



# Temporal Integration of Olfactory Perceptual Evidence in Human Orbitofrontal Cortex

Nicholas E. Bowman,<sup>1,\*</sup> Konrad P. Kording,<sup>2</sup> and Jay A. Gottfried<sup>1,\*</sup>

<sup>1</sup>Department of Neurology, Northwestern University Feinberg School of Medicine, Chicago, IL 60611, USA

<sup>2</sup>Rehabilitation Institute of Chicago, Northwestern University, Chicago, IL 60611, USA

\*Correspondence: [nickbowman80@gmail.com](mailto:nickbowman80@gmail.com) (N.E.B.), [j-gottfried@northwestern.edu](mailto:j-gottfried@northwestern.edu) (J.A.G.)

<http://dx.doi.org/10.1016/j.neuron.2012.06.035>

## Smells Like Teen Spirit

*Humans' bad odour discrimination put at use*

Loïc Matthey

Tea Talk n°10

13th September 2012

# Introduction

---

- ❖ Temporal evidence integration
  - ❖ Good strategy under noisy perception
  - ❖ Well-studied in animals, e.g. monkeys, visual system
    - ❖ Random-dot task
    - ❖ Drift Diffusion Model
  - ❖ What about other senses, what about humans?
- ❖ This paper:
  - ❖ Tests olfactory integration
  - ❖ Finds fMRI correlates of DDM-like ramping activity in humans

**N. E. Bowman, K. P. Kording, J. A. Gottfried**

**“Temporal Integration of Olfactory Perceptual Evidence in Human Orbitofrontal Cortex”  
Neuron 75, September 2012**

# Paper construction

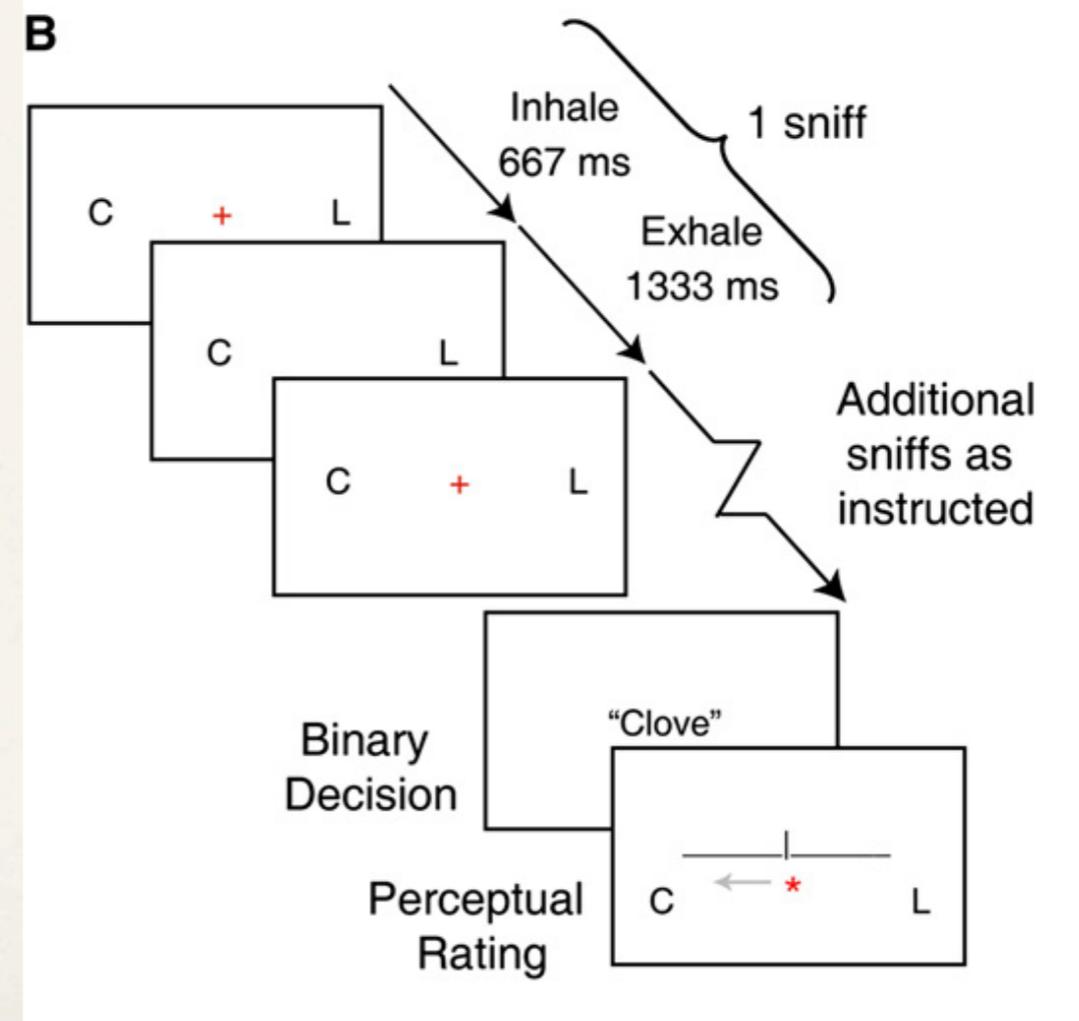
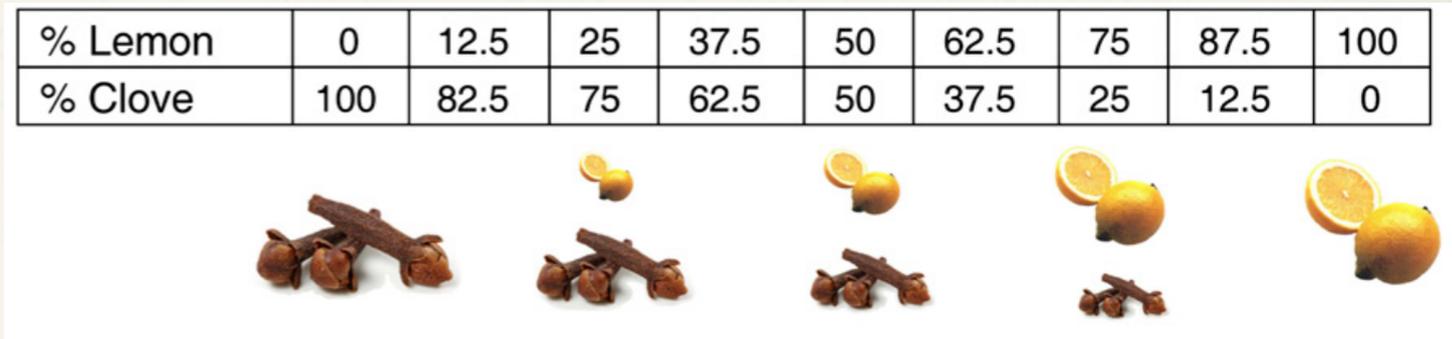
---

- ❖ Verify that odour identification is temporally integrated
- ❖ A DDM model is a good fit to observed RTs
- ❖ Look for brain regions showing integration-like responses, with fMRI
  - ❖ Human olfactory perception slow, especially for mixtures: able to see the signal using fMRI!

# Odour discrimination

- Identify dominant odour in mixture

- Two alternative forced-choice discrimination
- eugenol (“clove”) vs citral (“lemon”)
  - matched for perceived intensity
  - 8 channels air-dilution olfactometer
  - control proportion each channel contributing to airflow
- 10 participants
- If integration, more sniffs => better performance



# Odour discrimination

- ❖ Two blocks of trials

- ❖ “Fixed-sniff”

- ❖ cued, 1-3 sniffs

- ❖ “Open-sniff”

- ❖ As many as needed to make confident decision

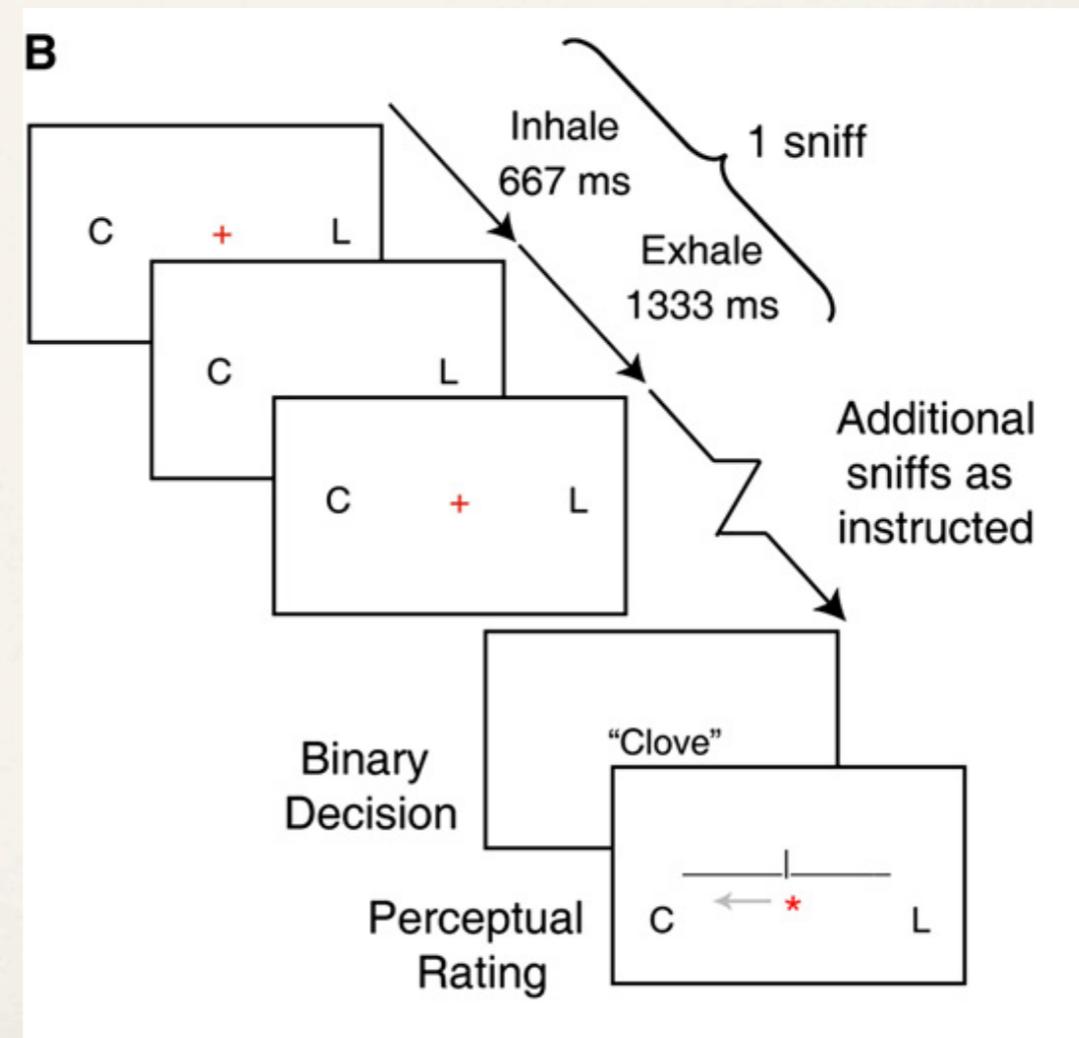
- ❖ Binary decision

- ❖ (also ask for perceptual rating but not used in results)

- ❖ 18 s between trials, 144 trials total

- ❖ No feedback

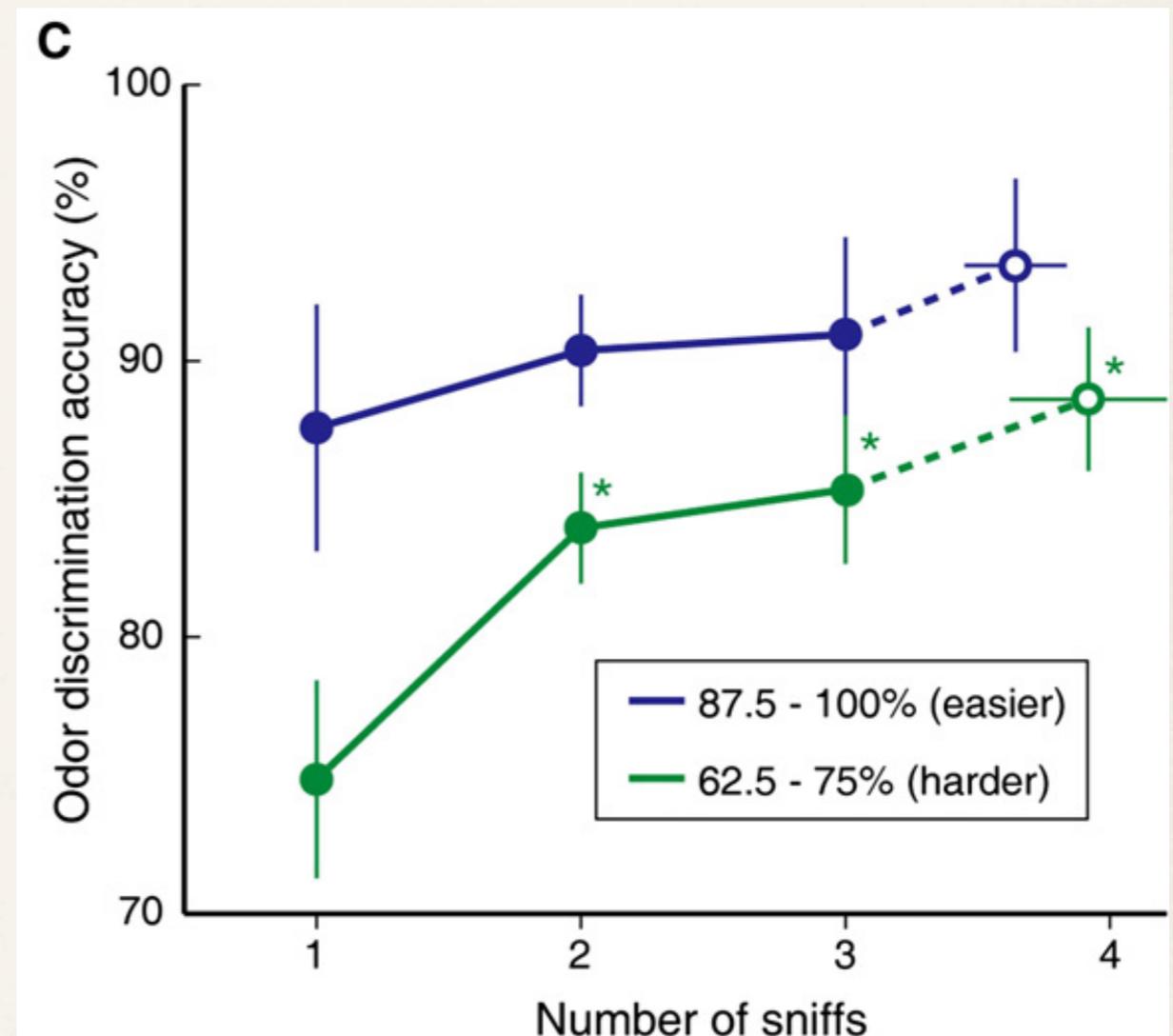
% Lemon	0	12.5	25	37.5	50	62.5	75	87.5	100
% Clove	100	82.5	75	62.5	50	37.5	25	12.5	0



# Odour discrimination

## ❖ Results

- ❖ Accuracy improvement with number of sniffs
- ❖ Depends on mixture difficulty

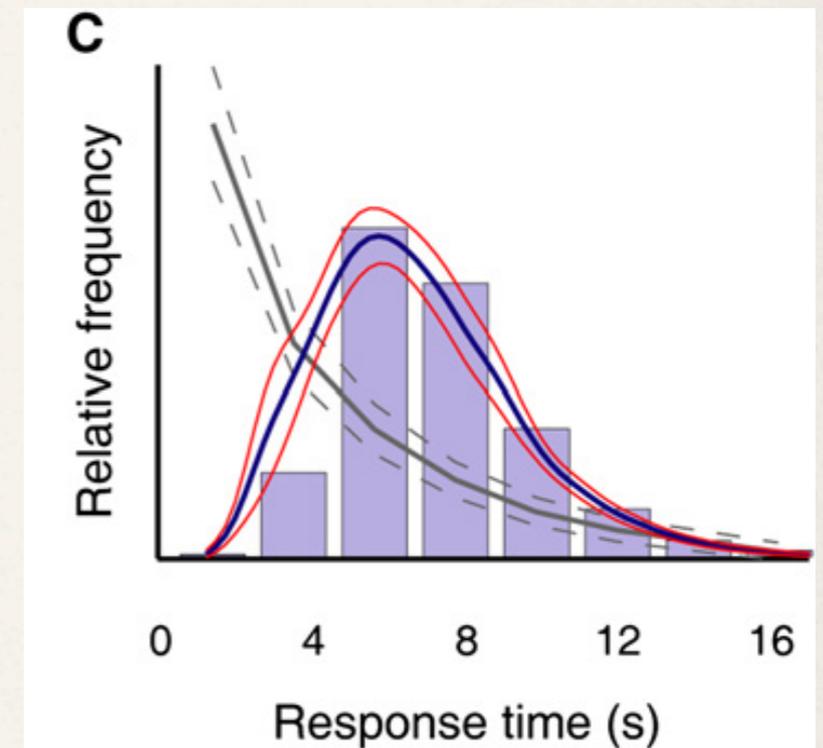
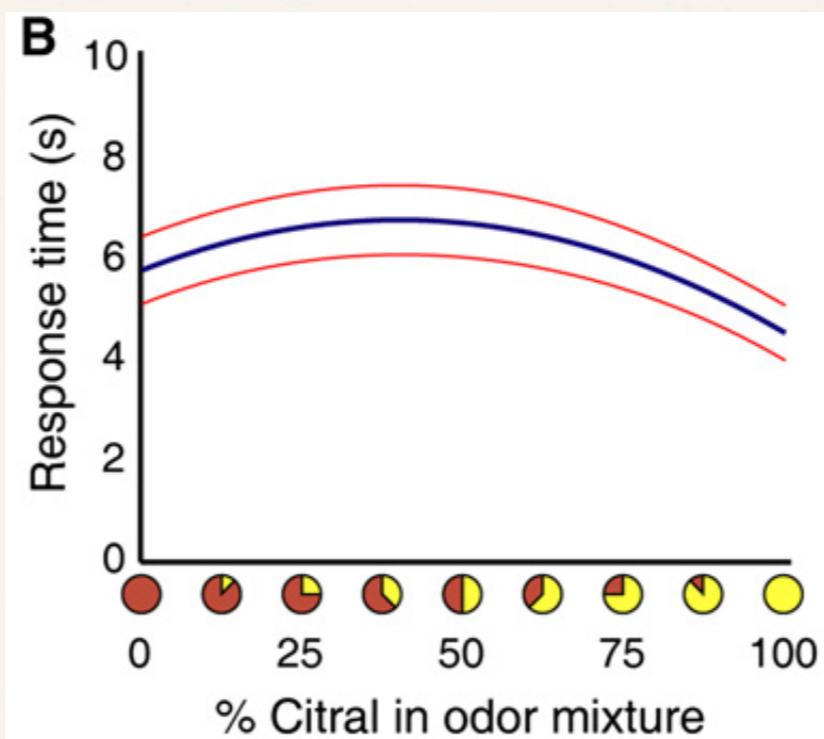
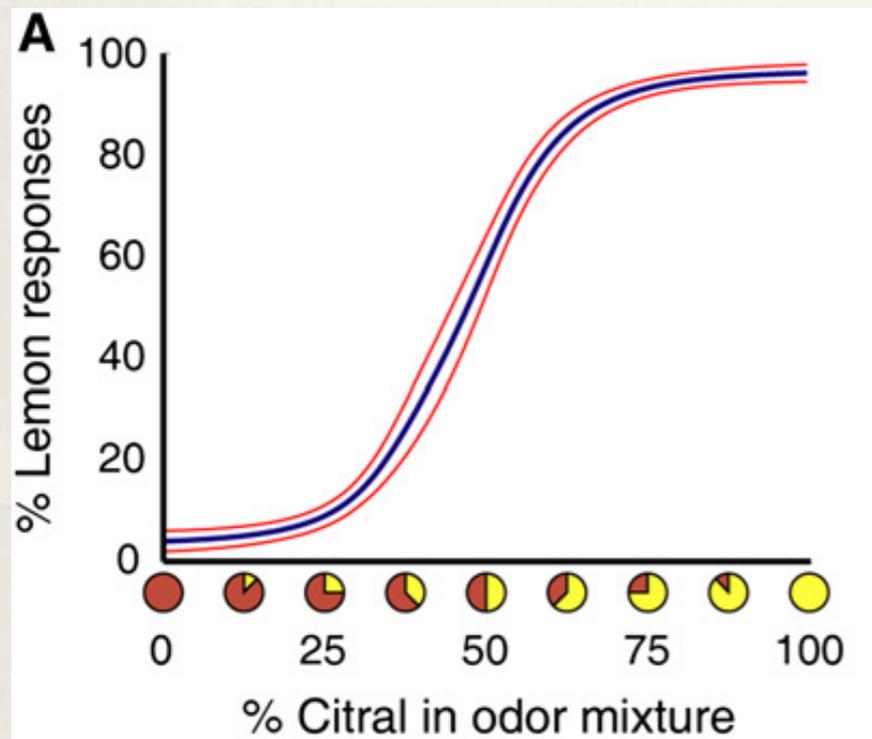


\* $p < 0.05$

# Odour discrimination

## ❖ Results

- ❖ Psychophysical data
- ❖ Consistent with Drift-Diffusion Model



# DDM

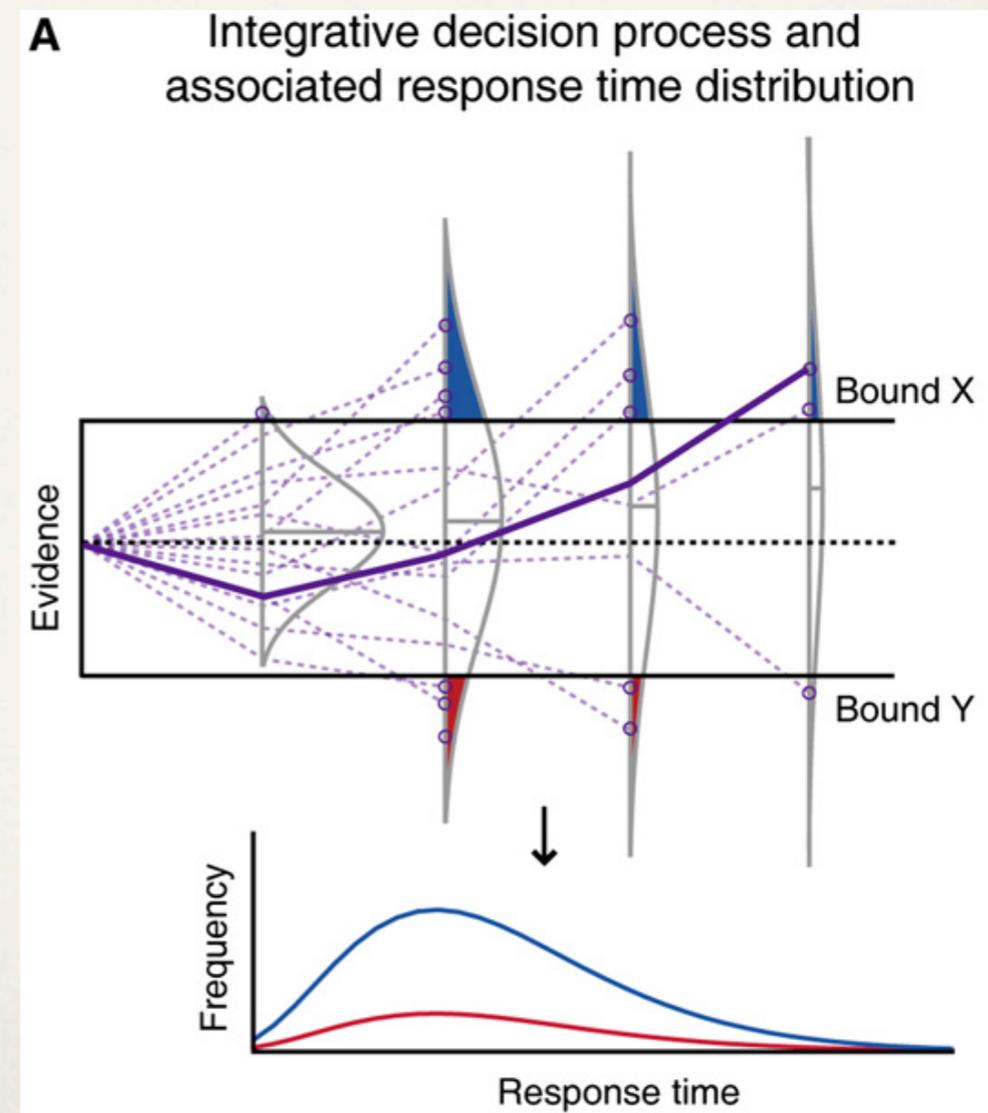
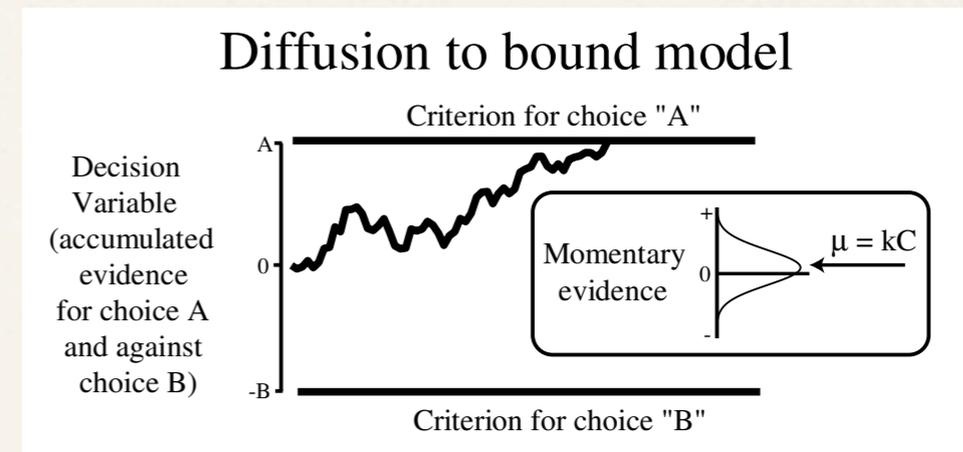
## ❖ Drift-diffusion model

- ❖ Simple 1D model of evidence integration

$$dx = A dt + c dW, \quad x(0) = 0$$

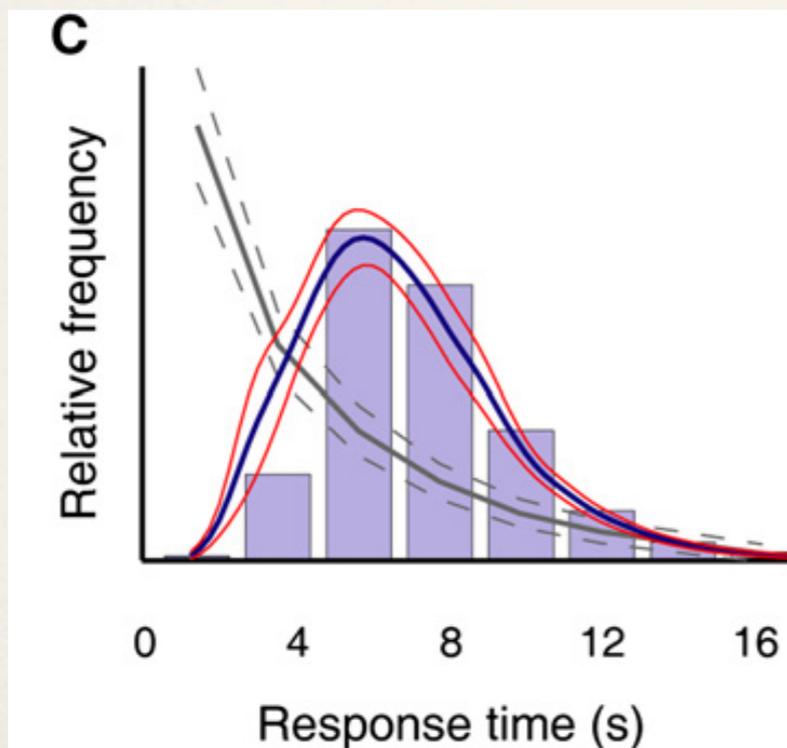
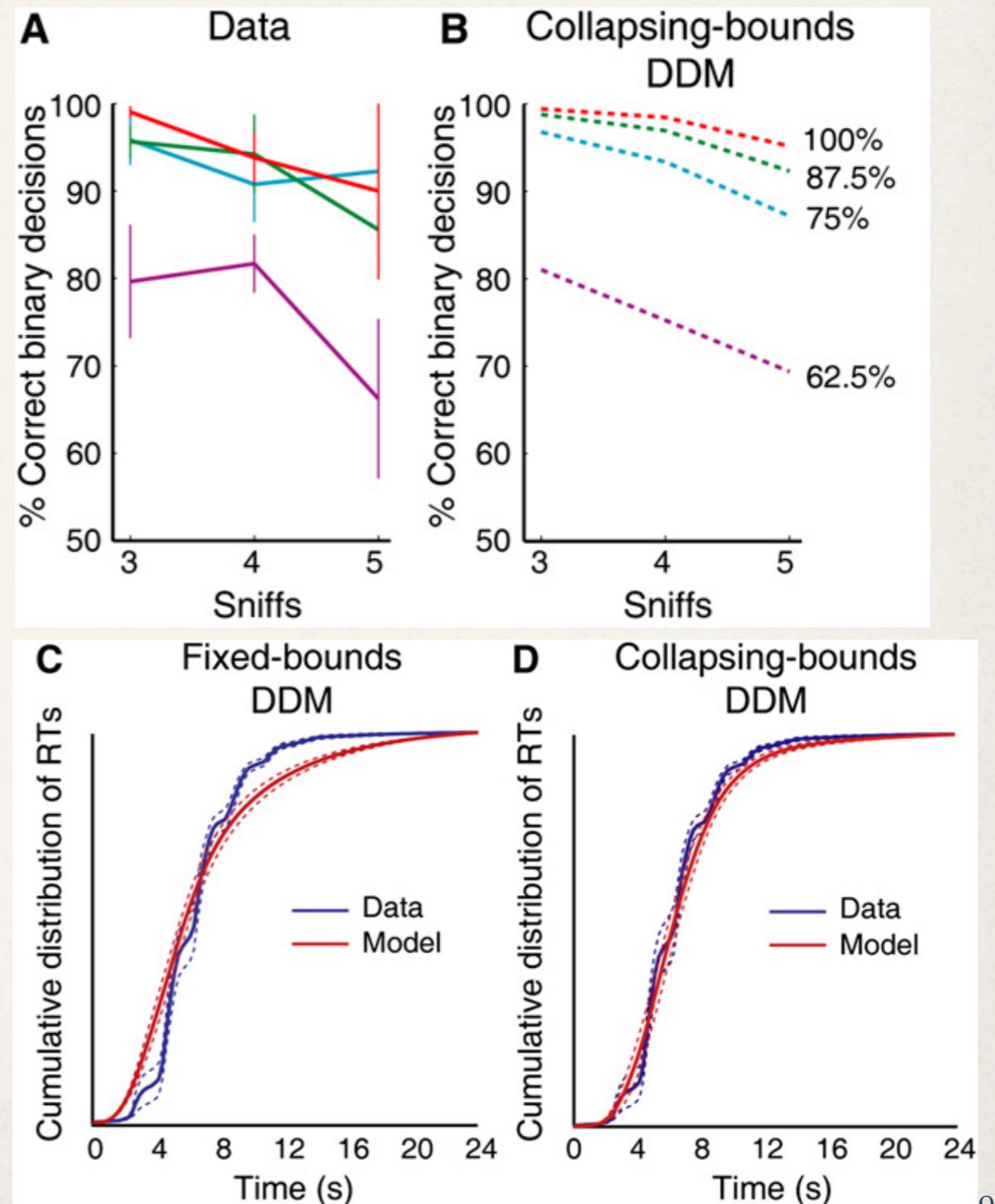
- ❖  $x$ : accumulated evidence.
  - ❖ Positive: towards choice A
  - ❖ Negative: towards choice B
- ❖  $A$ : drift term, “momentary evidence” biased towards A or B for a given trial.
- ❖ Noise:  $dW \sim N(0, c^2 dt)$
- ❖ Easy to solve for distribution, error rates and response times

$$p(x, t) = N(At, c\sqrt{t})$$



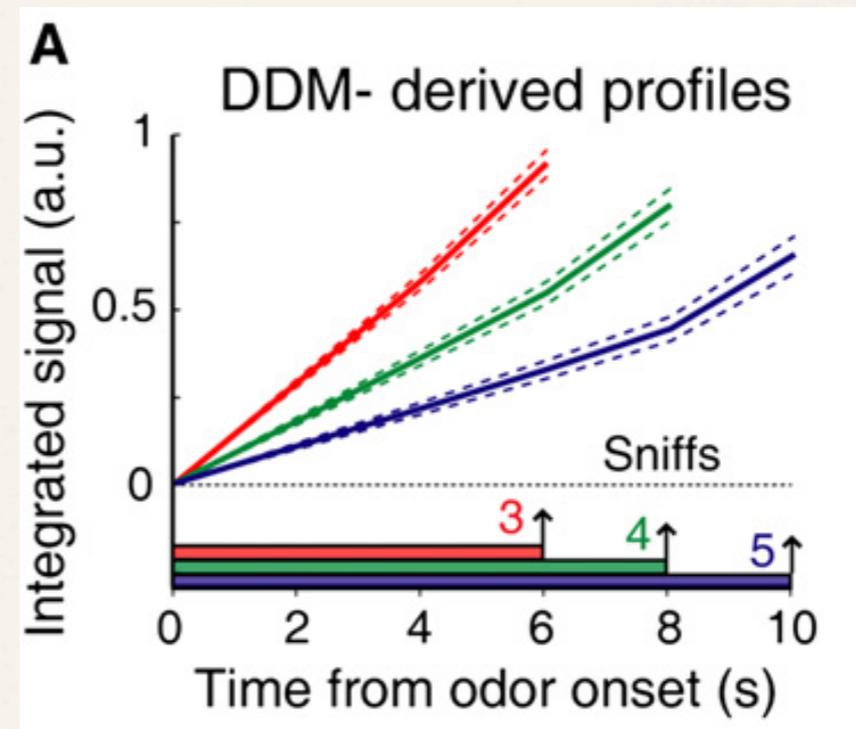
# Match with DDM

- ❖ RTs correspond to DDM
- ❖ Collapsing-bounds DDM actually better fit.



# fMRI correlates

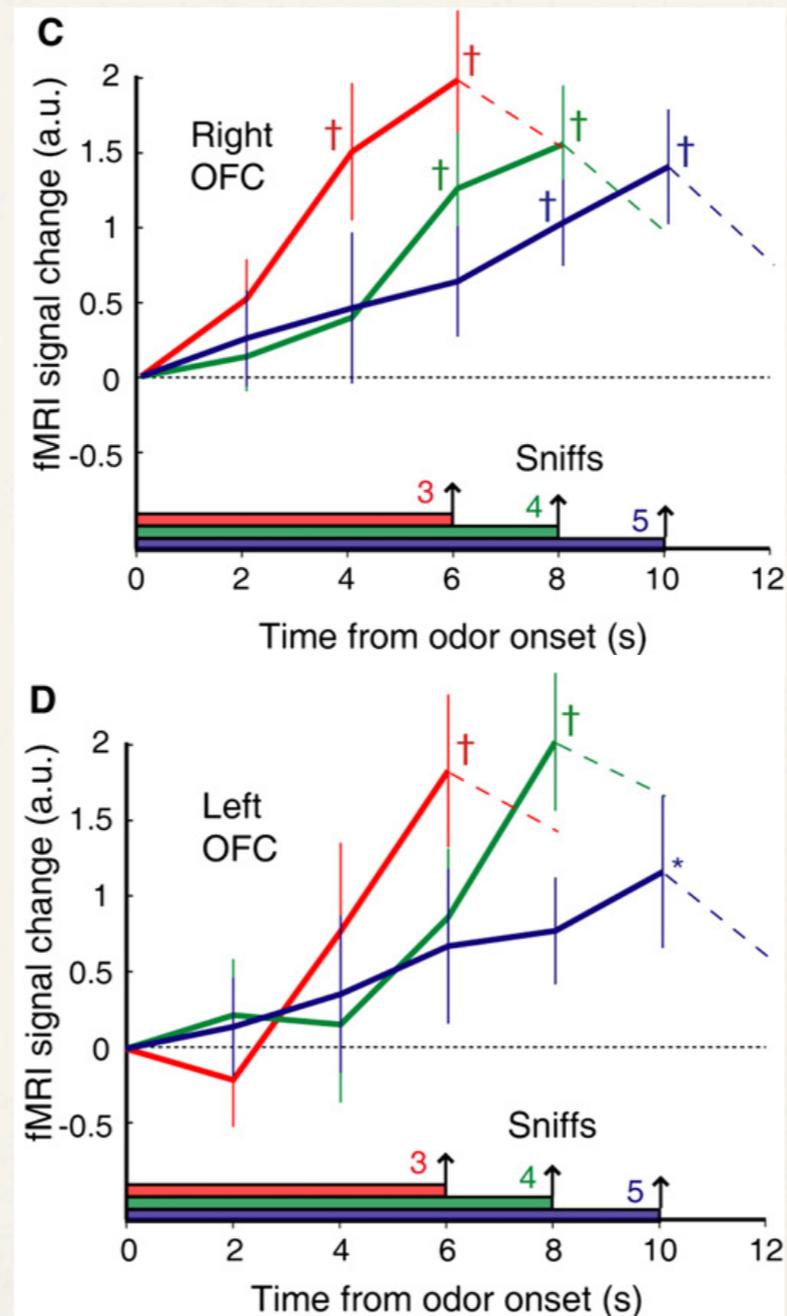
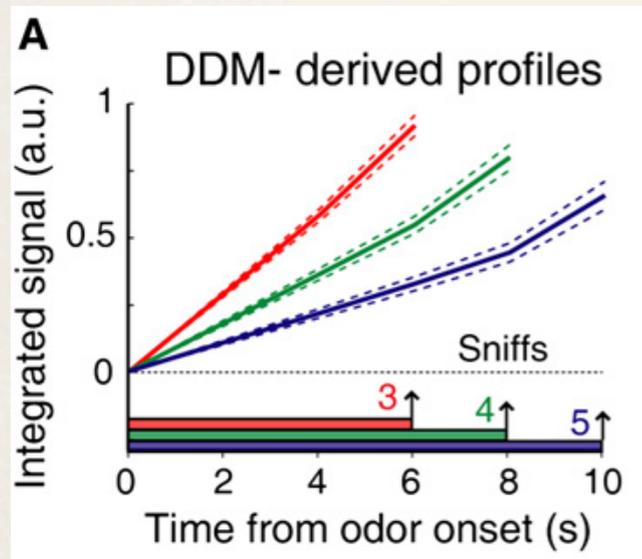
- ❖ Open-sniff experiment
- ❖ 2s repetition time, 128x120 voxels
  - ❖ 1 sniff = 2s as well
- ❖ Look for voxels correlating with DDM-derived integrated signal responses
  - ❖ Per subject fit, DDM profiles, voxel selection.
  - ❖ 14 time bins of 2s duration



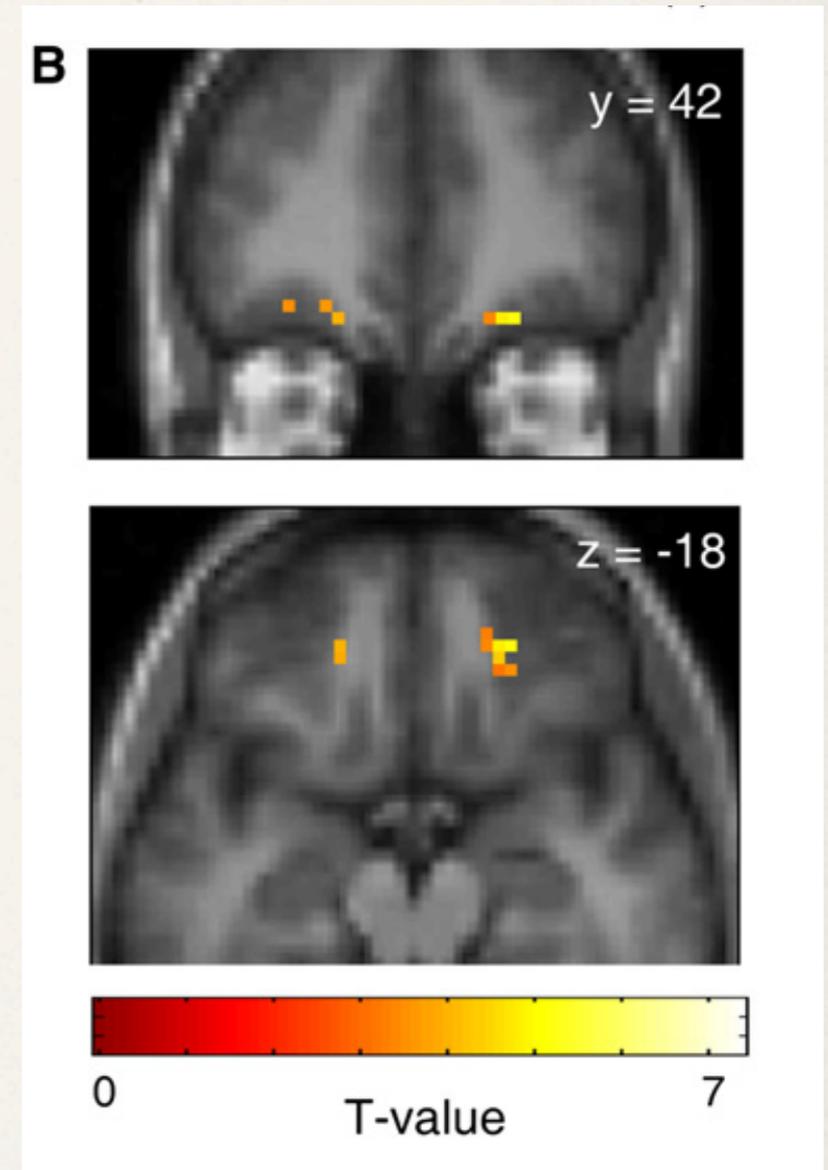
# fMRI correlates

## ❖ Results

- ❖ Found region showing integration-like profiles:
  - ❖ Medial Orbitofrontal Cortex (OFC)



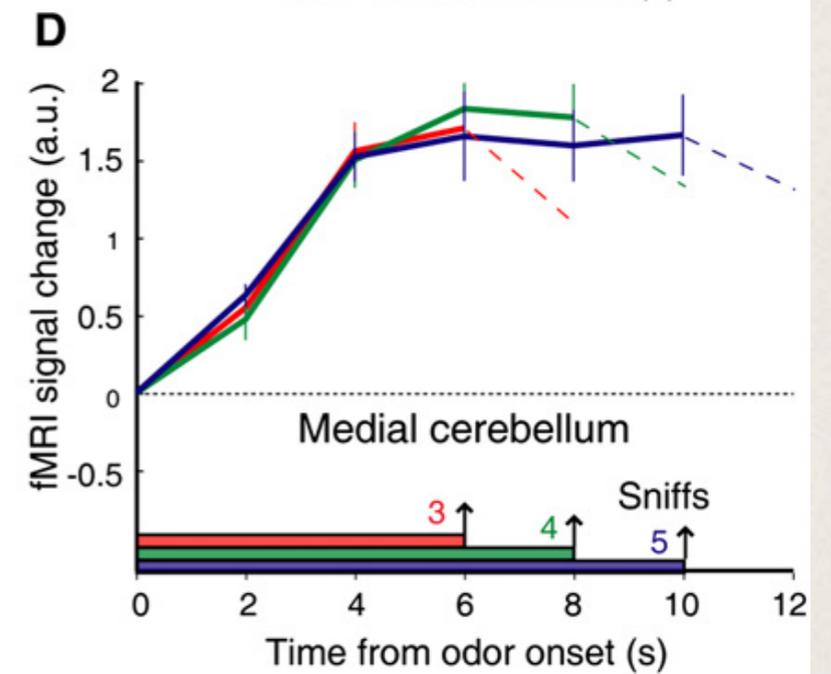
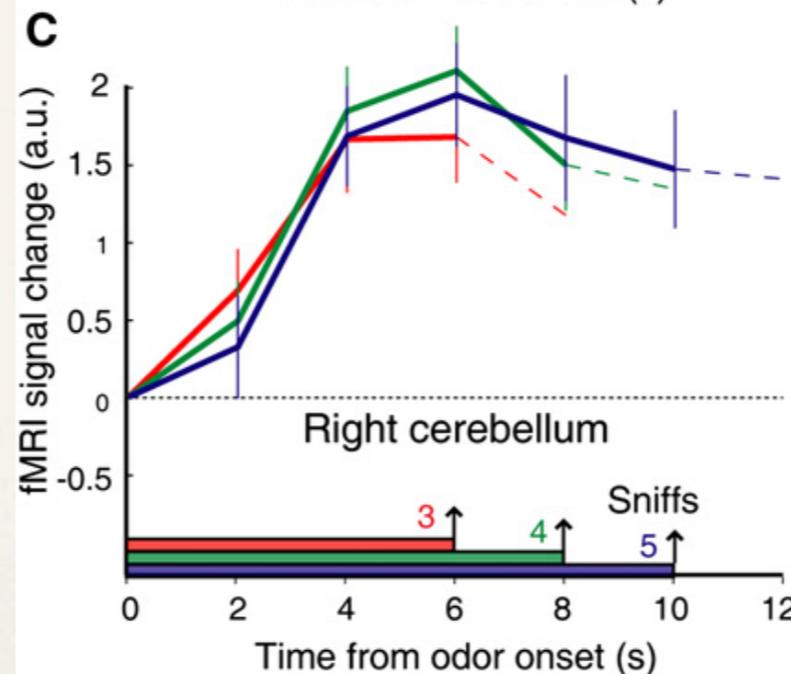
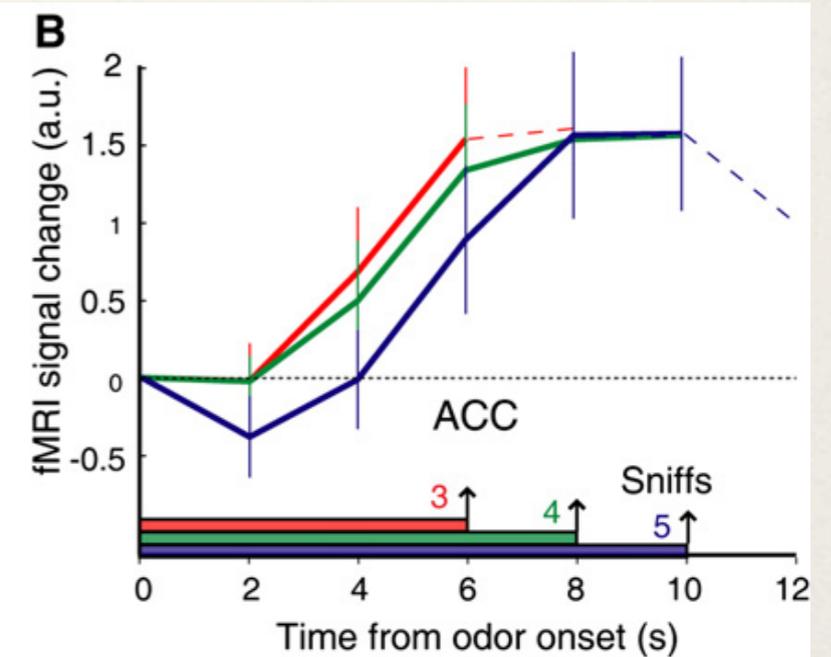
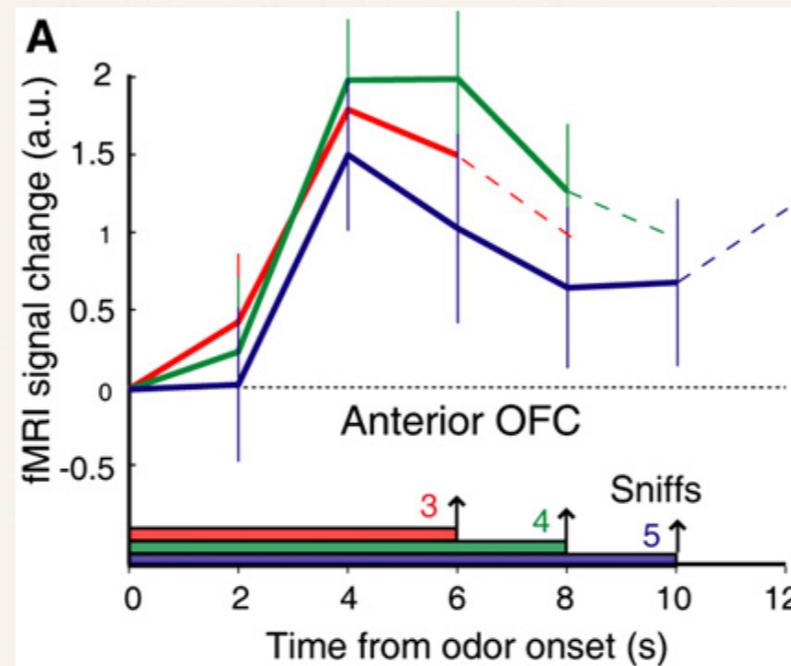
\* $p < 0.05$ ; †,  $p < 0.01$



# fMRI correlates

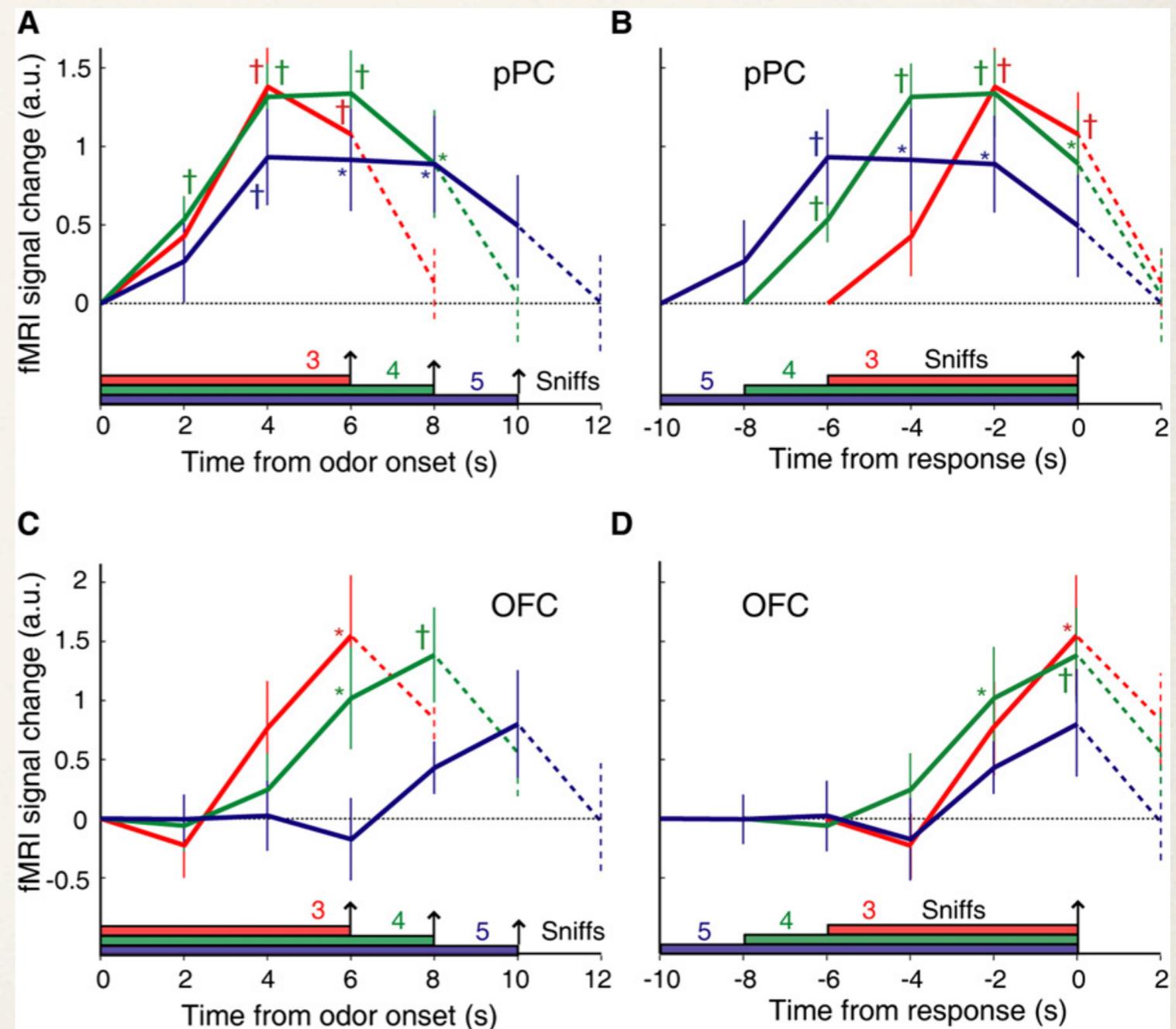
- ❖ Other activations

- ❖ Anterior Cingulate Cortex (ACC), Cerebellum
- ❖ Do not show significant interaction of condition and time



# fMRI correlates

- ❖ Hypothesis: pPC generates momentary olfactory evidence, to be integrated by OFC
- ❖ fMRI signal somehow consistent.



# Conclusion

---

- ❖ Found temporal integration of olfactory evidence (though weak)
- ❖ Make use of slow poor human performance to their advantage
- ❖ Found OFC correlates with DDM-like integration profiles
  - ❖ Identified region corresponds to putative olfactory projection site in human OFC
  - ❖ Rodent single-unit recording study on OFC:  
OFC report decision confidence during postchoice period
- ❖ pPC - OFC similar to MT - LIP in visual perceptual evidence integration in monkeys.

# The End

---

- ❖ Questions?

- ❖ References:

- ❖ M. B. Ahrens, J. M. Li, M. B. Orger, D. N. Robson, A.F. Schier, F. Engert, R. Portugues, “Brain-wide neuronal dynamics during motor adaptation in zebrafish”, Nature, published online May 2012

---

❖ Supplementary slides

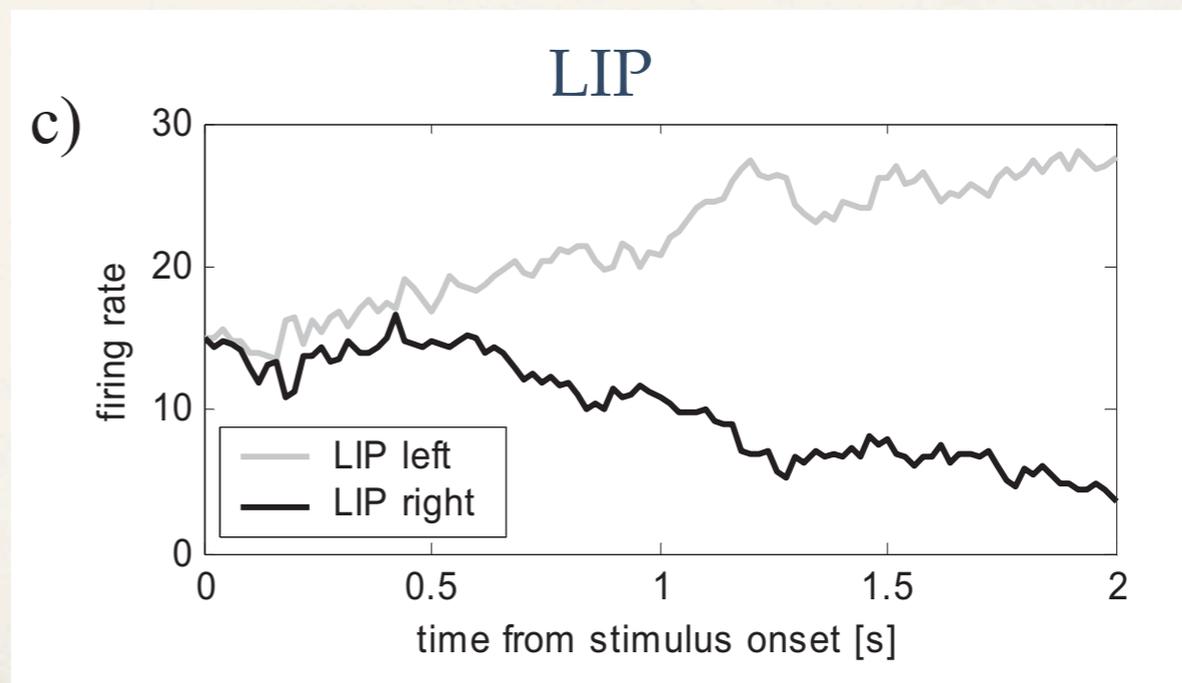
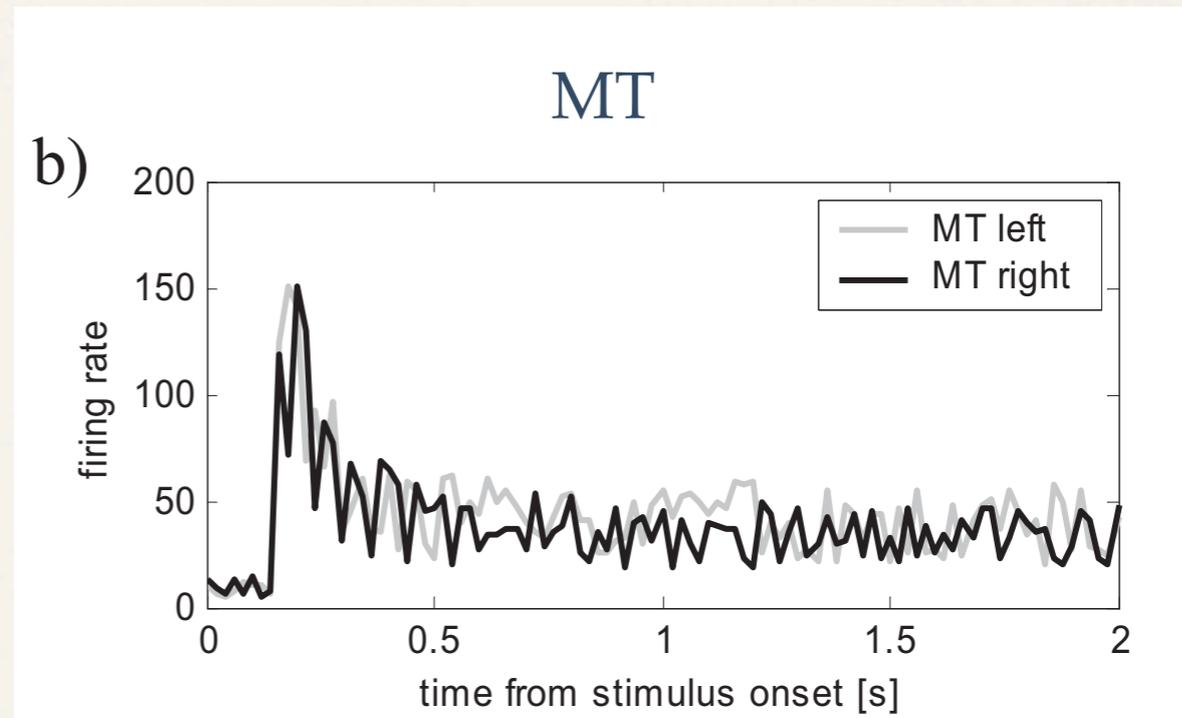
# Experimental data

- ❖ Behavioral data

- ❖ Performance as function of coherence (~difficulty of task)
- ❖ Distribution of response times.

- ❖ Neuronal recordings:

- ❖ Middle Temporal Area (MT)
  - ❖ ~ momentary evidence
- ❖ Lateral Intraparietal Area (LIP)
  - ❖ ~ accumulated evidence



# Sequential Probability Ratio Test

---

- ❖ Assume two populations reporting evidence for two alternatives (left / right):  $I_1$  and  $I_2$
- ❖ Let  $Y = I_1 - I_2$ .
  - ❖ If “right” hypothesis is true:  $Y \sim p_1(y)$ , with mean  $\mu_1 > 0$
  - ❖ If “left”:  $Y \sim p_2(y)$ , with mean  $\mu_2 < 0$
- ❖ Get iid samples from  $p_i(y)$ .
- ❖ Goal: Decide as soon as possible which hypothesis is true.

- ❖ Optimal solution: Likelihood-ratio test:  $Z_2 < \frac{p_1(y_1)p_1(y_2) \dots p_1(y_n)}{p_2(y_1)p_2(y_2) \dots p_2(y_n)} < Z_1$

- ❖ Taking  $\log$ , equivalent to random walk

$$\log Z_2 < \log \frac{p_1(y_1)}{p_2(y_1)} + \dots + \log \frac{p_1(y_n)}{p_2(y_n)} < \log Z_1$$
$$\Rightarrow I^n = I^{n-1} + \log \frac{p_1(y_n)}{p_2(y_n)}$$

# Drift Decision Model

---

- ❖ Simple:  
Analytical formulas for the Error rate and Response time.

Fixed time

$$ER = \Phi\left(-\frac{A}{c}\sqrt{T}\right)$$

Free-response

$$ER = \frac{1}{1 + e^{\frac{2Az}{c^2}}}$$

$$DT = \frac{z}{A} \tanh\left(\frac{Az}{c^2}\right)$$

z: bound

- ❖ Optimal model, as implements the Neyman-Pearson test.
- ❖ Extensions:
  - ❖ Drift variability:  
 $A \sim N(m_A, s_A)$
  - ❖ Initial position variability:  
 $x_0 \sim U[-s_x, s_x]$