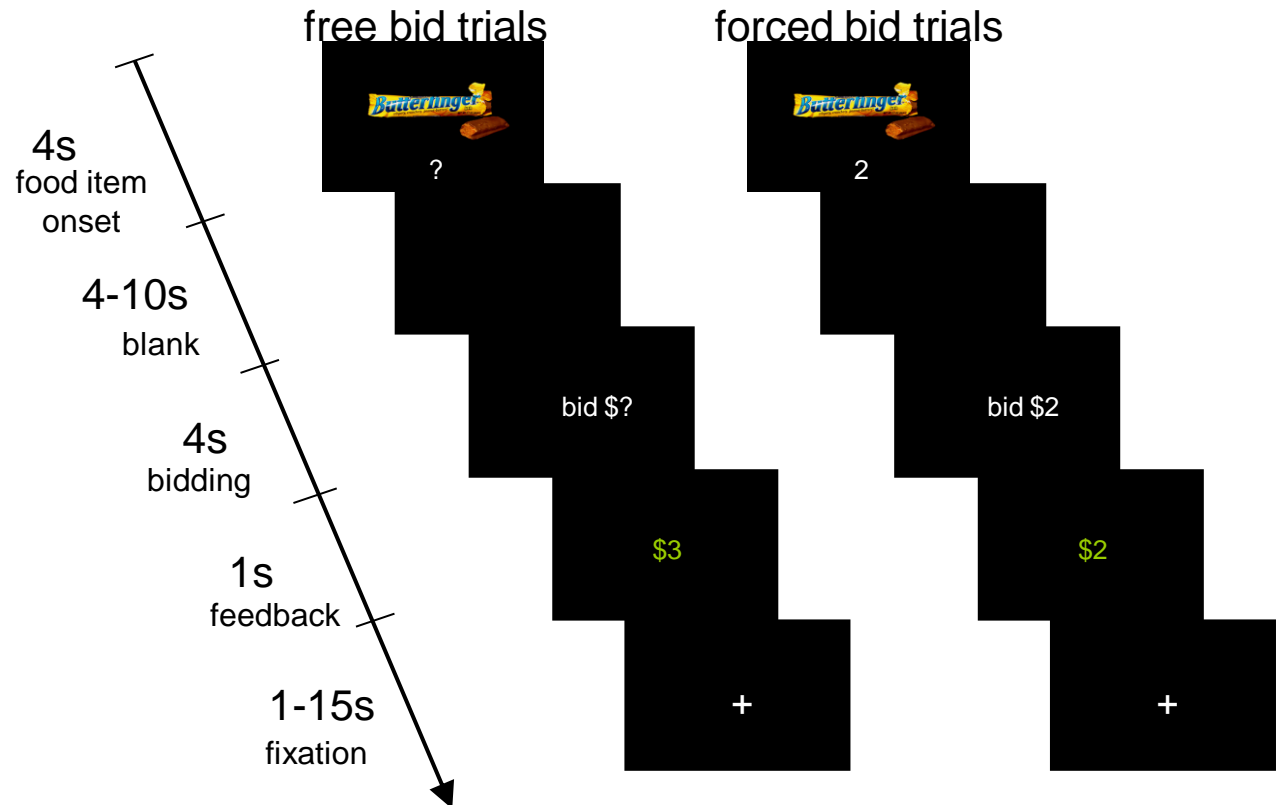
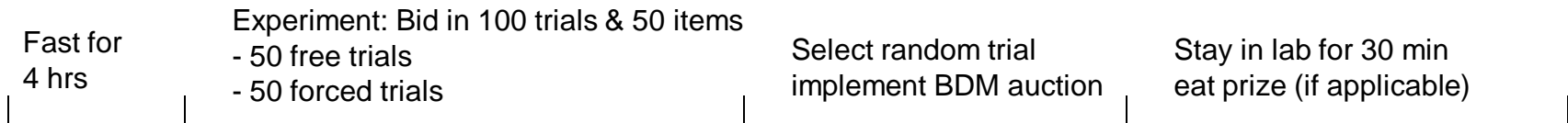
Useful conceptual framework



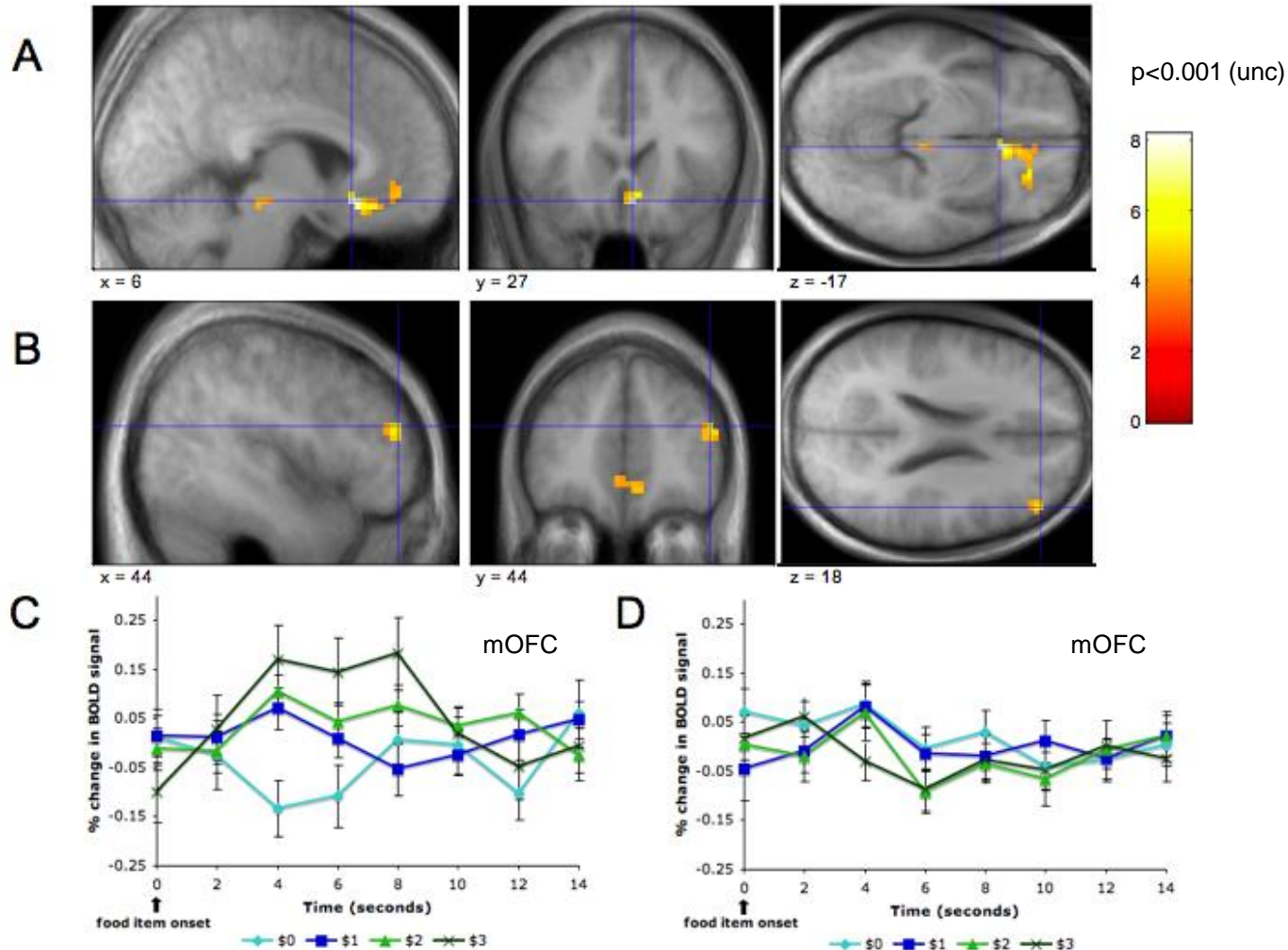
II Valuation

Experiment 1

JNeuro 2007, Plassmann O'Doherty Rangel



MAIN RESULT: mOFC and DLPFC encode for WTP in free trials, but not in forced trials



Experiment 2

JNeuro 2010, Plassmann O'Doherty Rangel

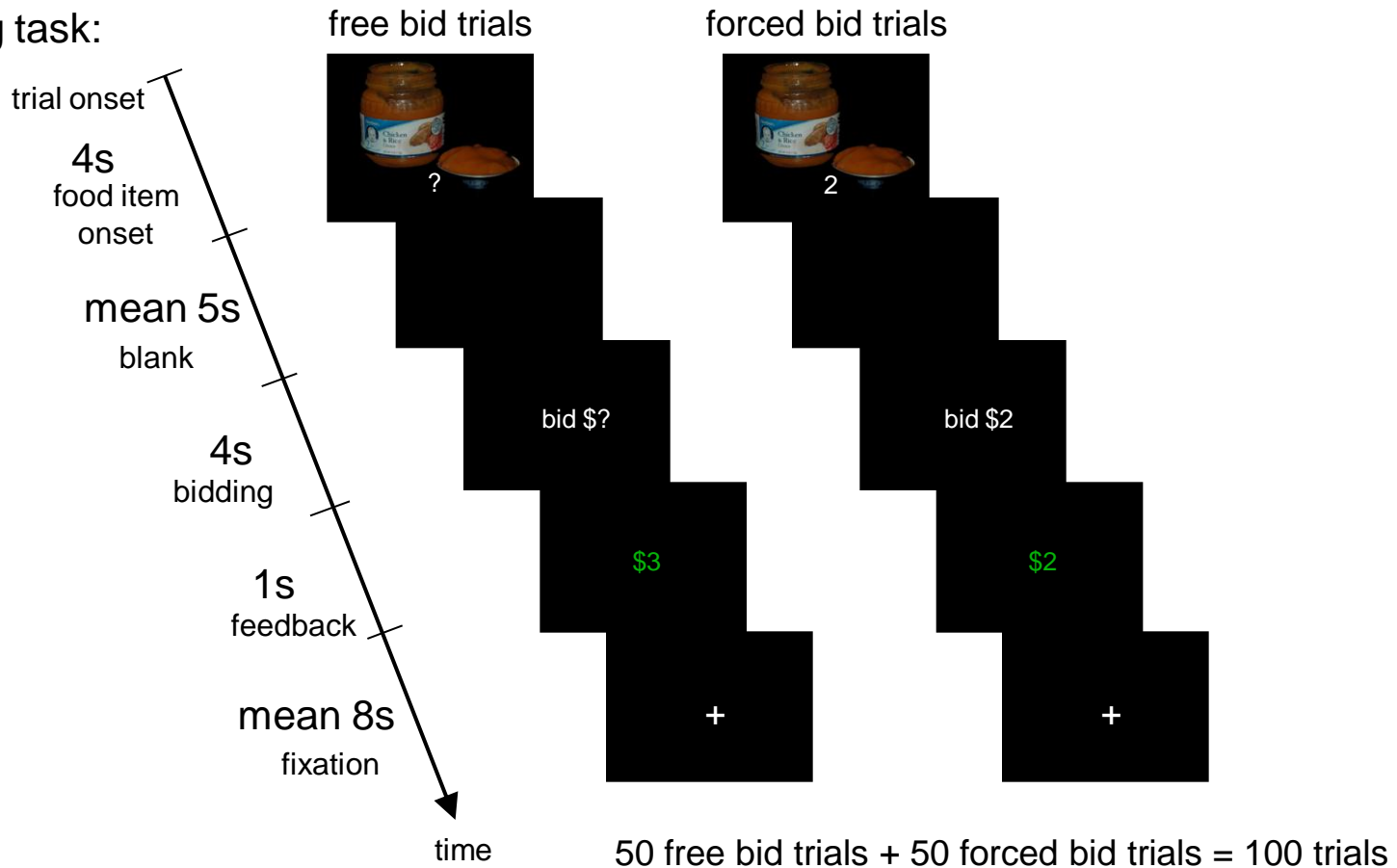
Subjects ate 2 hours
prior to the experiment

Bidding task
in fMRI scanner

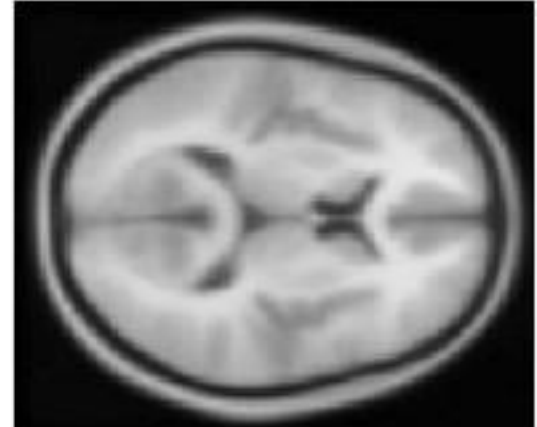
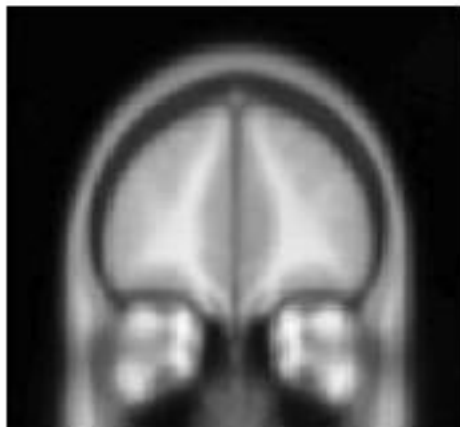
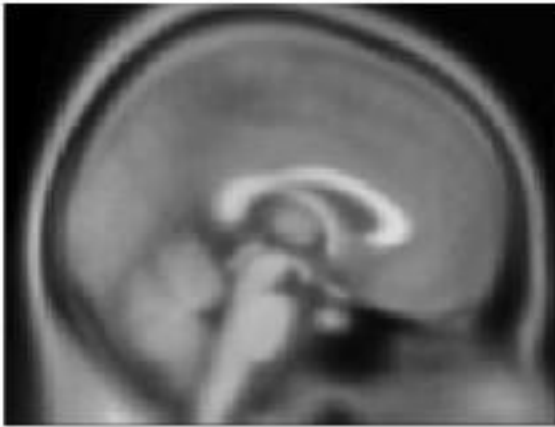
Select random trial
& implement auction

Stay in lab for 30 min
eat item (if applicable)

Bidding task:

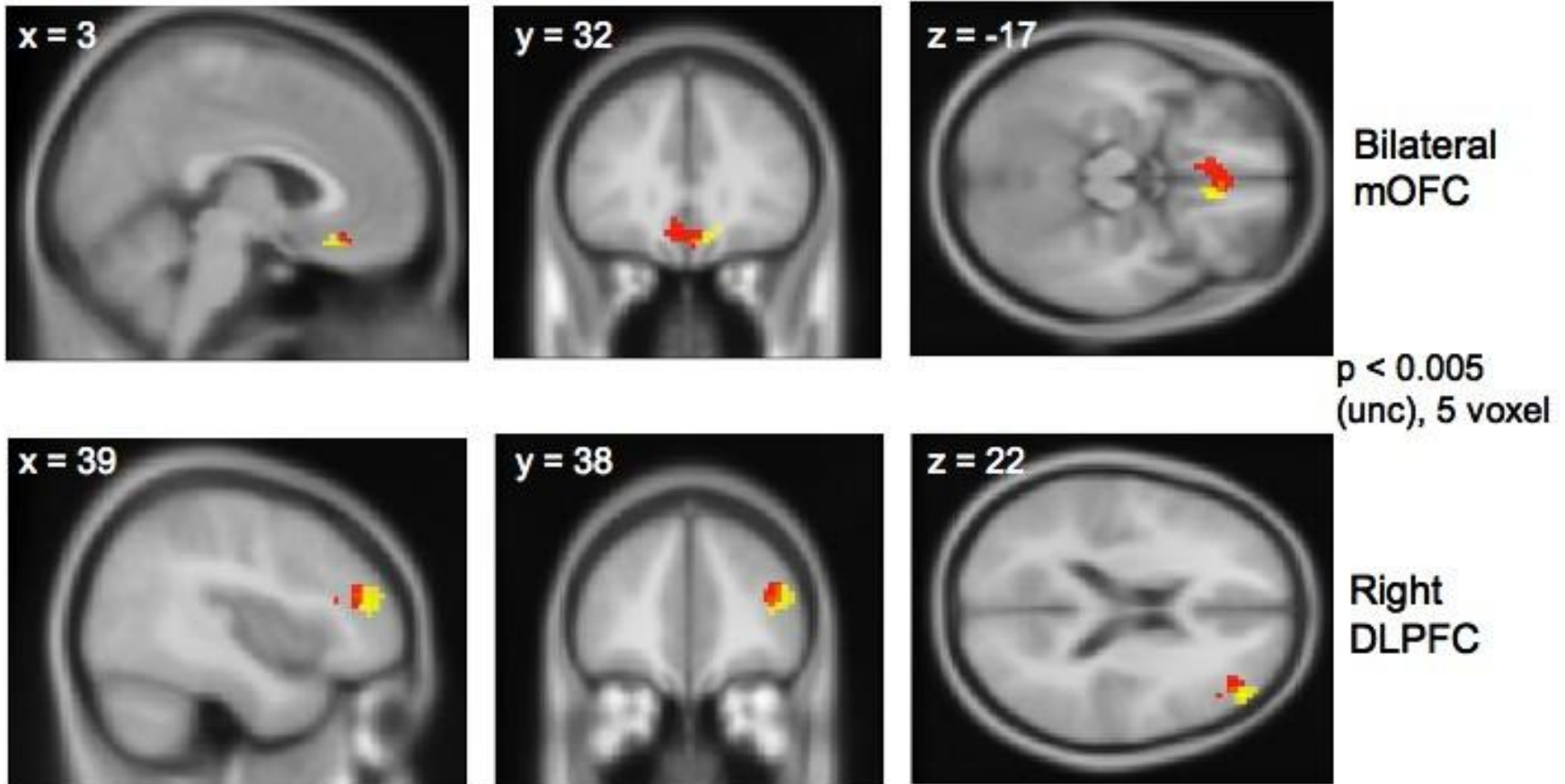


Areas with increased activity with bid (i.e., with aversive value)



$p < 0.01$ (uncor), 5 voxel

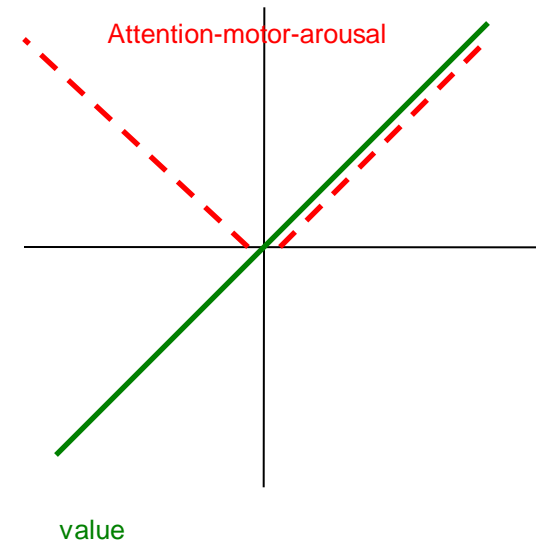
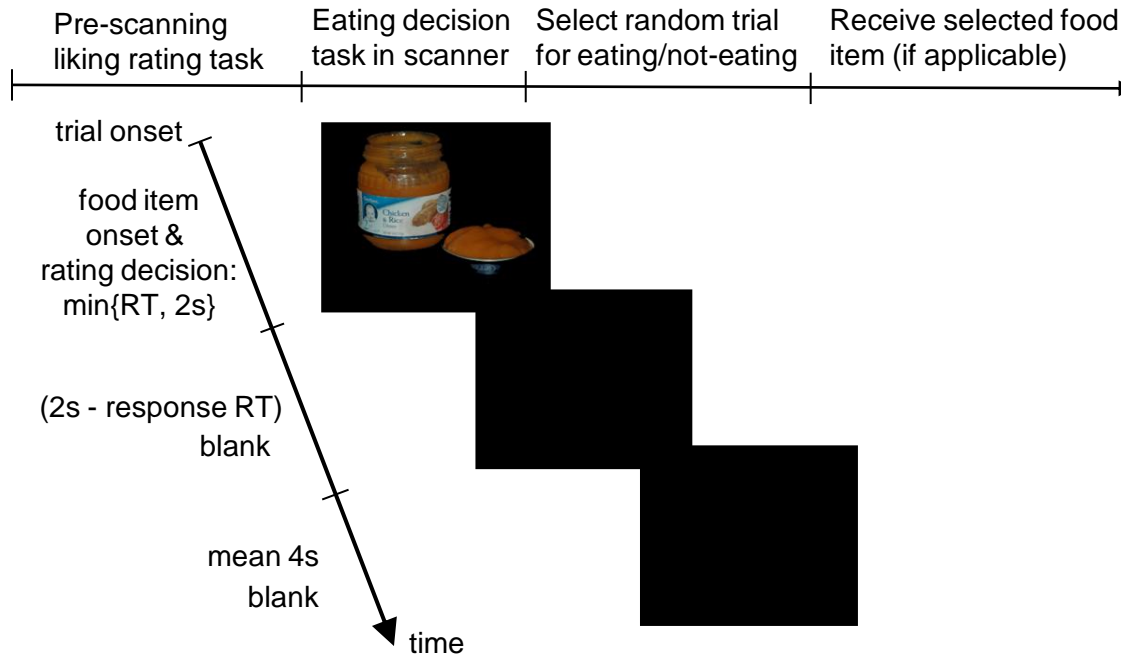
Conjunction of appetitive vs aversive goal value signals



- Increased activity changes: modulation by appetitive DVs in free bid trials
- Decreased activity changes: modulation by aversive DVs in free bid trials

Experiment 3

Cerebral Cortex 2010, Lit et al

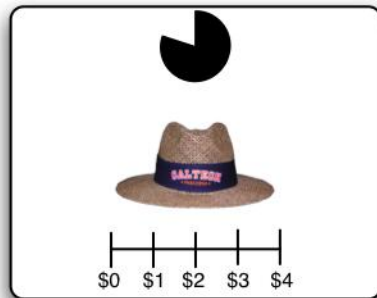


Experiment 3

Chib, O'Doherty, Rangel, JNeuro, 2009

Pre-scanning

free response time
BDM auction

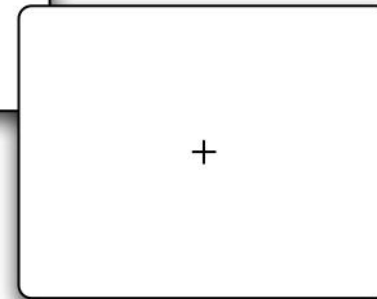


During fMRI Scanning

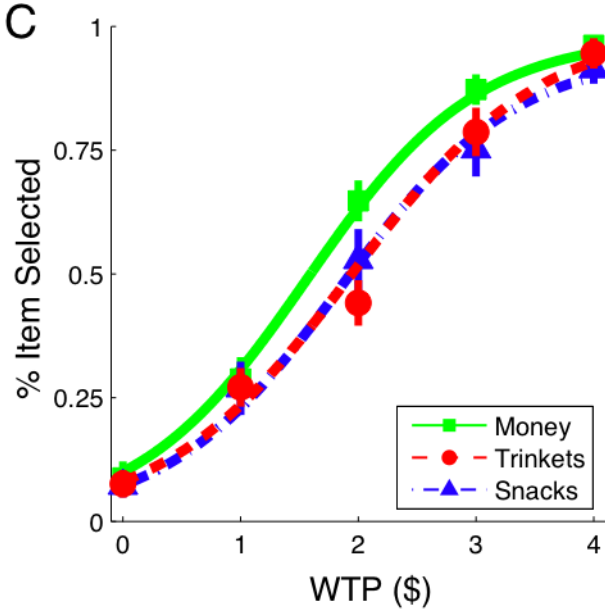
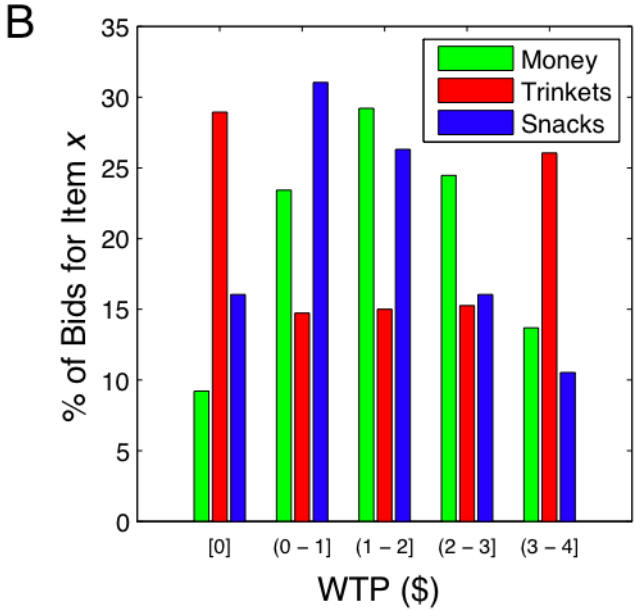
2 sec
binary choice



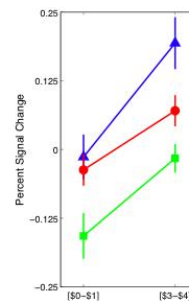
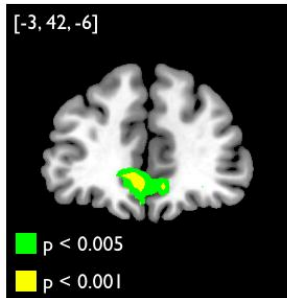
1-10 sec
fixation



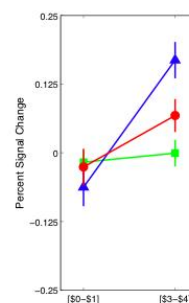
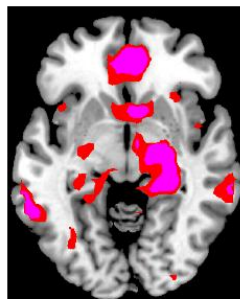
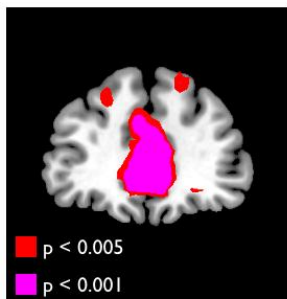
Behavior



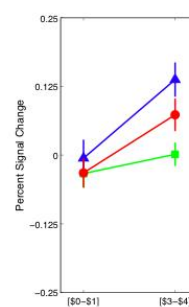
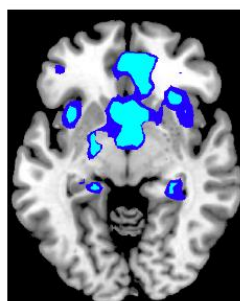
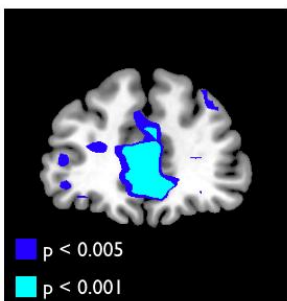
Money



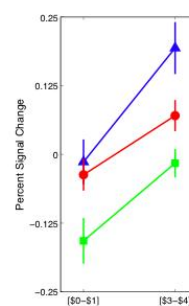
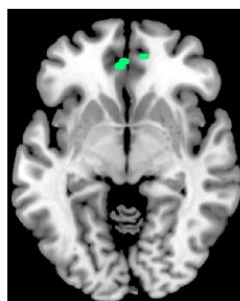
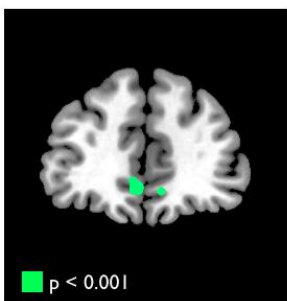
Trinkets



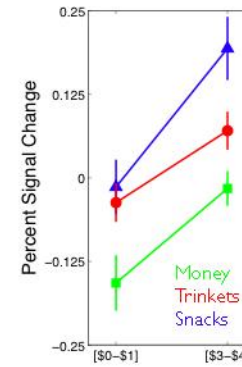
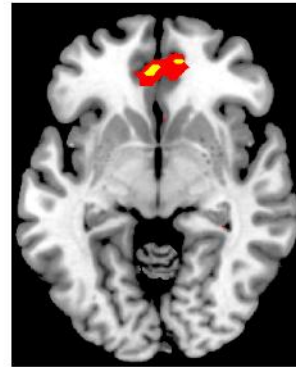
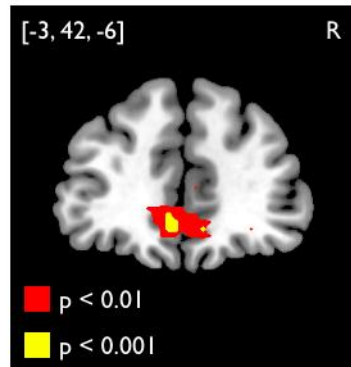
Snacks



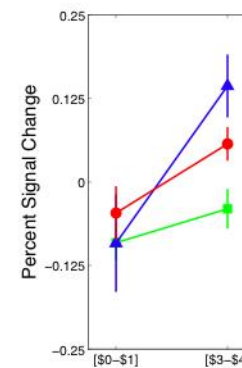
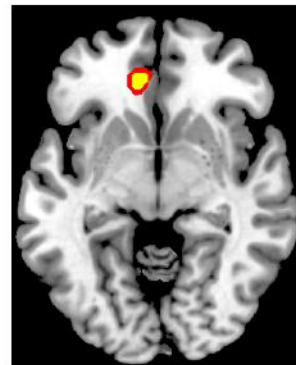
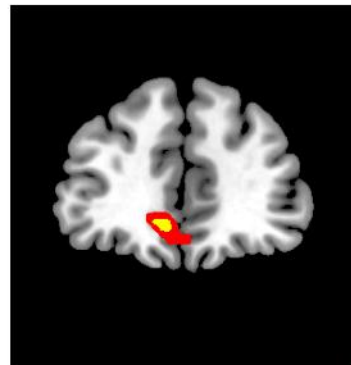
Commonly active areas



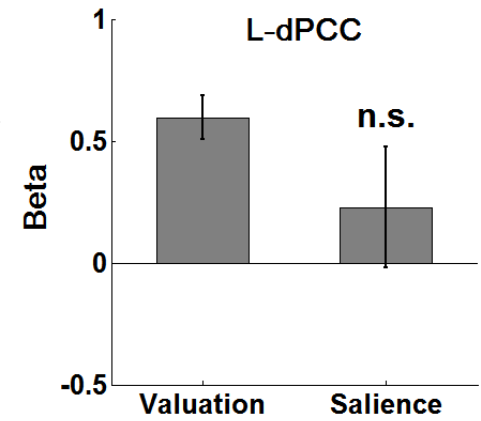
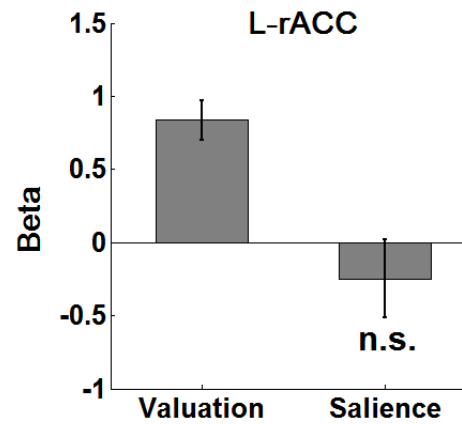
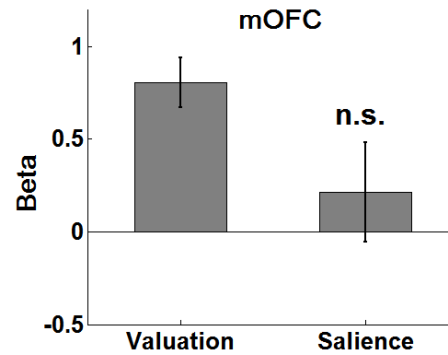
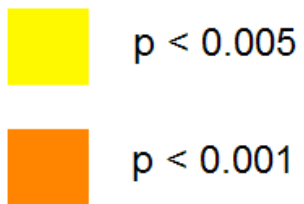
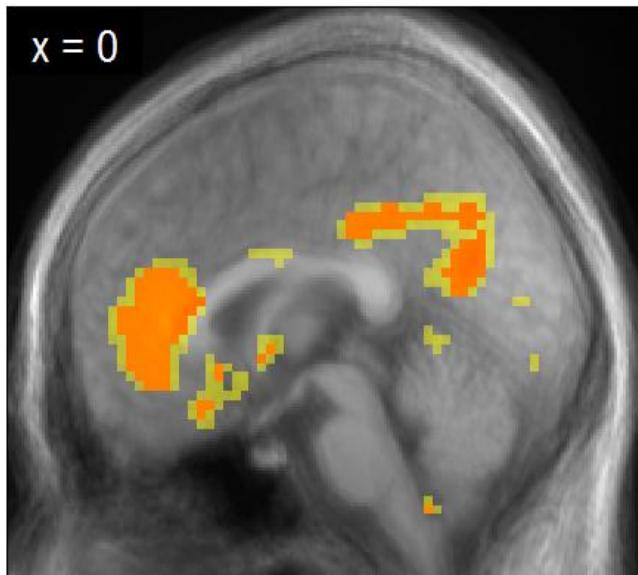
choices against a
fixed monetary bid



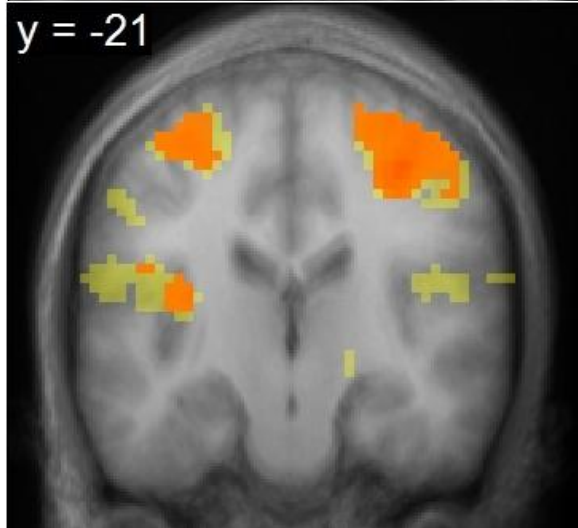
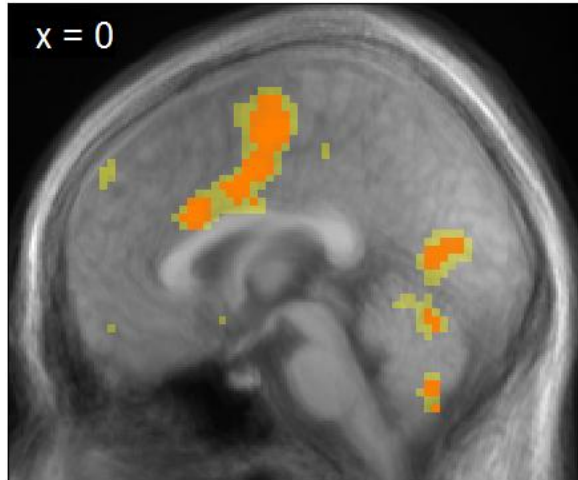
choices against a
fixed snack item



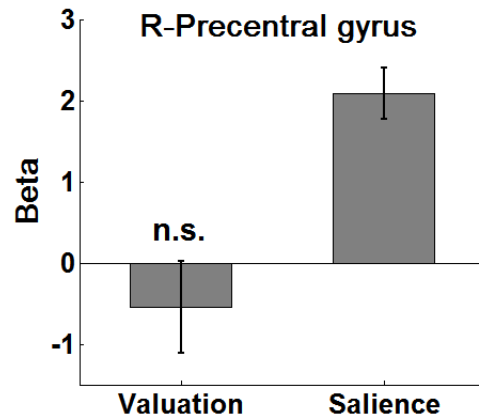
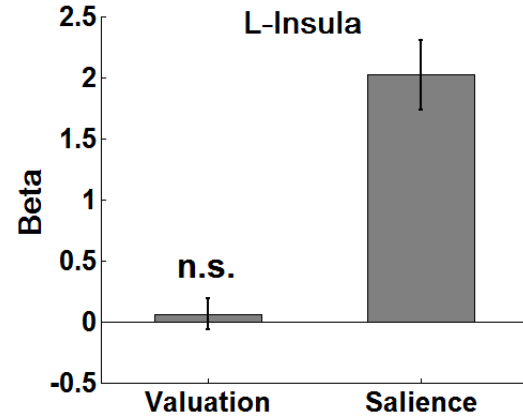
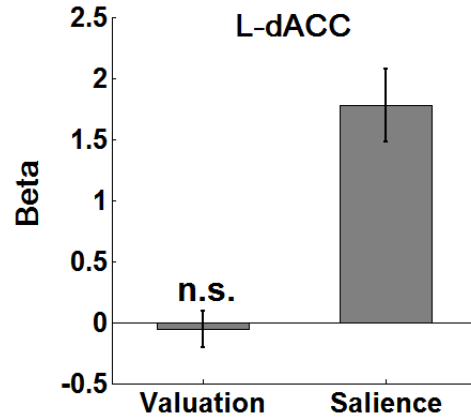
Value only activity



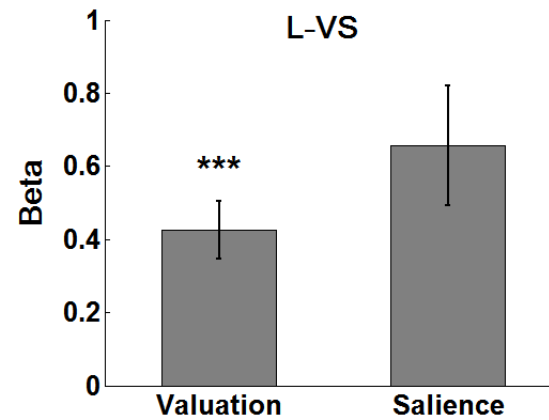
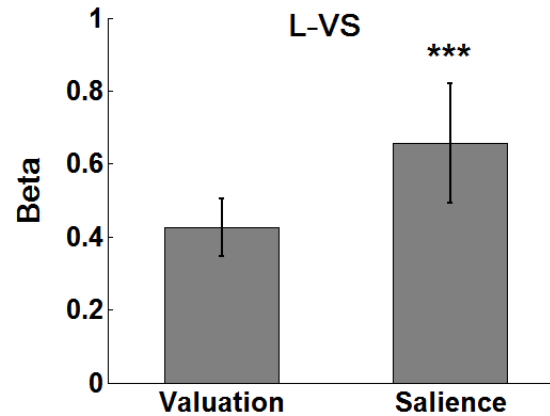
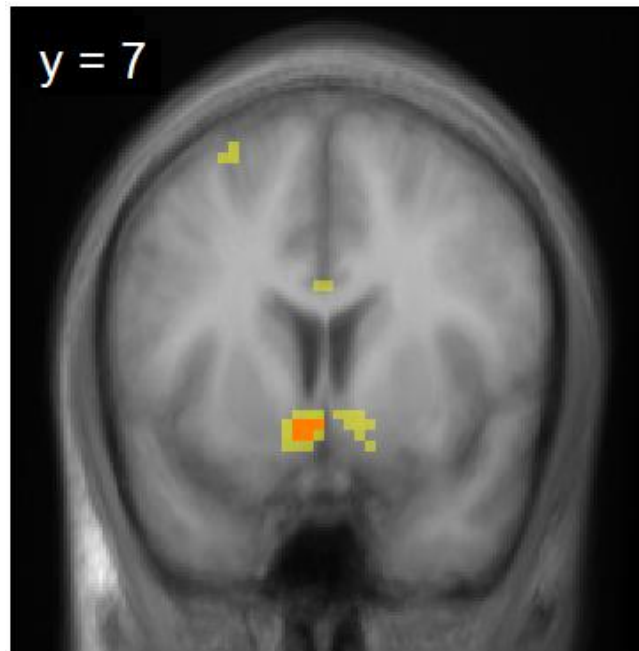
Saliency only activity



Yellow: $p < 0.005$ Orange: $p < 0.001$

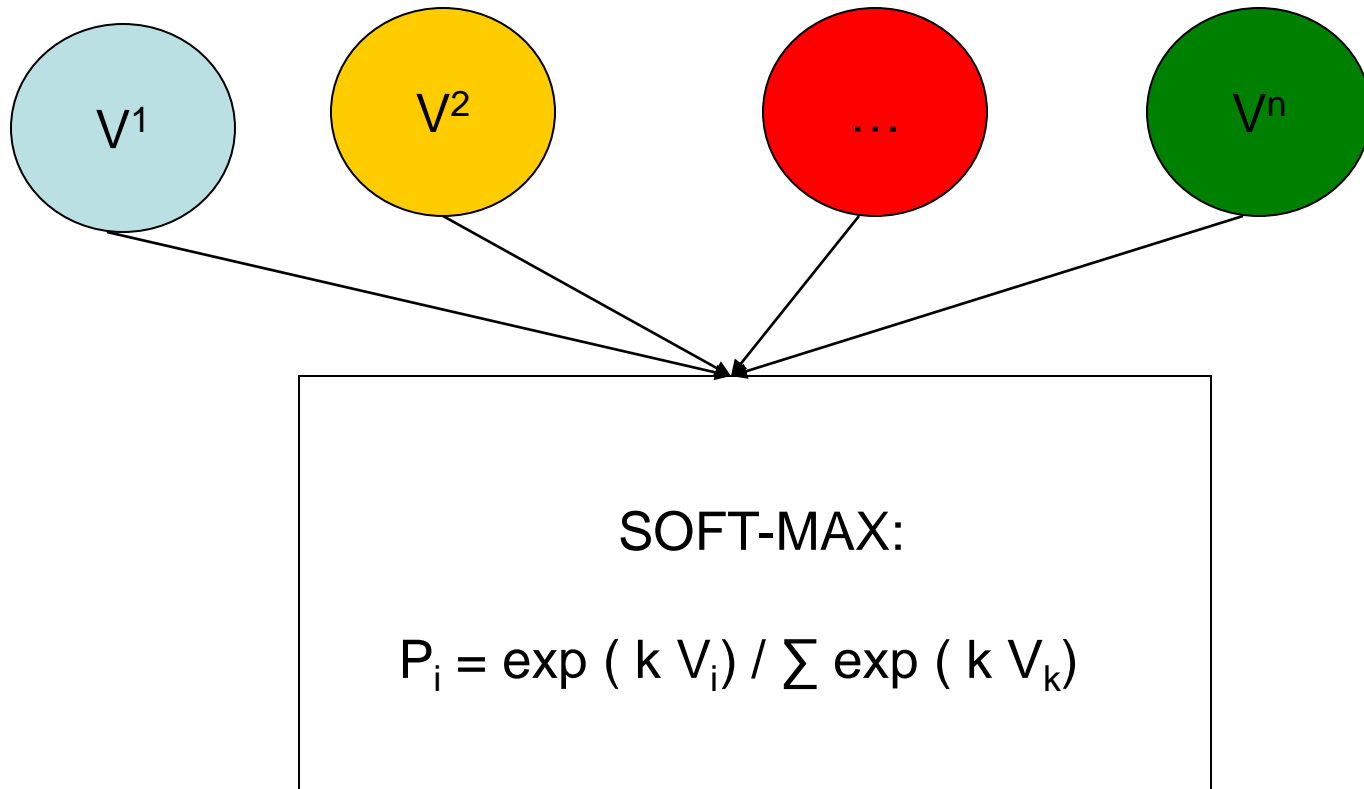


Value & saliency related activity

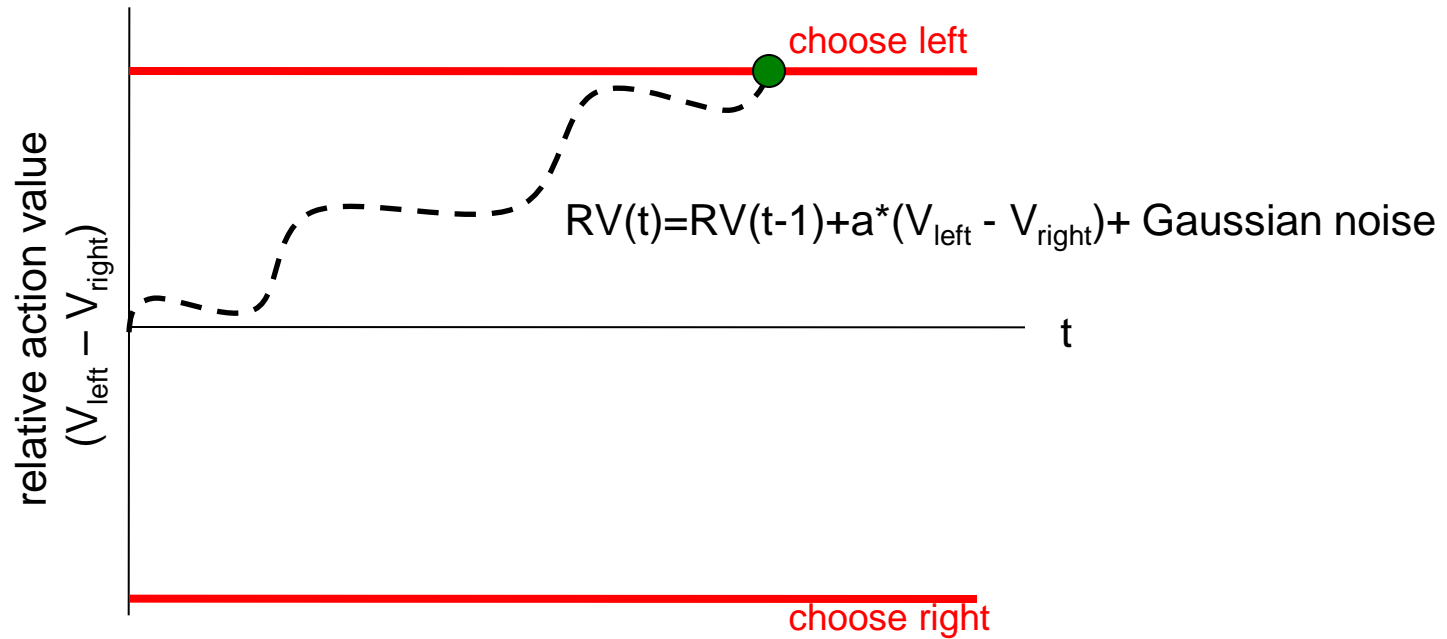


III Comparison

Common reduced from view

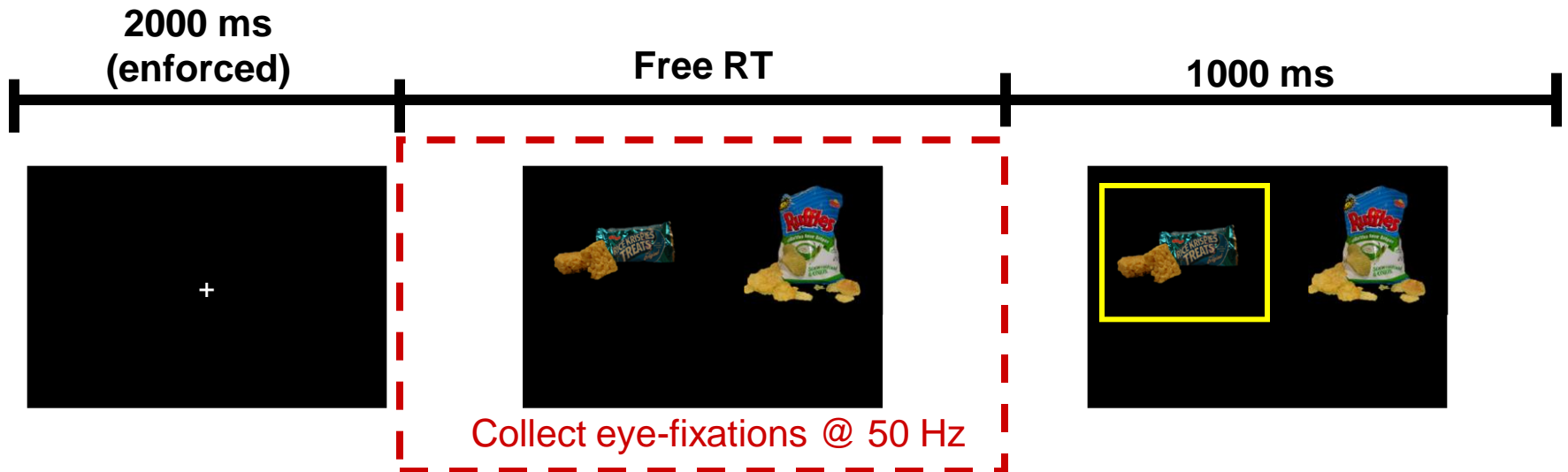
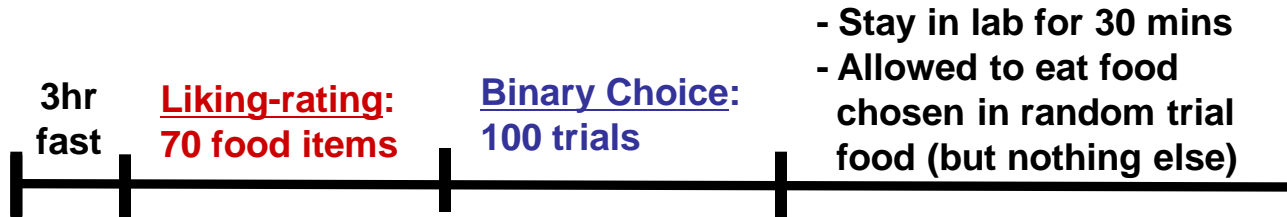


Ratcliff's Drift-Diffusion Model

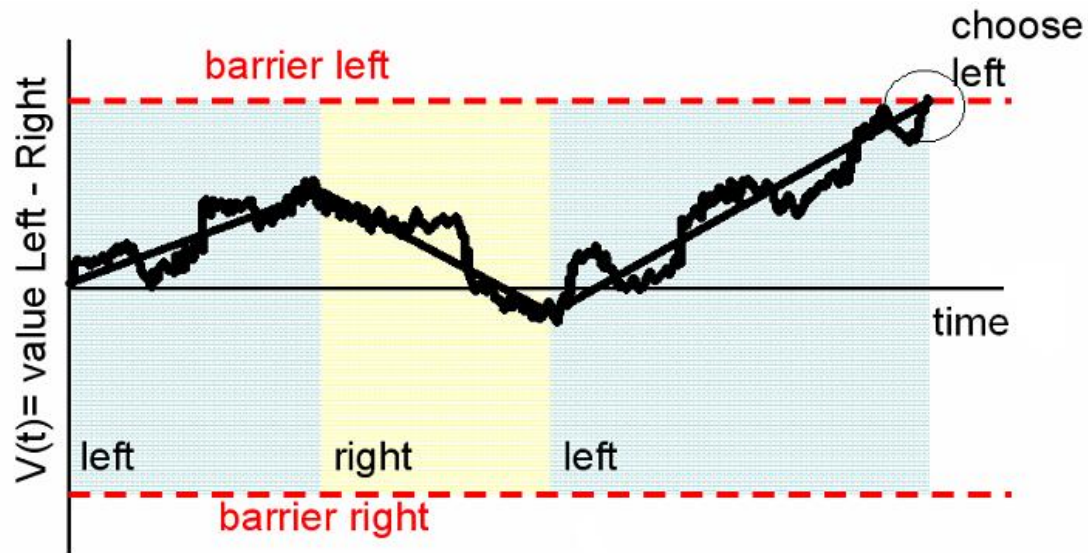


Experimental 4

Nature Neuro 2010, Krajbich Armel Rangel

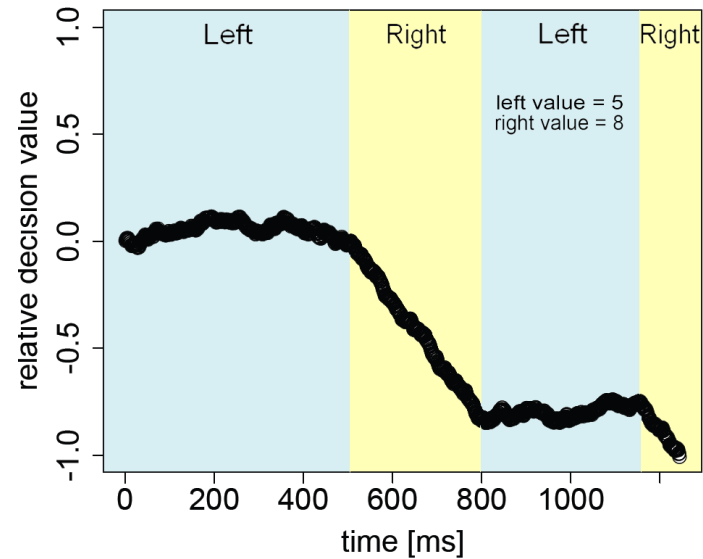
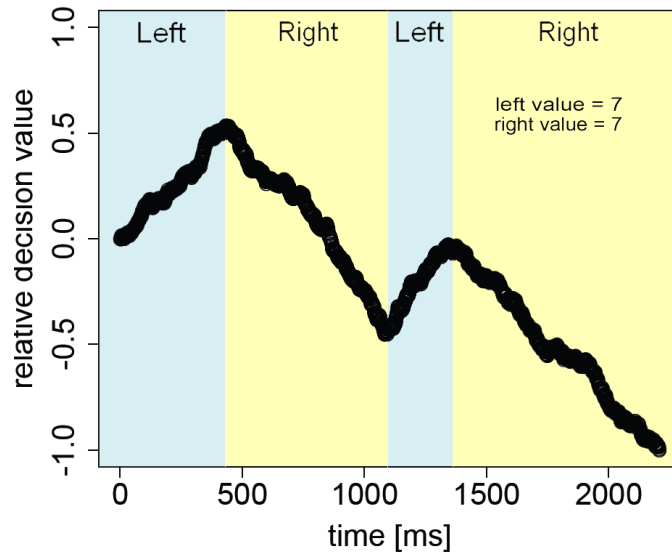


Computational model



$$V(t) = V(t-1) + a(v_{\text{target}} - \theta v_{\text{non-target}}) + u_t$$
$$u_t \sim N(0, s^2)$$

Examples of simulations



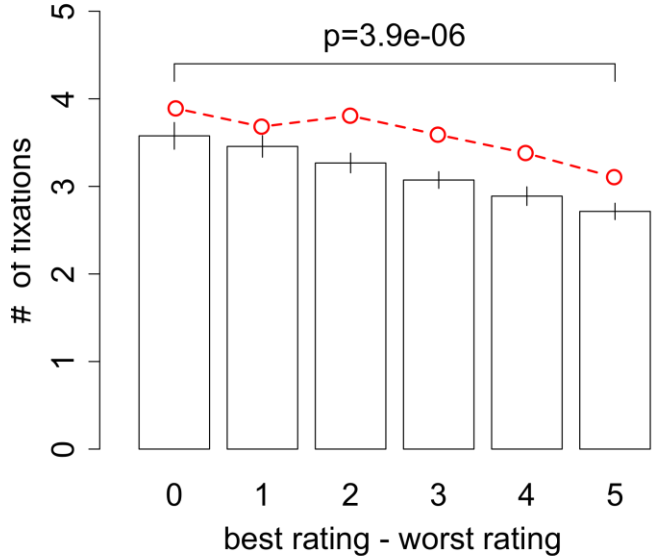
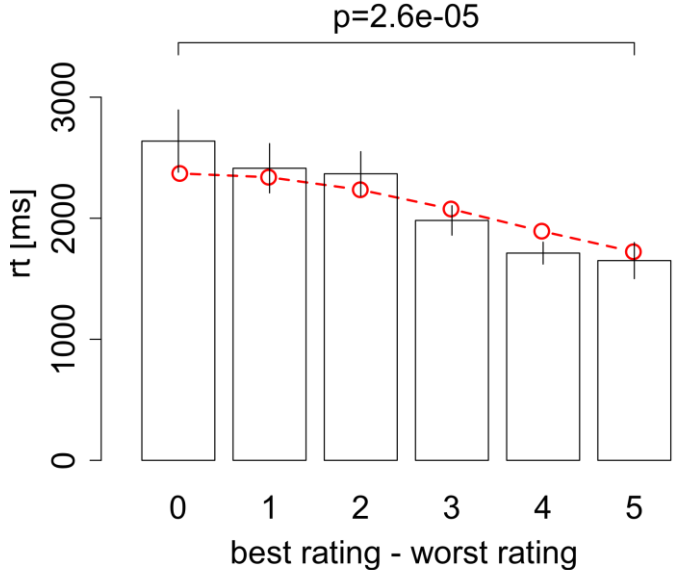
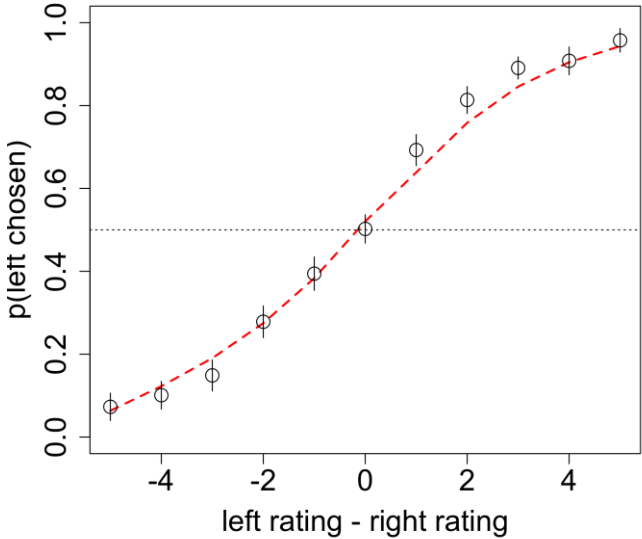
Key features:

- Fixation lengths drawn from common distribution
- Integrator follows a random walk with slope $r_{\text{target}} - 0.3r_{\text{non-target}}$

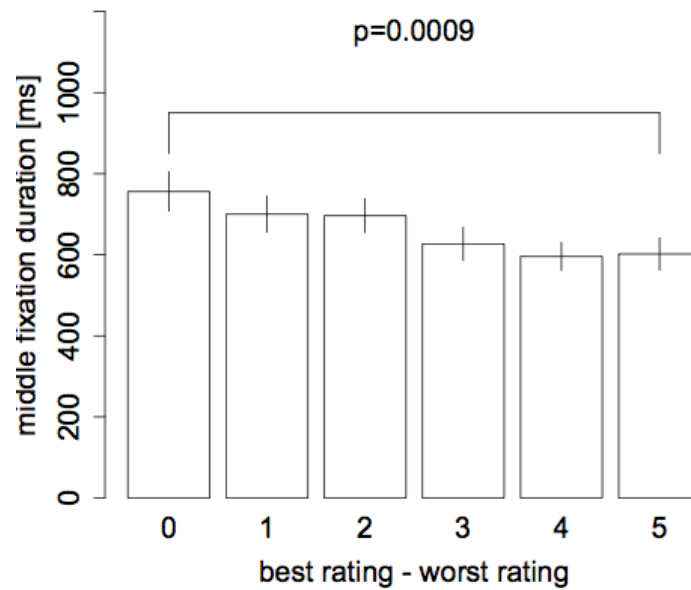
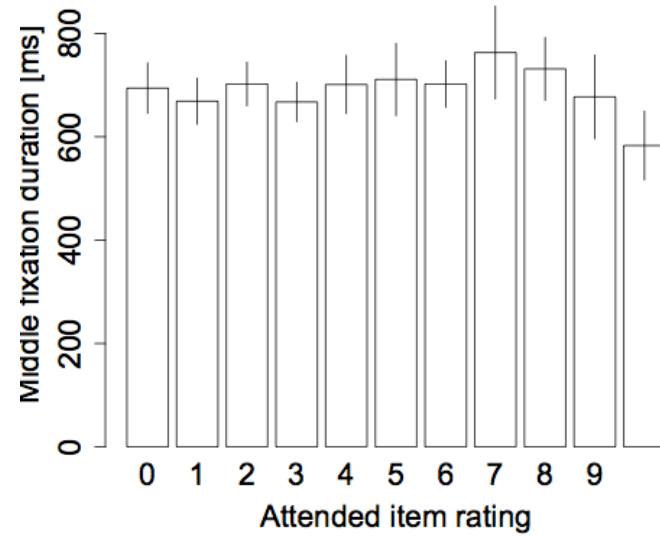
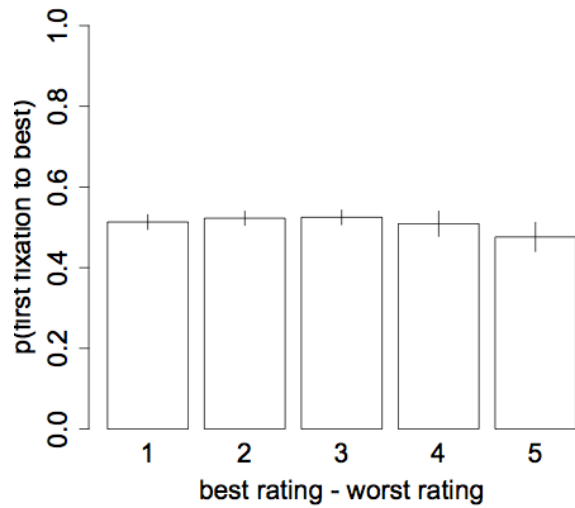
Estimation-prediction exercise

- Free model parameters:
 - a = slope of integration
 - s^2 = noise variance
 - θ = attentional bias
- Estimate parameters in even trials using ML
Match: choices and reaction times
- Simulate model in odd trials

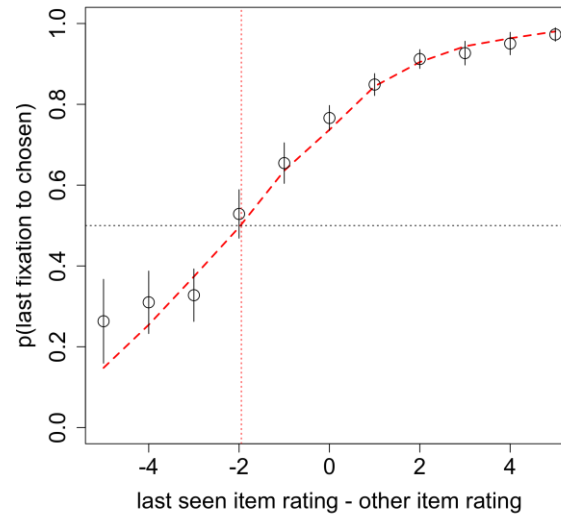
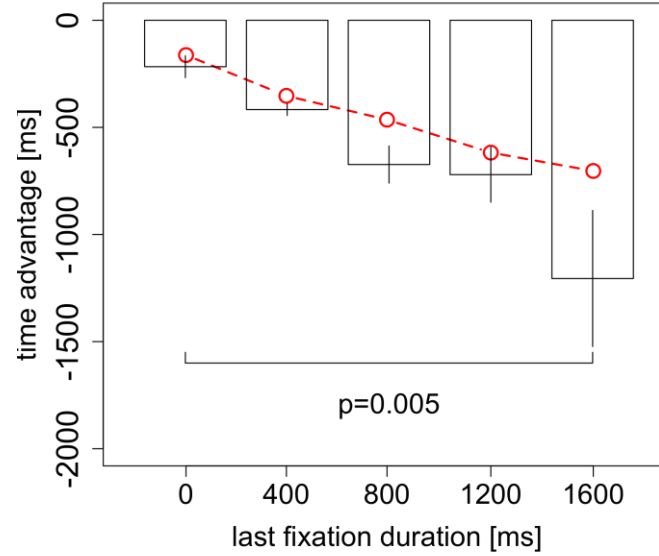
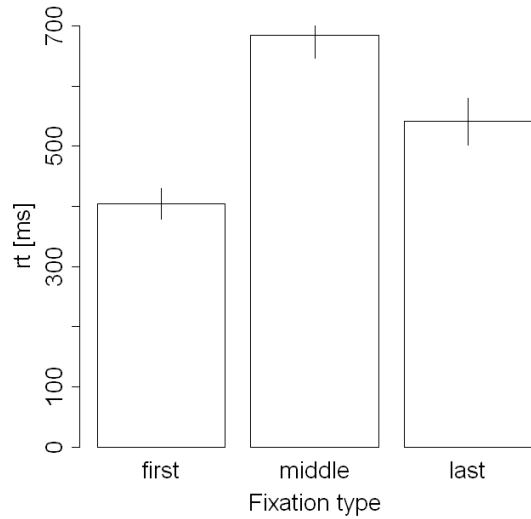
Basic psychometrics



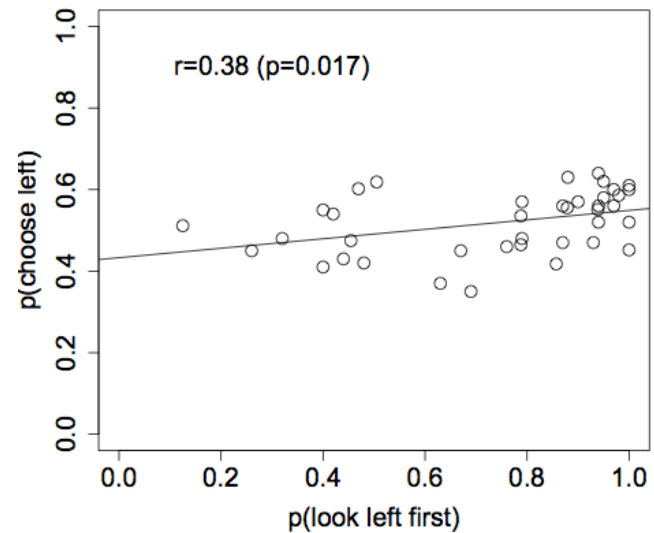
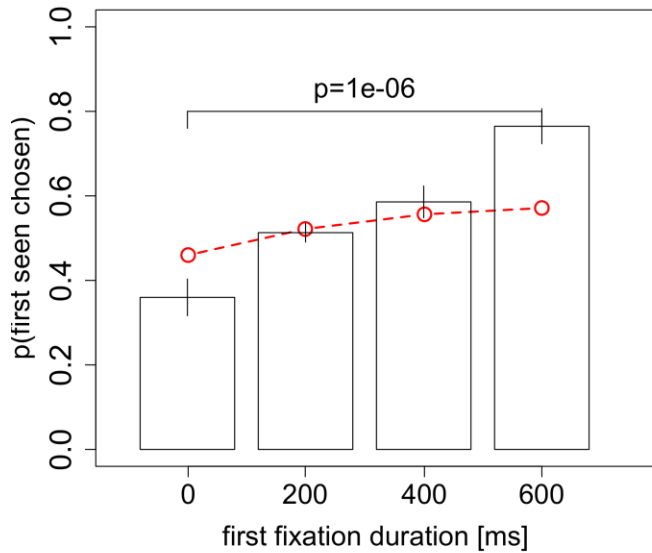
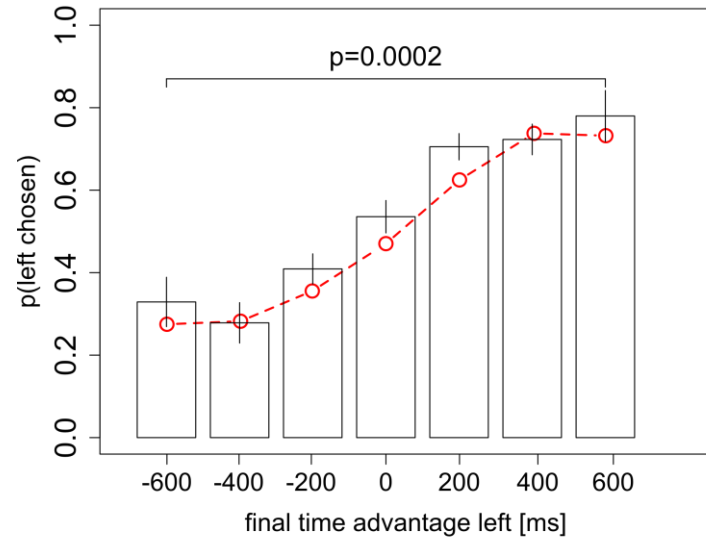
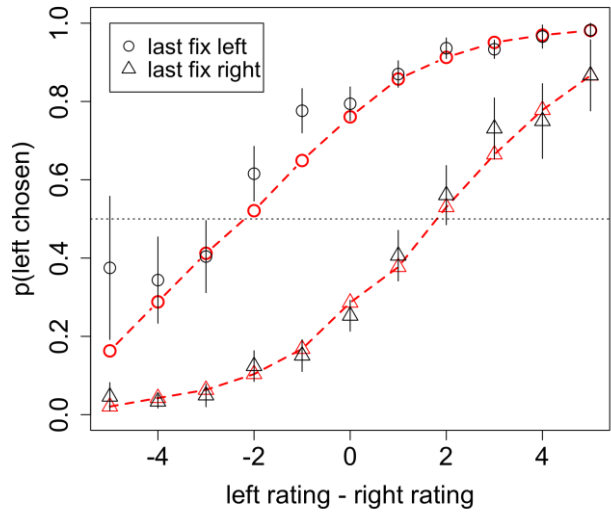
Basic fixation patterns



Key tests of the model

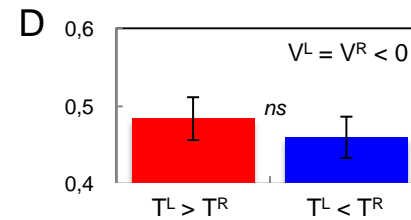
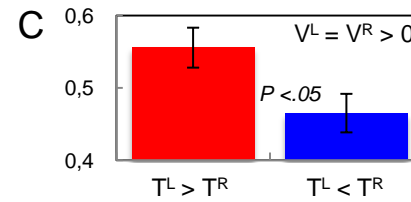
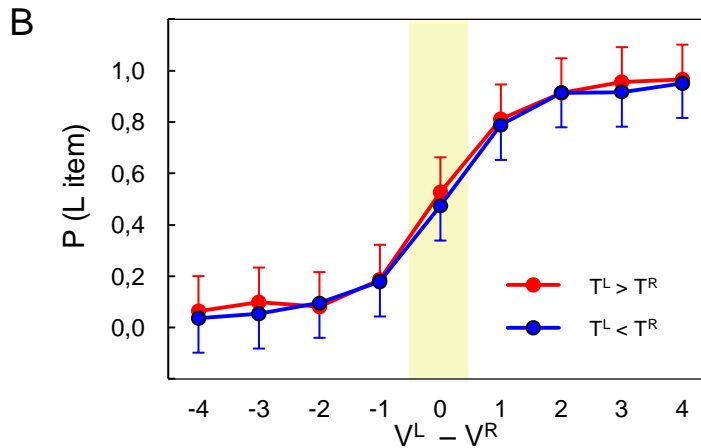
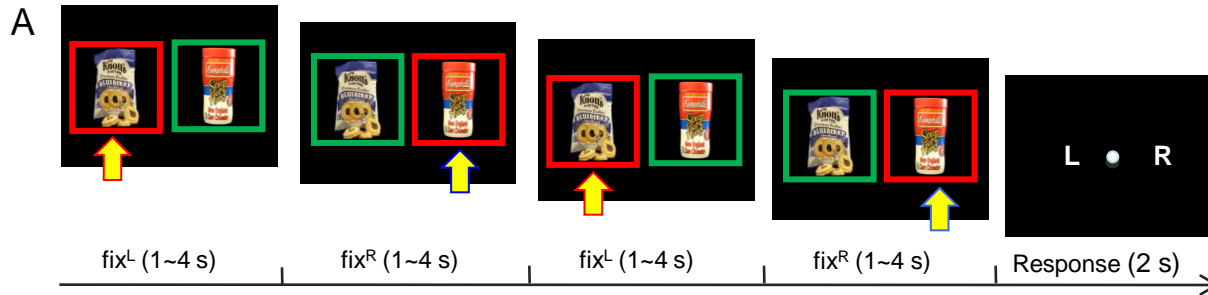


Predicted choice biases

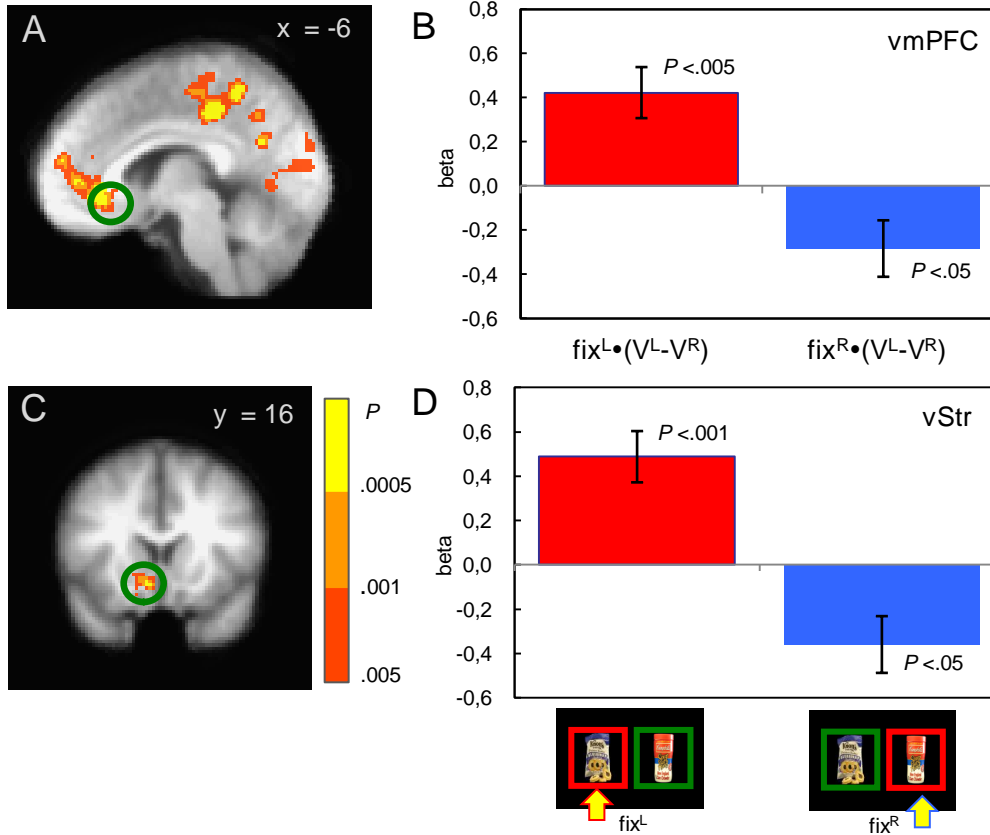


Experiment 5

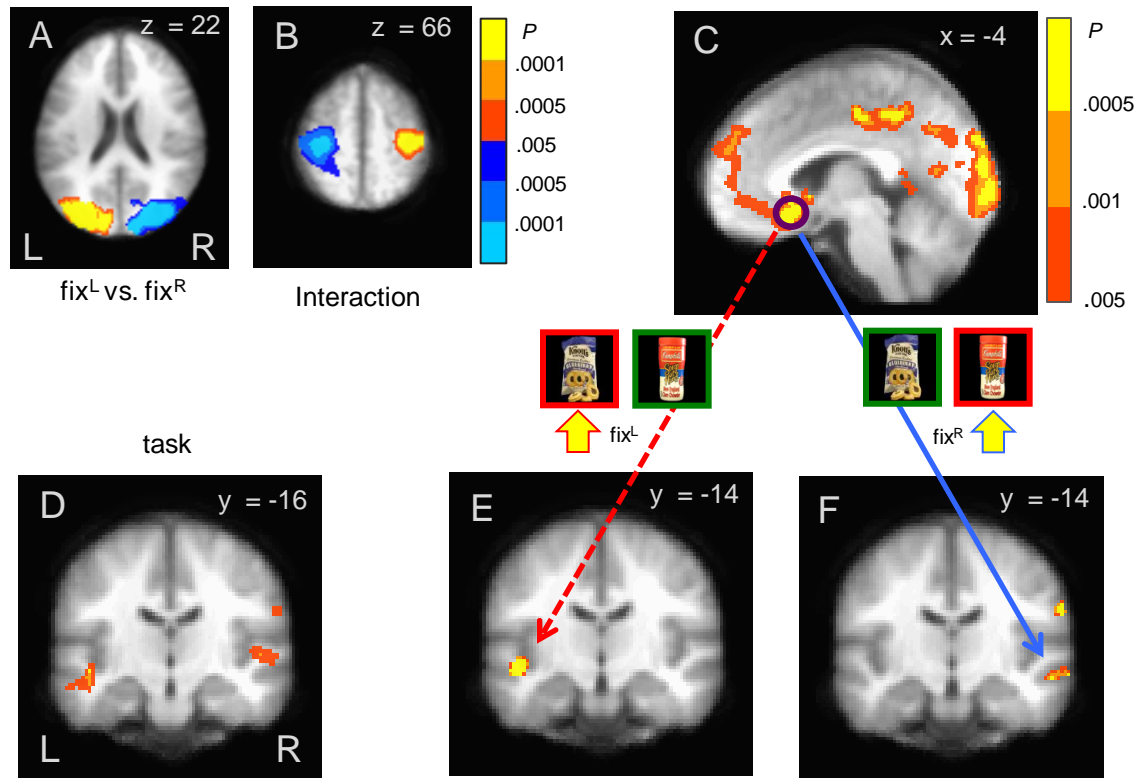
under review, Seung O'Doherty Rangel



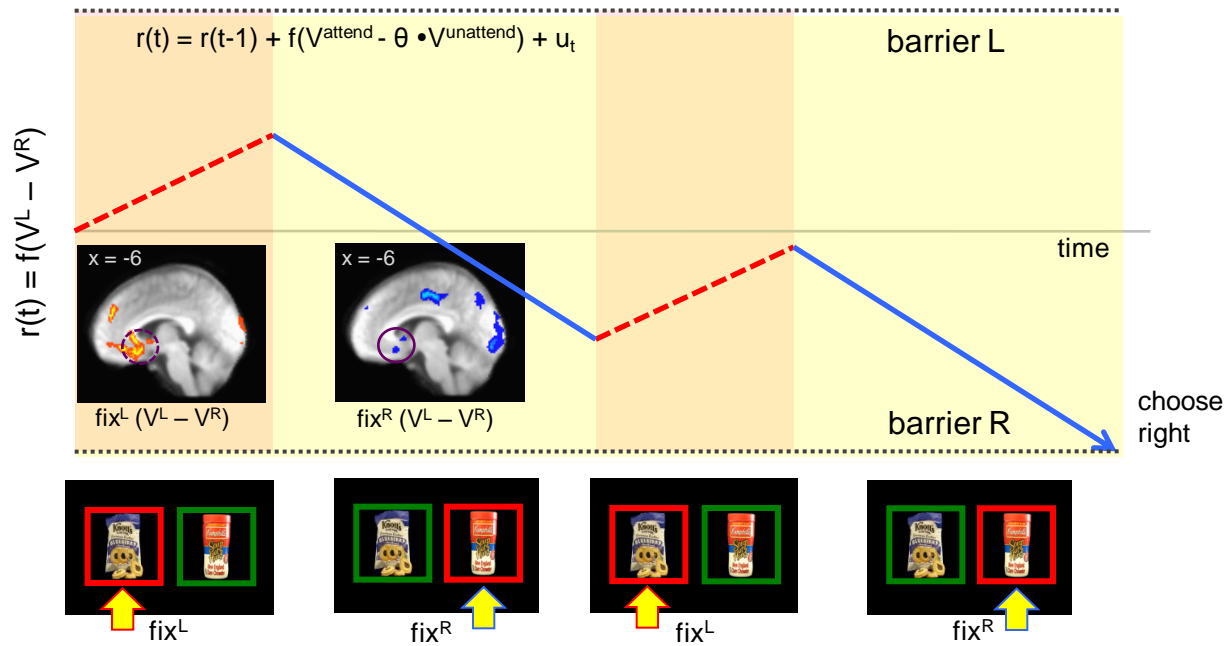
mOFC encodes attention modulated relative value signals



Attentional effect modulated by the STS



Relationship with computational model



IV

From choices to motor output

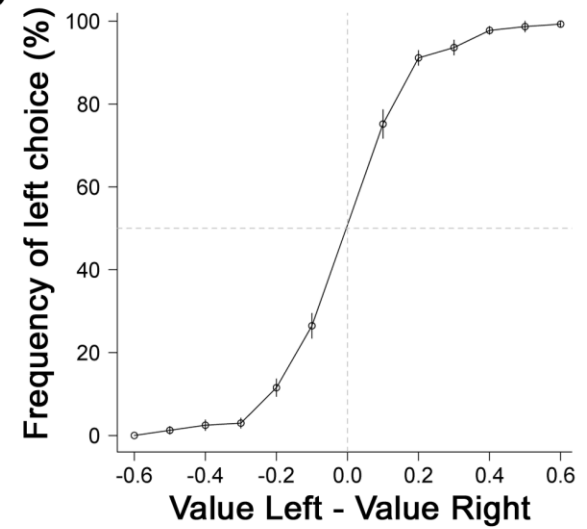
Experiment 6

under review, Hare O'Doherty Schulz Rangel

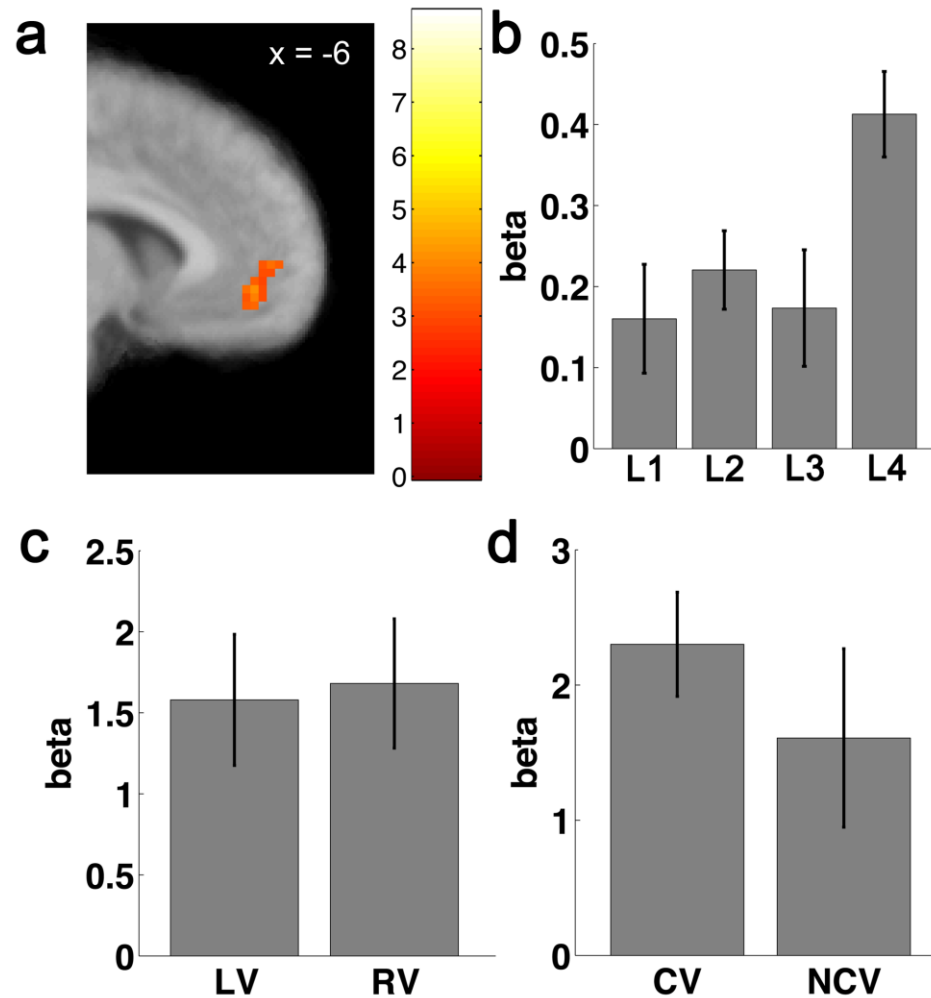
a



b



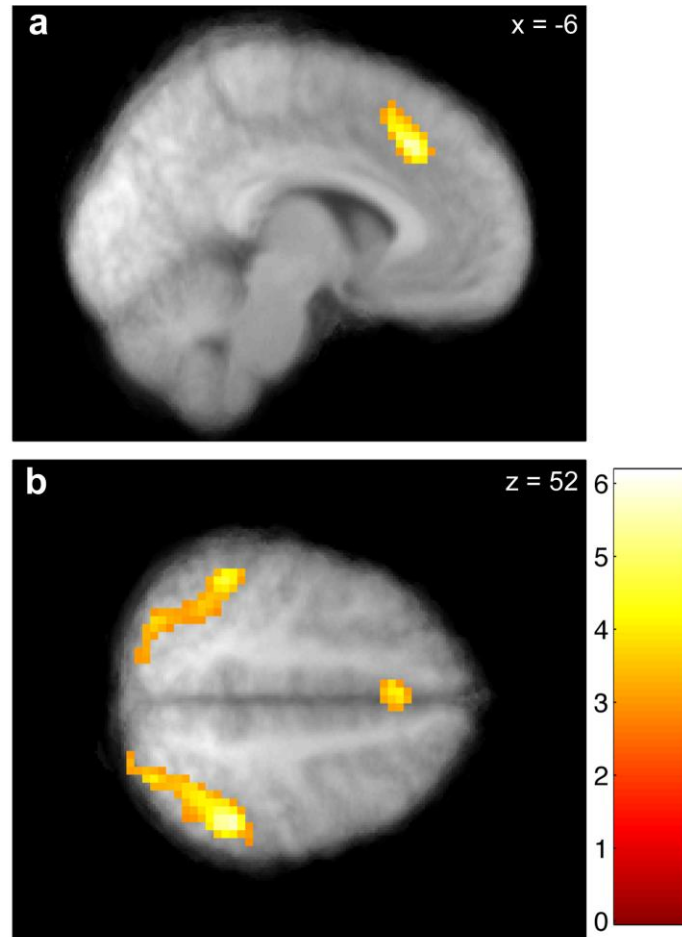
mOFC correlates with stimulus values



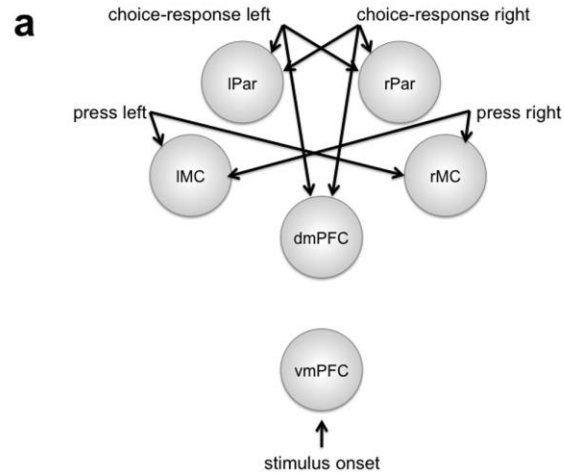
Markers of a region involved in comparison

1. Should exhibit aggregate activation pattern consistent with predictions of plausible neural implementations of the DDM
2. Should exhibit connectivity w/ vmPFC valuation areas at time of choice
3. Should exhibit choice dependent connectivity with motor cortex output areas

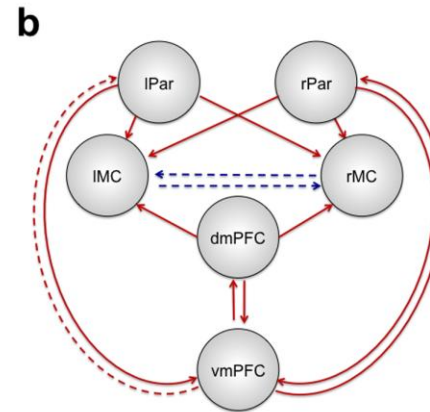
dmPFC activity correlates with predictions
of simple neural implementation of
best fitting DDM



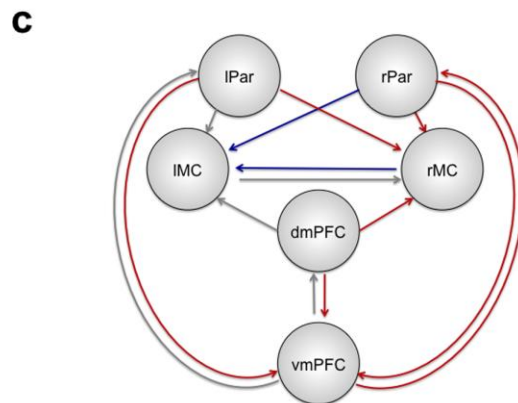
dmPFC modulates transformation of values into motor responses



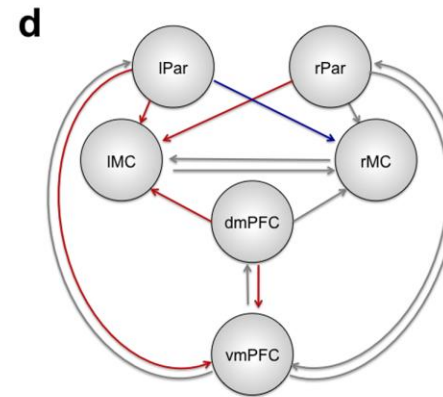
Driving inputs



Intrinsic connections



Choose left



Choose right

V
Self-control

Neural mechanisms of dietary self-control

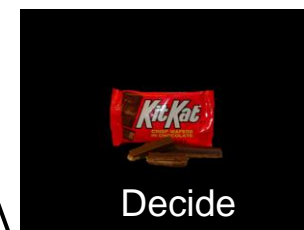
Hare, Camerer, Rangel (Science 2009)

Health Session

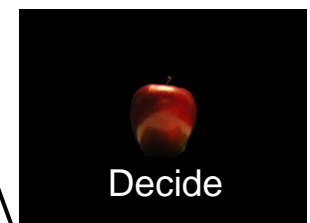
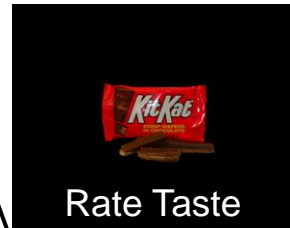
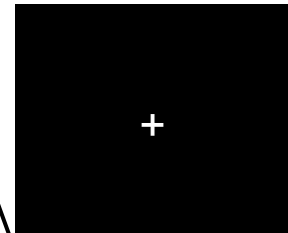
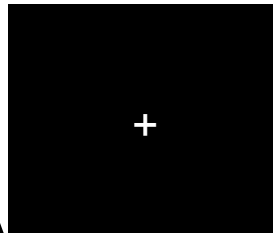
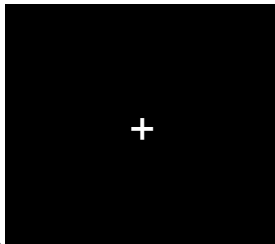
↔ Taste Session

Decision Session

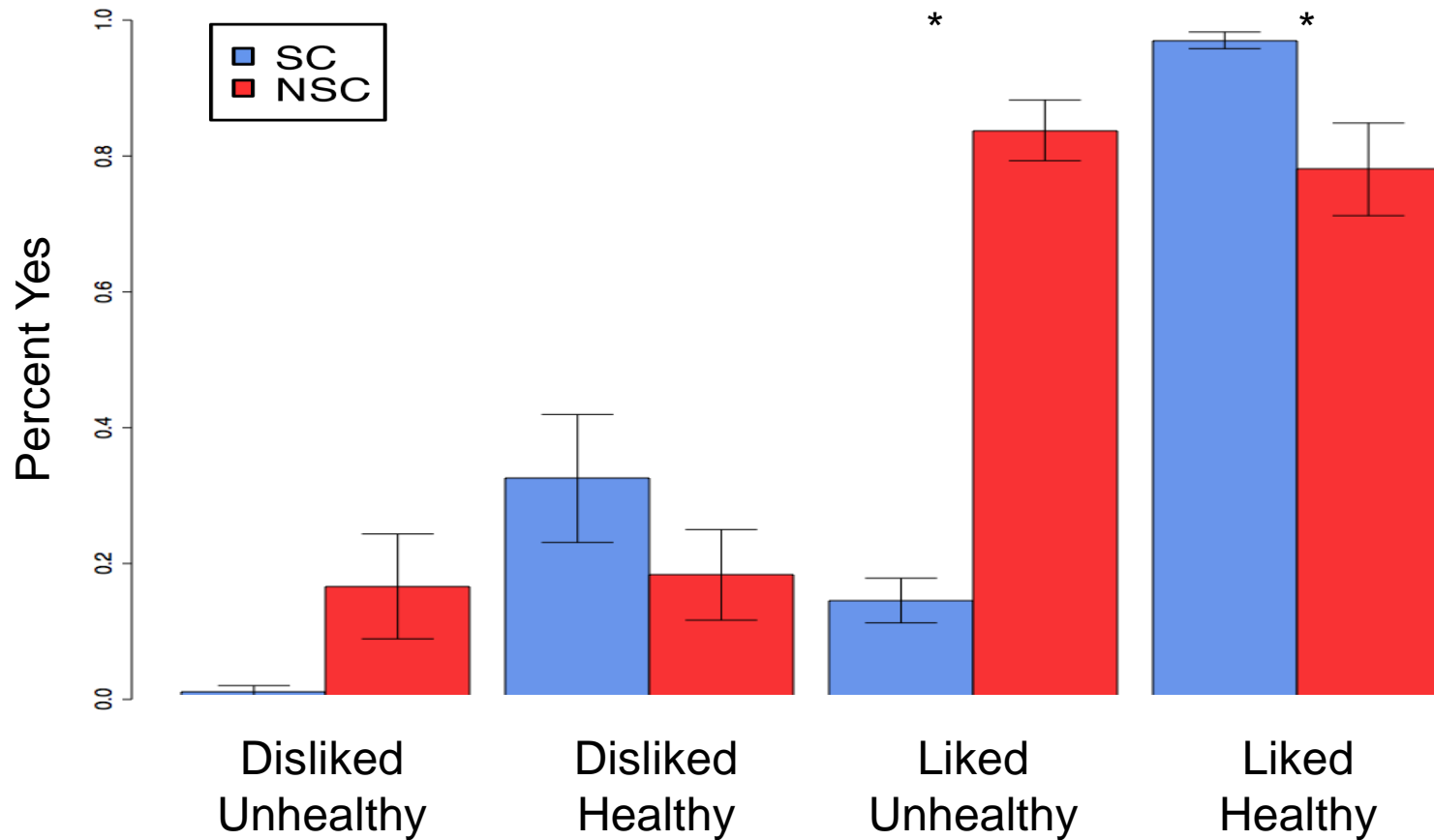
4s
food item
presentation



?-?s
fixation



Behavioral differences across groups

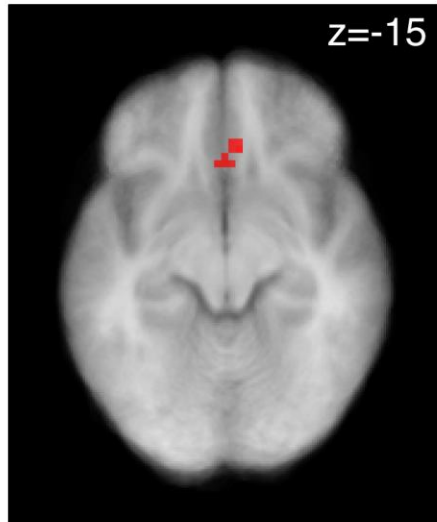
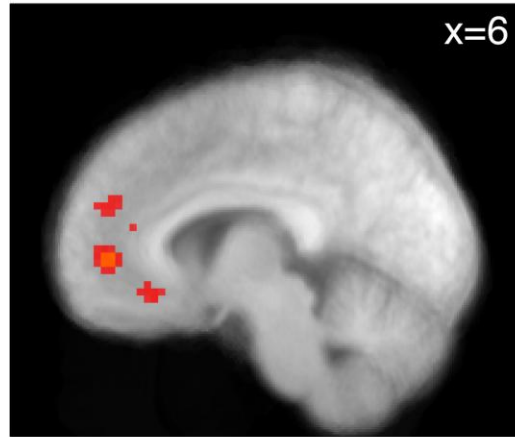
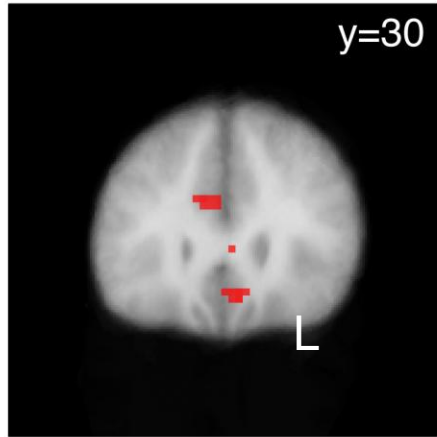


Hypotheses

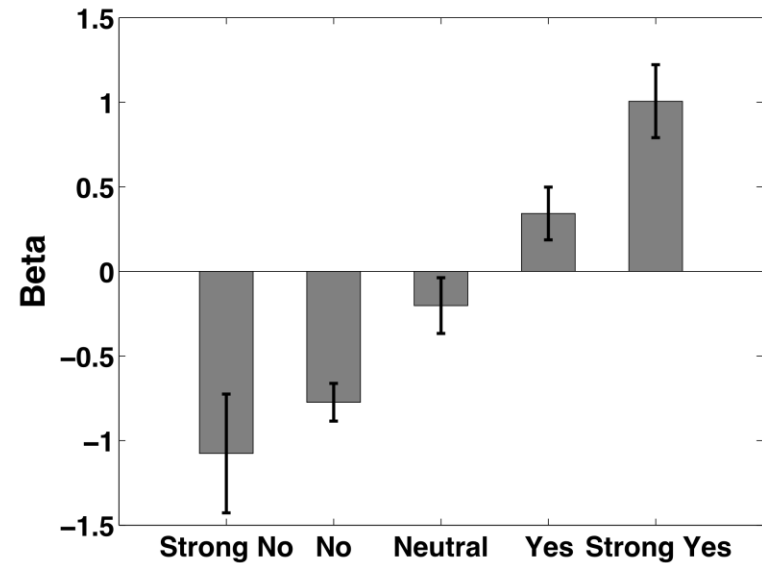
H1) vmPFC encodes a common decision value signal that has different properties in good and poor self-controllers

H2) Attentional self-control involves DLPFC modulation of the vmPFC valuation system

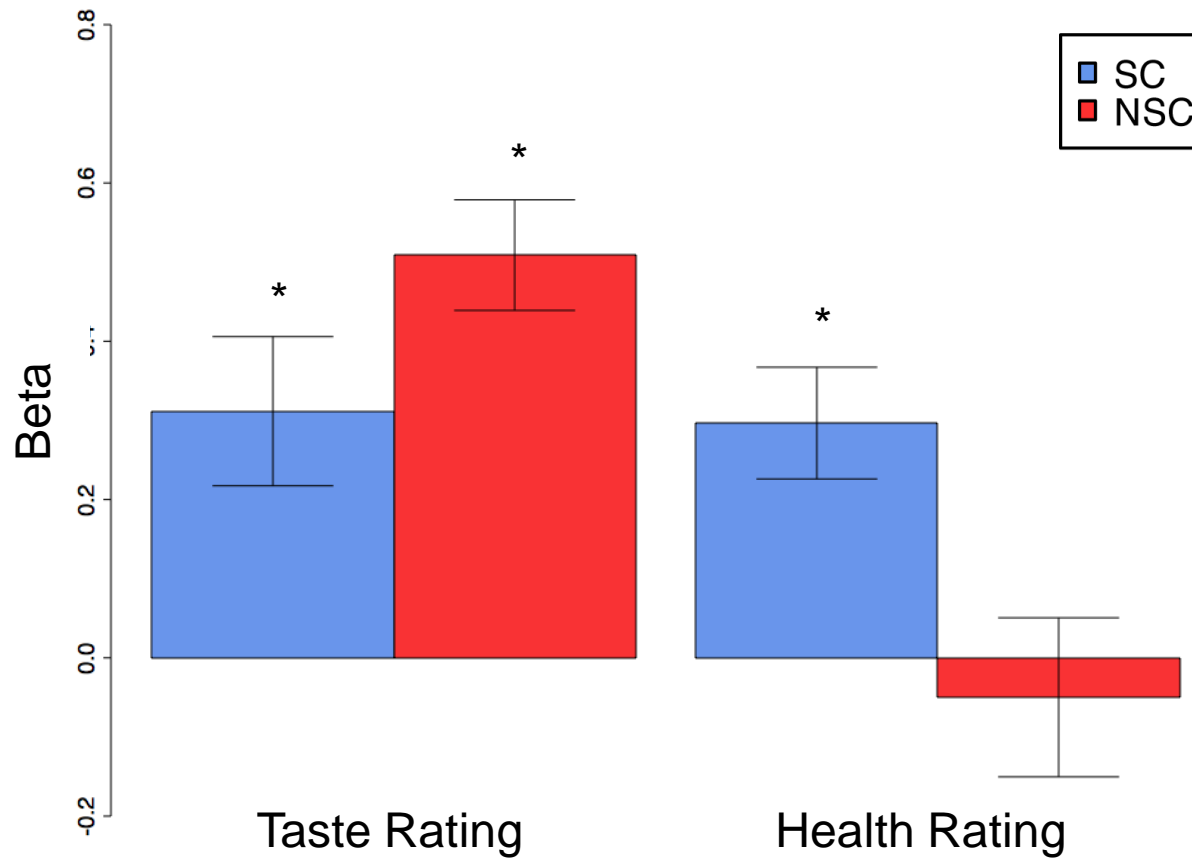
Activity in vmPFC is correlated with a behavioral measure of decision value (regardless of SC)



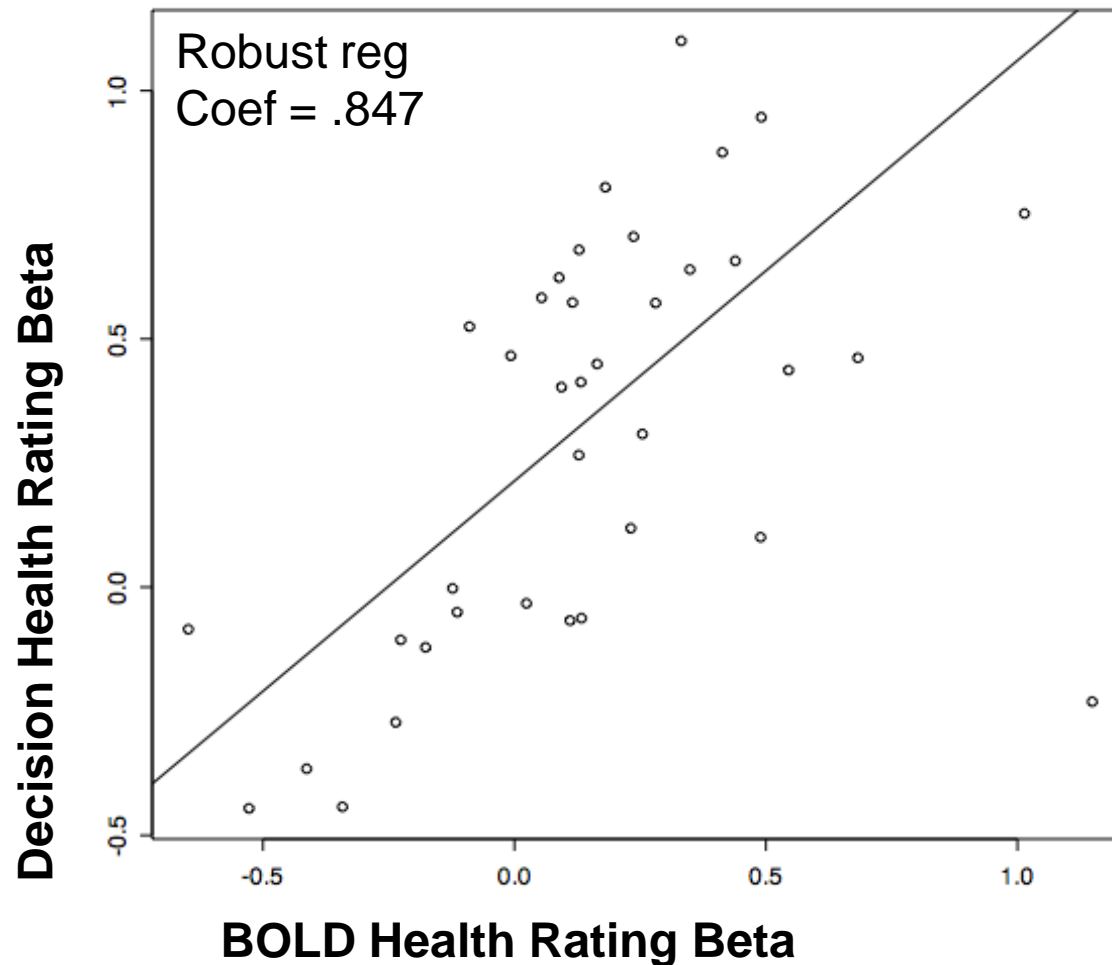
■ $p < .001$
■ $p < .005$



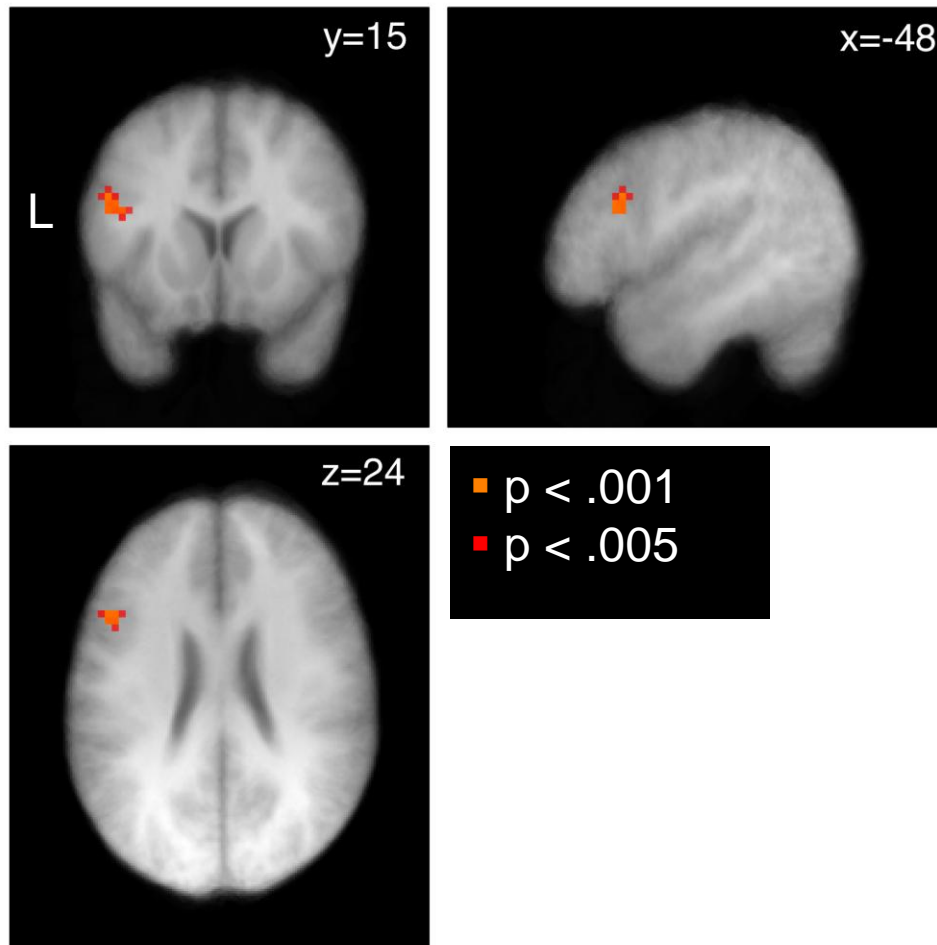
vmPFC BOLD signal reflects both taste and health ratings



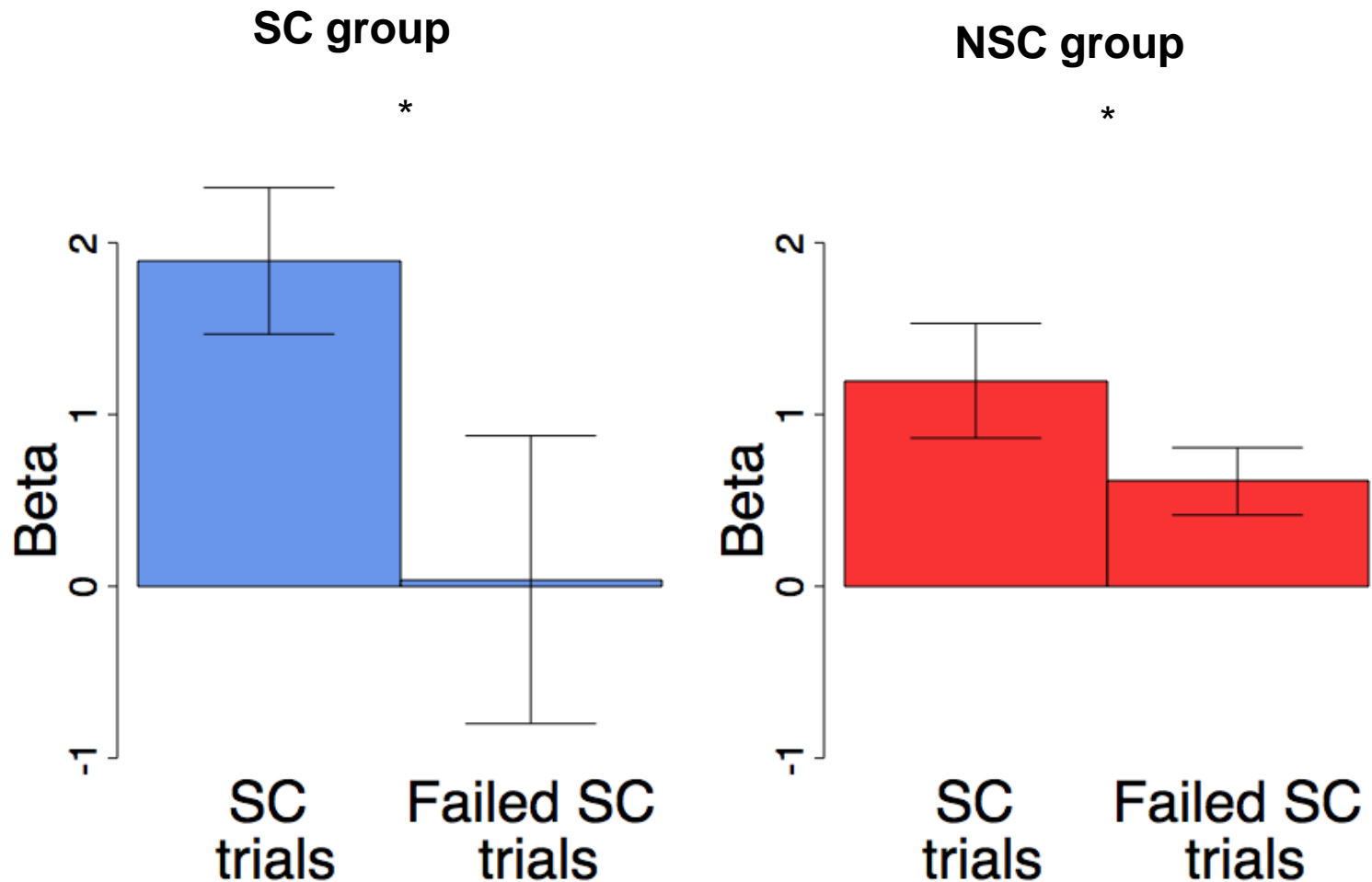
The effect of HR in the vmPFC is correlated with its effect on behavior



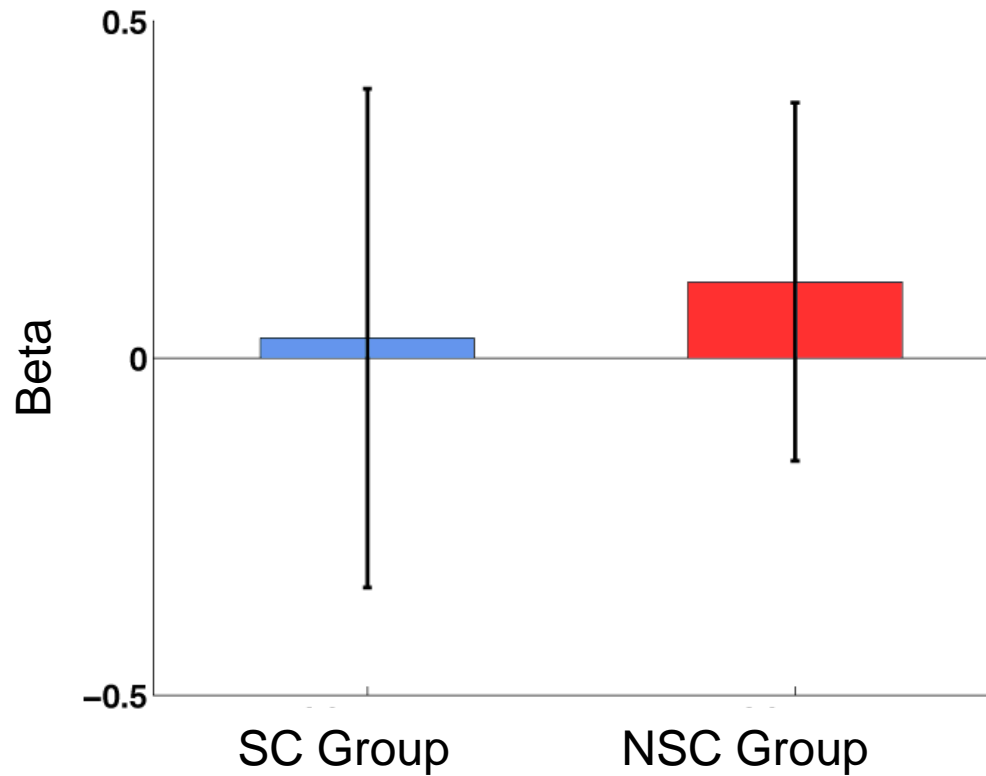
SC group has greater DLPFC than NSC when implementing self-control



More activity in DLPFC in successful SC trials than in failed SC trials



DLPFC activity does not correlate with HR

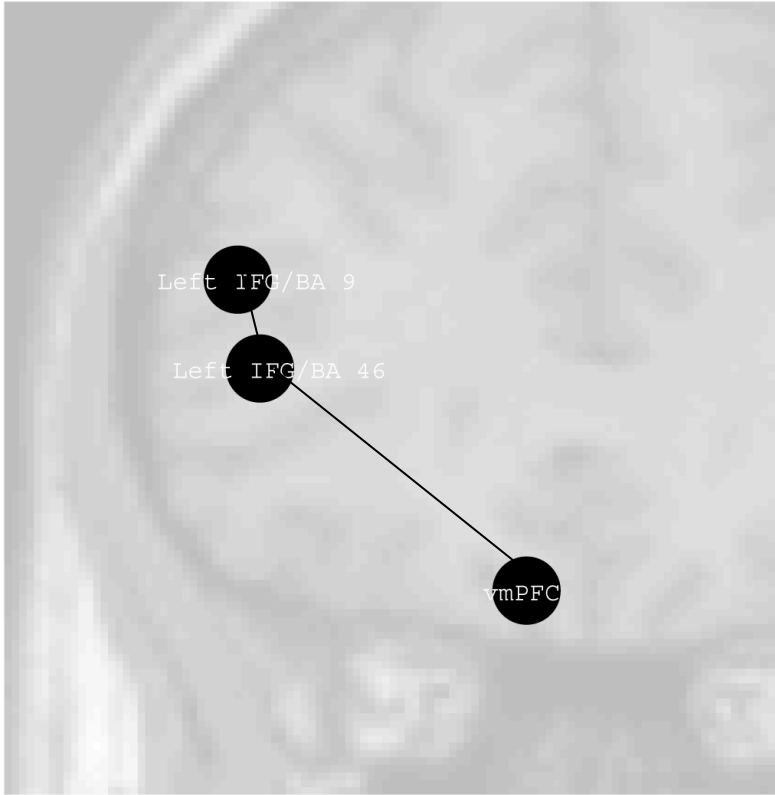
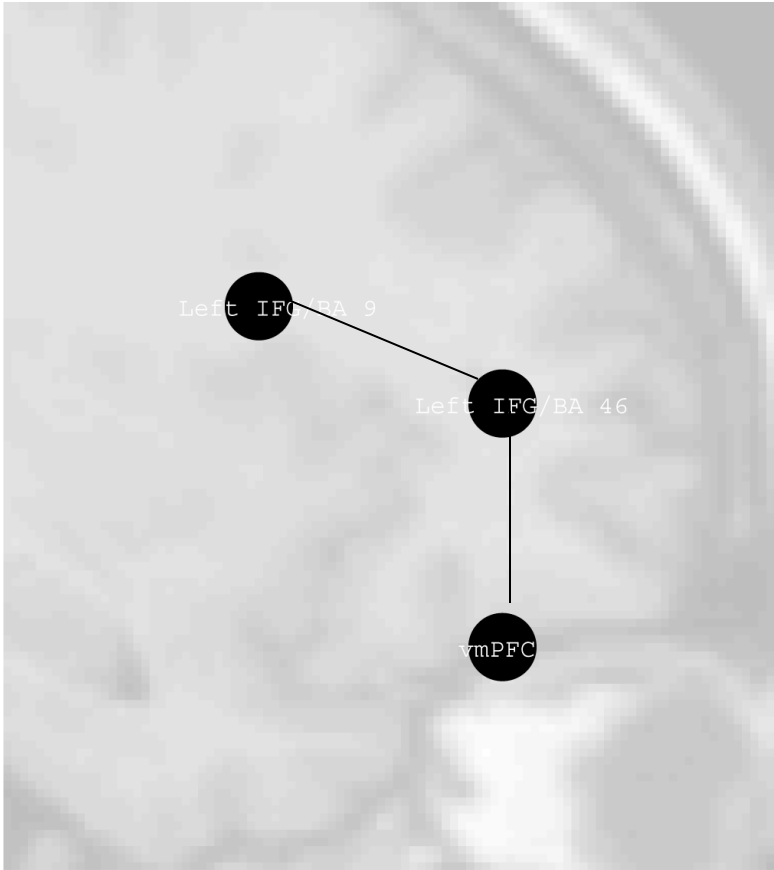


*Error bars = 95% confidence intervals

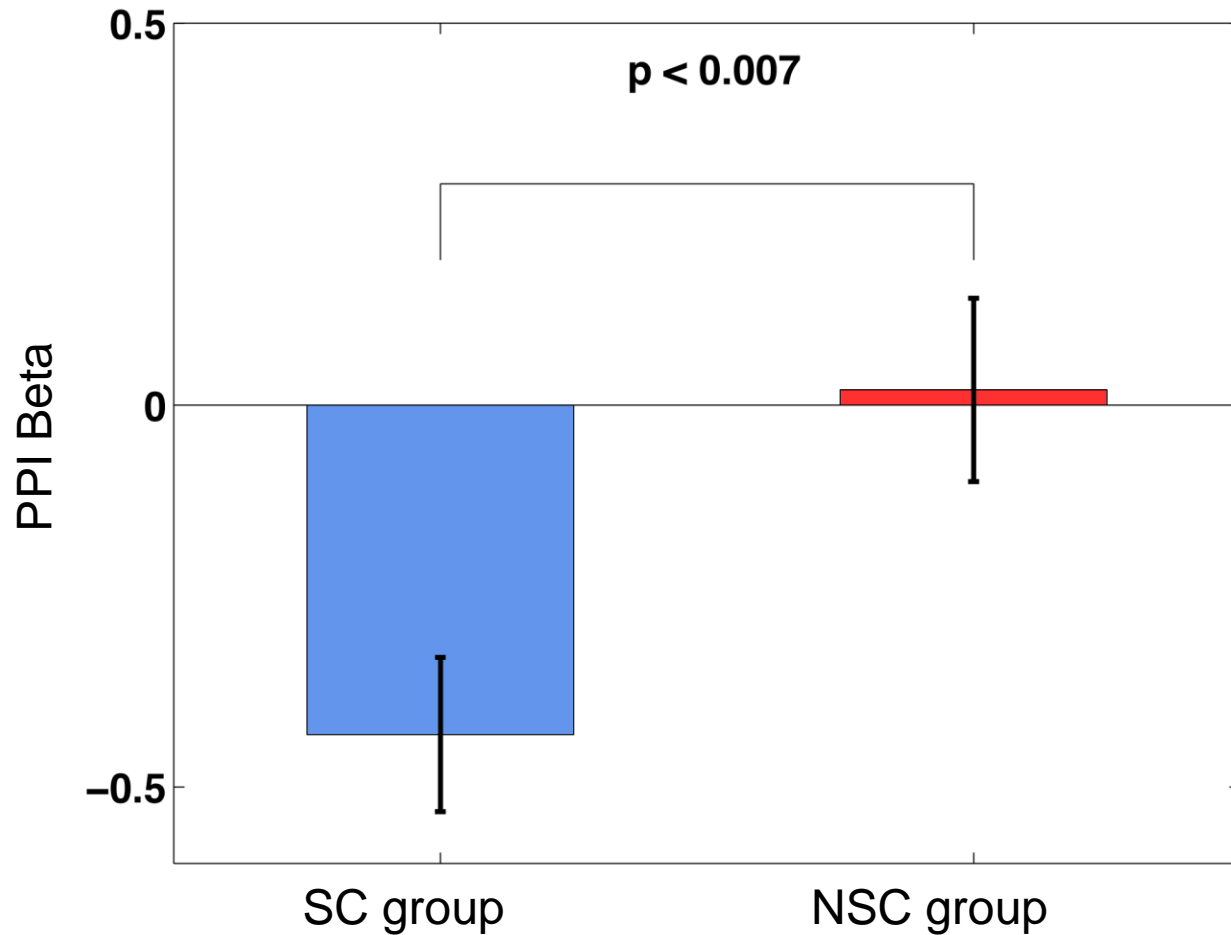
Attentional self-control network

Sagittal

Coronal



Group difference in PPI



Remarks

- Evidence attentional self-control involves modulation of vmPFC value signals by dlPFC so that they incorporate all dimensions of stimuli
- Healthy eaters in sample can do this
Unhealthy eaters cannot do this
- Have replicated results in a monetary discounting task

V
Final Remarks

Key points

- mOFC/vmPFC plays critical role in valuation during decision-making, probably by computing relative values
- A modified DDM provides very high accuracy description of psychometric data
- Both the valuation and comparison process are modulated by visual attention
- Evidence that dmPFC might be part of the comparator process that transforms values into motor responses

Next steps: Examples of critical open questions

Valuation:

- > How EXACTLY are the value signals in mOFC computed at time of choice?
- > What is the network of inputs that help at work in different decision problems and situations?
- > What EXACTLY is the code used in OFC to represent value of a stimulus?
- > How are the various components of the valuation learnt?
- > How does the brain know when to start and stop valuing a stimuli & which stimuli to evaluate?

-

Next steps: Examples of critical open questions

Comparison:

> More detailed models of comparator process and neurobiological basis

Ex:

- how are multiple value neurons integrated in comparison
- how is the DDM mapped to underlying neurobiology

-

Next steps: Examples of critical open questions

Motor:

- > How are stimuli and action representations mapped to each other?
- > Role of Supplementary Motor Areas
- > Role of basal ganglia- thalamic- cortical loops
- > Computational role for IPS

-

**Interested in post-doc
or PhD studies in neuroeconomics?
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