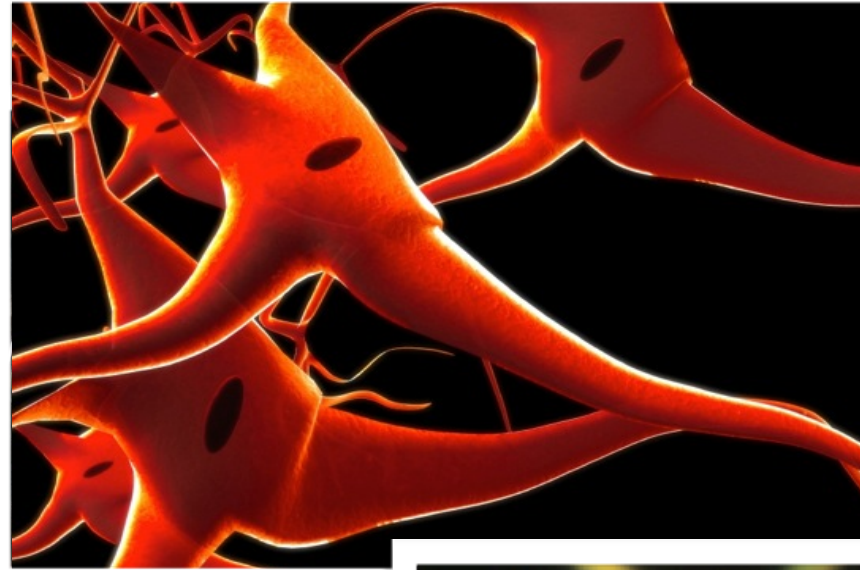
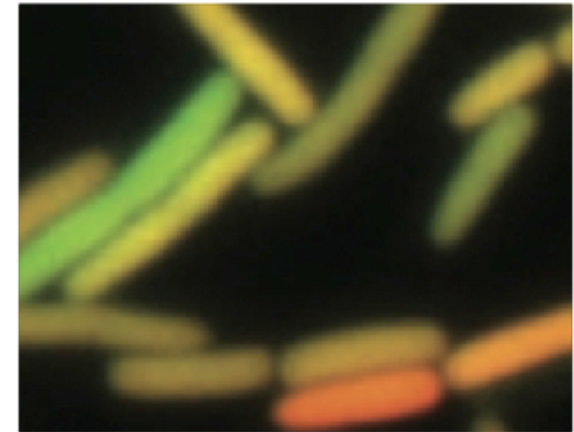


THOMAS SCHLEGEL



UNIVERSITY OF BRISTOL



PETER SWAIN

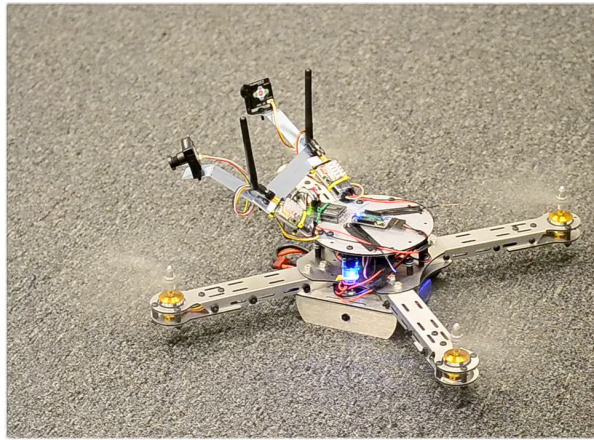
FROM HOUSE-HUNTING HONEYBEES TO NEURAL MODELS AND PSYCHOPHYSICS

JAMES A. R. MARSHALL

DEPARTMENT OF COMPUTER SCIENCE AND KROTO RESEARCH INSTITUTE,
UNIVERSITY OF SHEFFIELD

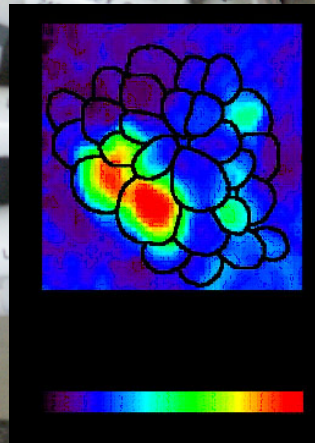
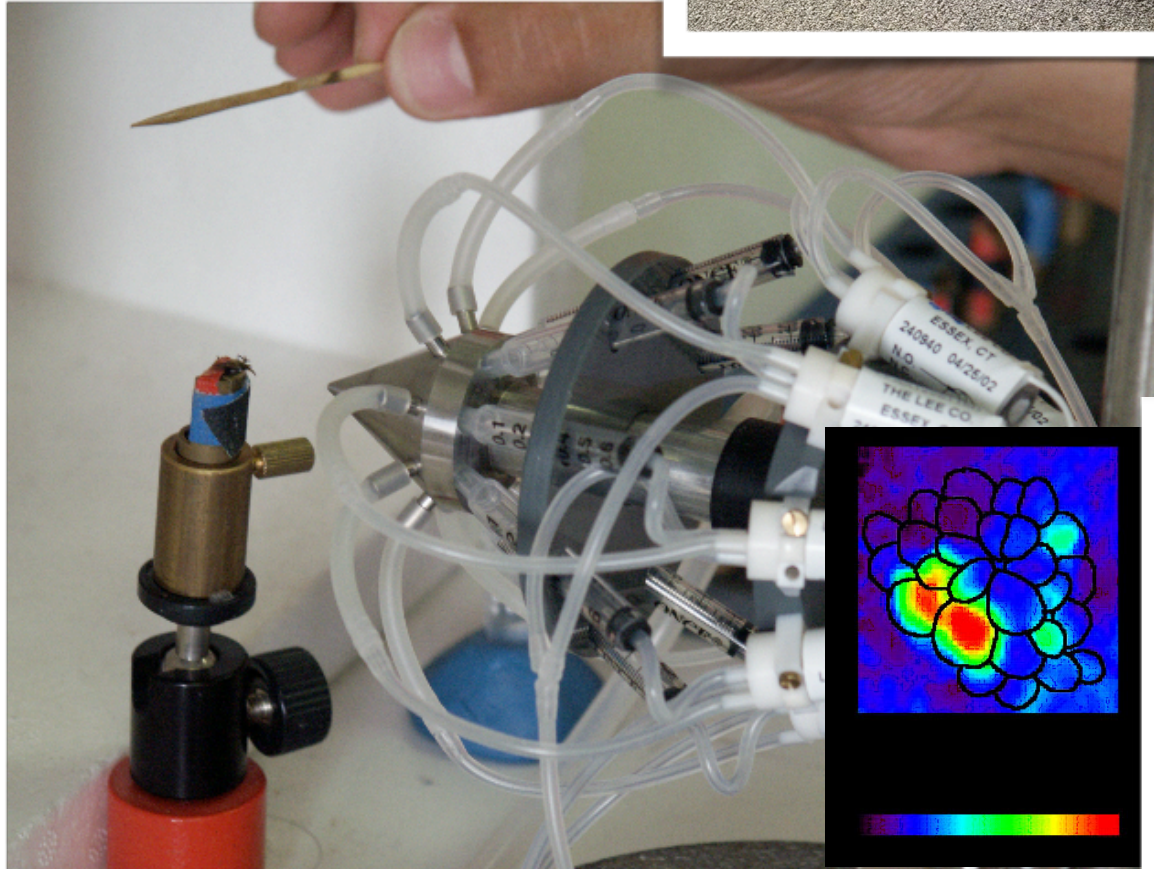


GREEN BRAIN



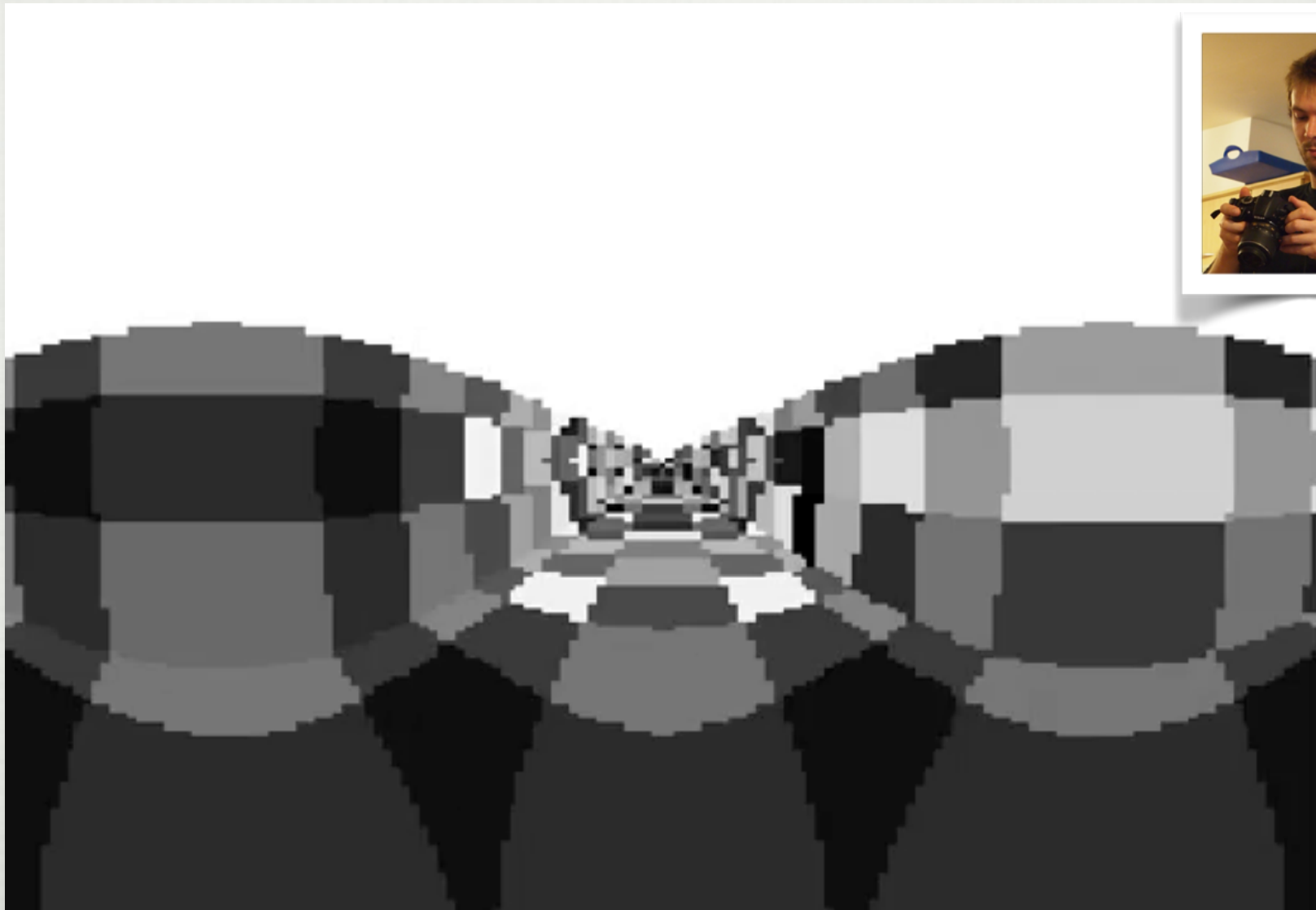
FU BERLIN

GREENBRAINPROJECT.CO.UK



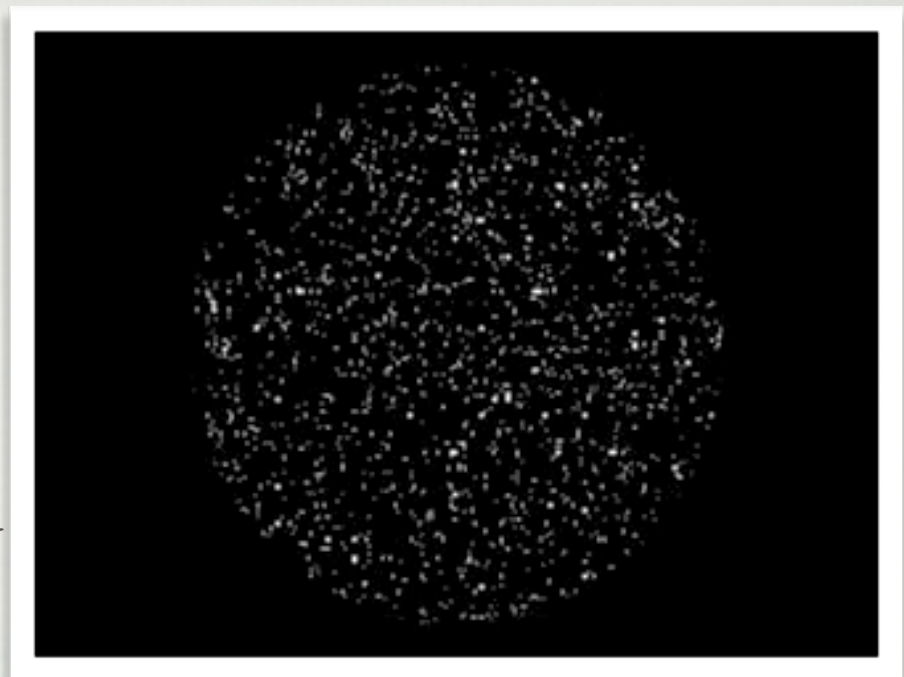
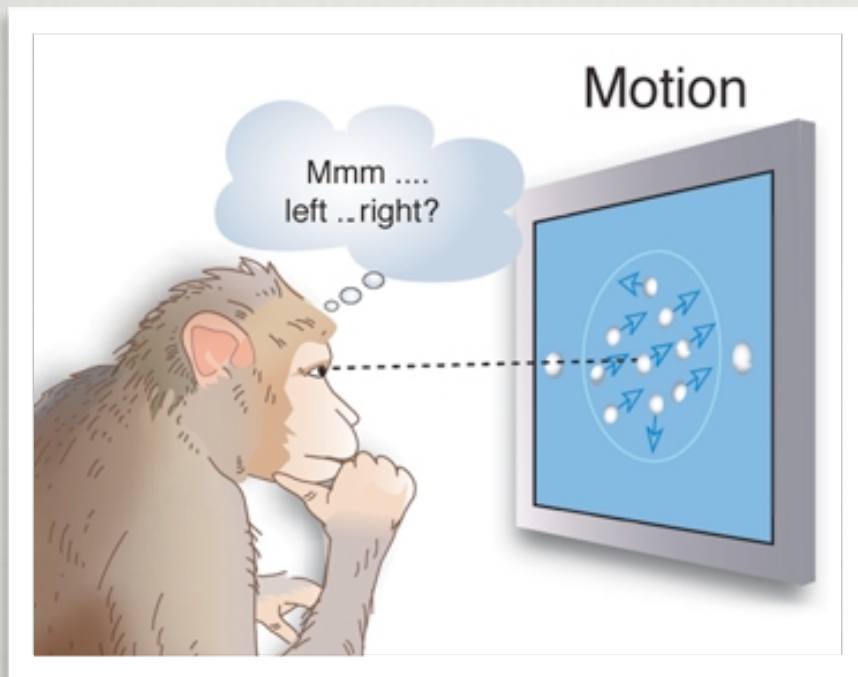
NVIDIA

SIMULATING BEE VISION AND FLIGHT CONTROL

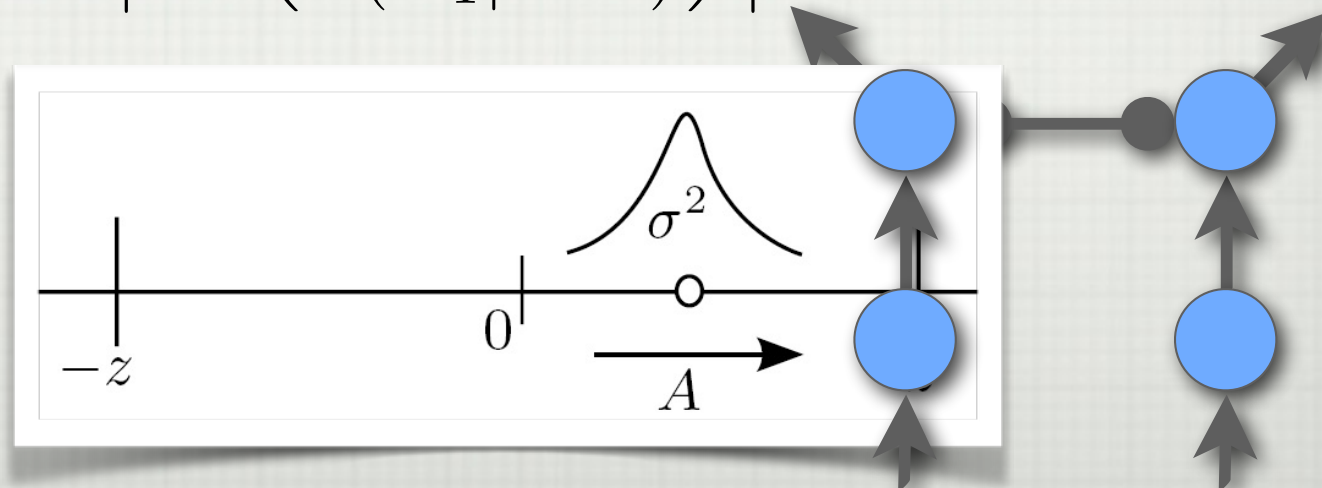




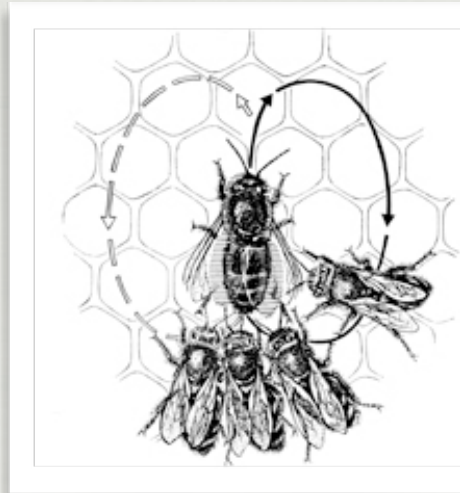
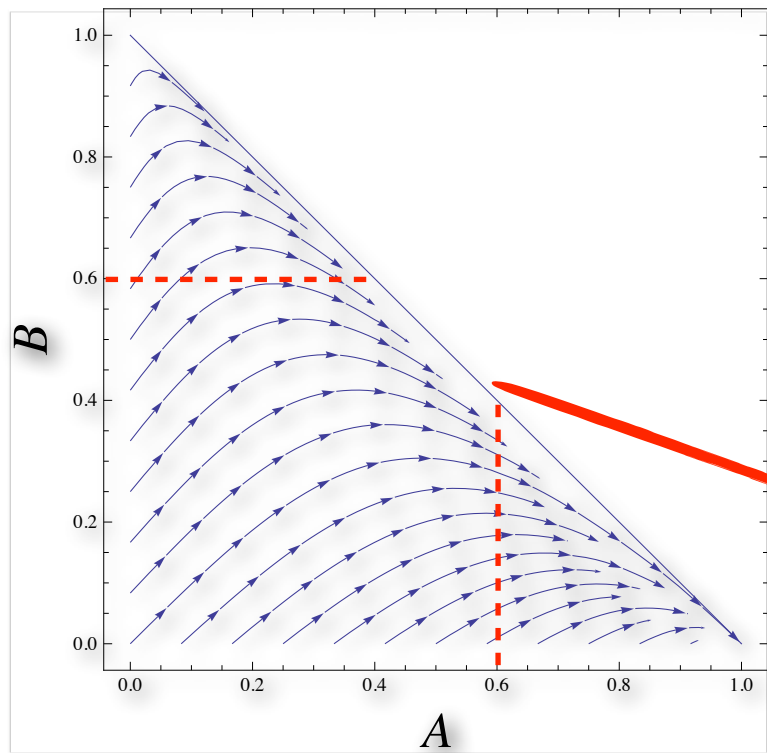
TOM SEELEY



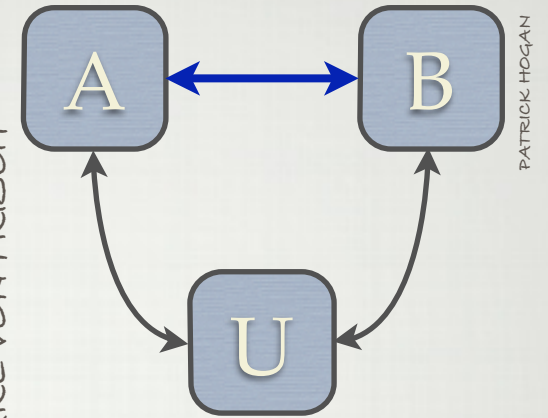
$$\left| \log \left(\frac{P(H_0|\text{data})}{P(H_1|\text{data})} \right) \right| > z$$



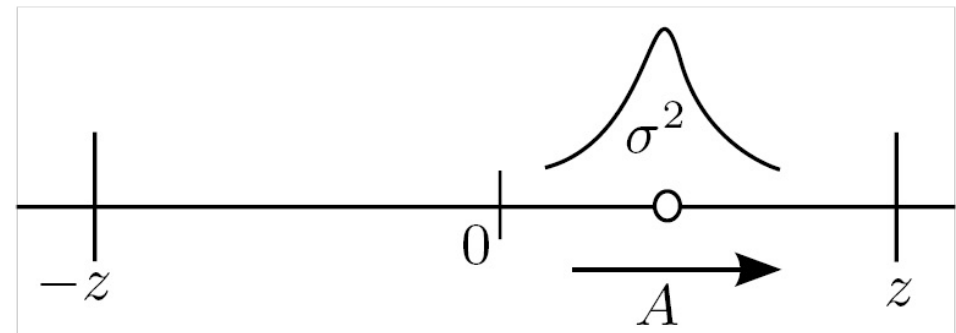
PATRICK HOGAN



KARL VON FRISCH



PATRICK HOGAN



JOURNAL
OF
THE ROYAL
SOCIETY

Interface

On optimal decision-making in brains and social insect colonies

James A. R. Marshall^{1,*}, Rafal Bogacz¹, Anna Dornhaus², Robert Planqué³,
Tim Kovacs¹ and Nigel R. Franks⁴

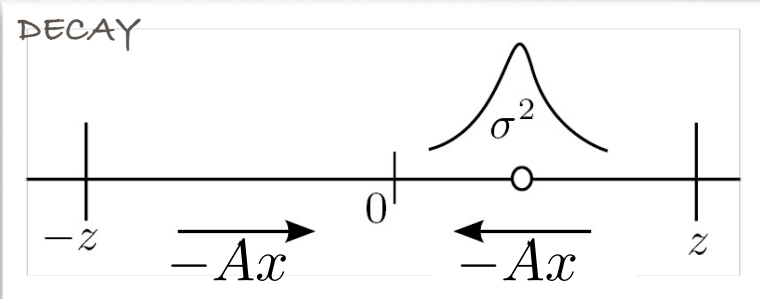
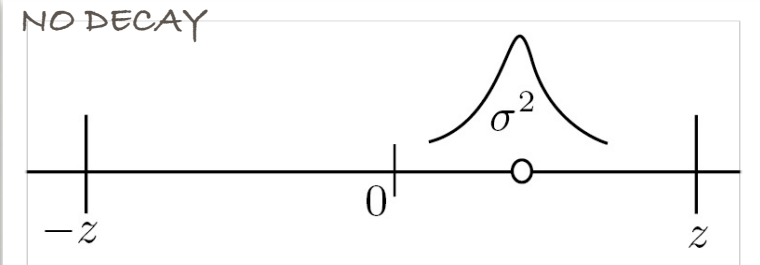
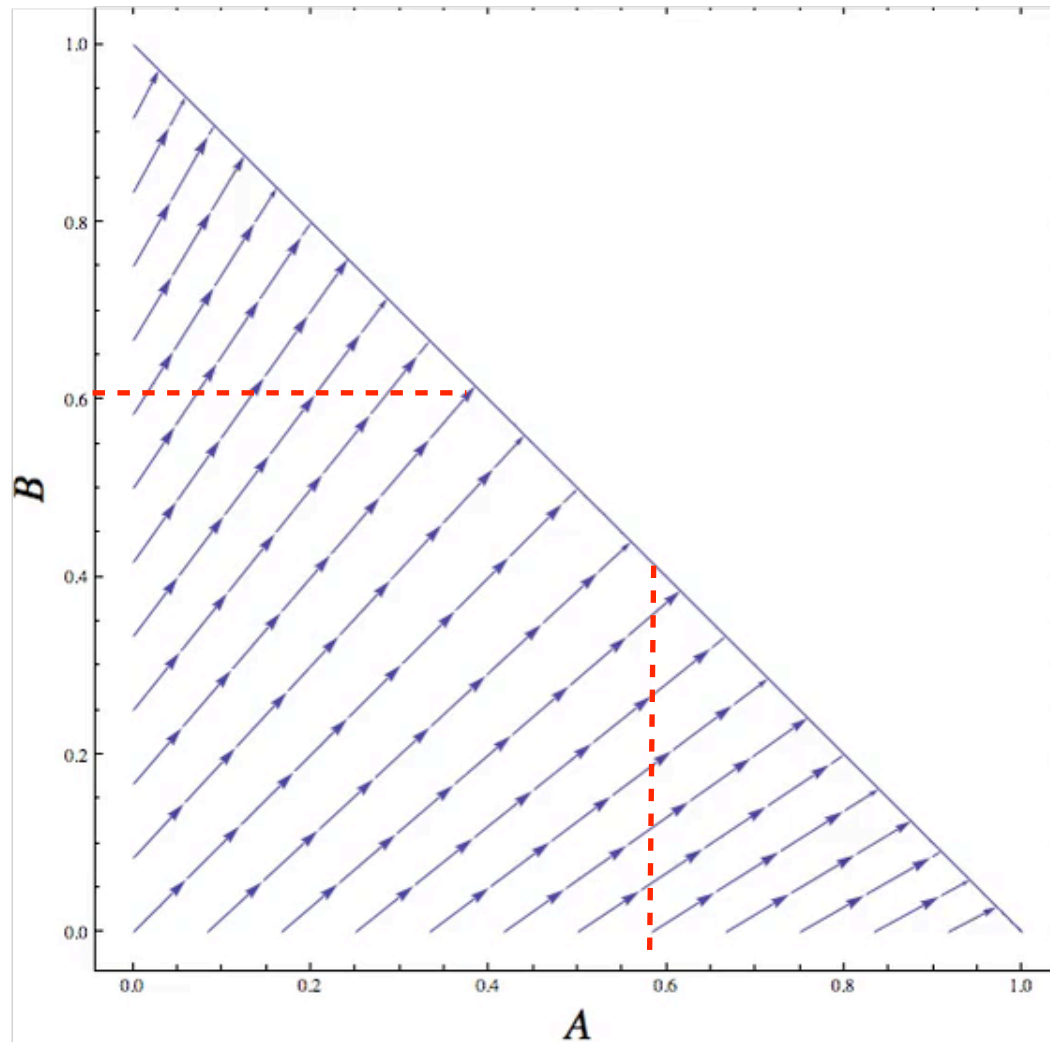
Department of Computer Science, University of Bristol, Woodland Road,

*J. R. Soc. Interface (2009) 6, 1065–1074
doi:10.1098/rsif.2008.0511
Published online 25 February 2009*





EQUAL NESTSITE QUALITIES



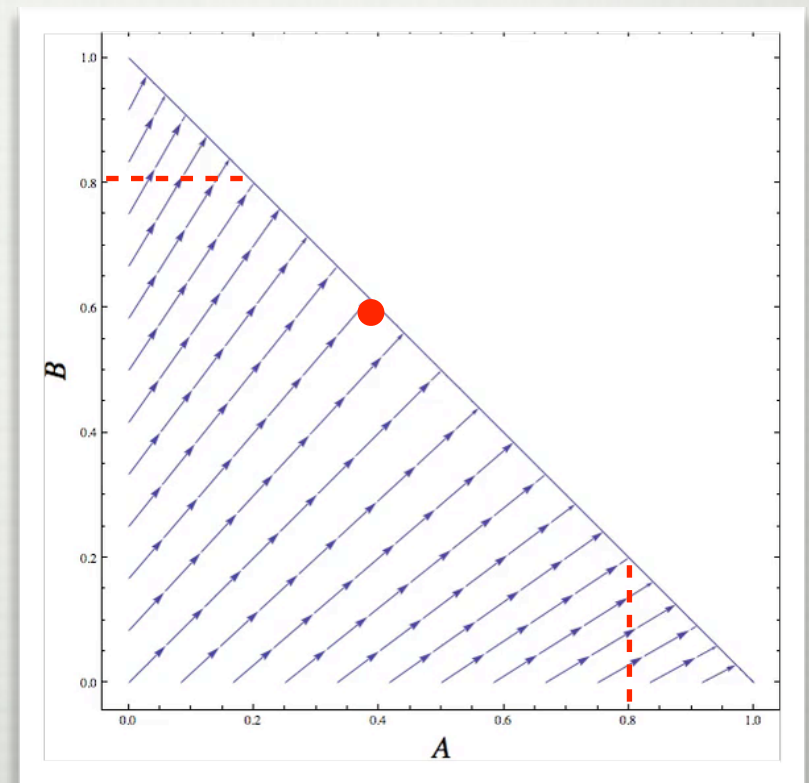
EXISTING NEURAL PROPOSALS

☐ COLLAPSING DECISION THRESHOLDS

☐ URGENCY SIGNAL

☐ TIME-DEPENDENT GAIN

☐ ASYMMETRIC INHIBITION



BEHAVIOR

How Honeybees Break a Decision-Making Deadlock

Jeremy E. Niven

For a honeybee swarm of potentially thousands of individuals, choosing a home is a momentous decision. Failing to choose a single location may cause the swarm to split and the queen to be lost (1); choosing poorly may limit the swarm's growth or expose it to freezing temperatures during the winter (2). Studies over the past 60 years have shown that honeybee swarms use quorum sensing, a form of decentralized decision-making, to choose a suitable nest site, but many gaps remain in our understanding of this process. On page 108 in this issue, Seeley *et al.* (3) show that an inhibitory signal between bees advocating different locations allows them to make a decision even when potential nest sites are equally favorable.

Honeybee colonies reproduce through budding, whereby the queen and some workers leave the nest and bivouac on a branch. Some of the most experienced workers leave to locate suitable nest sites (4). Upon their return, these scouts advertise potential locations and their qualities by performing a waggle dance. During the dance, the scout walks straight across the bivouacking bees, making side-to-side waggles of her body. She then stops, turns left or right, and walks a semicircular return path to her starting point. The waggle run's duration and orientation encode the length and the angle of the outward flight, respectively, whereas the number of dance circuits

encodes the quality of the potential nest site (5). Waggle dances recruit additional scouts to a site until a quorum number is reached and the swarm prepares to move to its new home (2).

Scouts advocating less attractive sites produce fewer dance circuits and make fewer trips to the site (6). Along with the recruitment of uncommitted scouts to more attractive sites, this was assumed to be sufficient to enable the bees to reach a quorum, thereby deciding which site to choose (2). However, foraging workers use an additional type of signal to communicate with other bees. Upon returning from a feeder that is crowded or where a predator is present, forager bees produce a brief vibrational signal that discourages other bees from producing waggle dances that advertise the location of that feeder (7). Hypothesizing that a similar signal may be used by house-hunting bees, Seeley *et al.* set out to observe scout behavior. They found that scouts received "stop" signals—head butts mainly to their head and thorax—from other bees during the return run of the waggle dance (see the figure). These stop signals occurred more frequently just before a scout stopped dancing.

The authors next established swarms on Appledore Island (Maine), which lacks natural nest sites, and gave them a choice of two identical nest boxes. Scouts visiting one box were marked with yellow paint; those visiting the other were marked with pink paint. Most of the bees giving "stop" signals

During the search for a new nest site, use of an inhibitory signal enables honeybees to reach a decision.

selection process, dancing scouts with yellow paint received many more stop signals from scouts with pink paint and vice versa, showing that scouts from one site preferentially inhibit the dances of those advertising a competing site (see the figure, panel A). Once the scouts started implementing the decision, dancing scouts received stop signals from scouts that had visited either site. When swarms were given only one nest box, scouts received few stop signals during the decision phase but many during the implementation phase. This general inhibition of dancing during the implementation phase presumably ensures that all the bees are present when the swarm takes flight.

To demonstrate a role for the observed cross inhibition between scouts advertising competing sites, Seeley *et al.* constructed a series of computational models of the collective decision-making process, based on the interaction rules they had observed among the scouts. Models that incorporated no or indiscriminate stop signaling predicted that the scouts would reach a stable deadlock, failing to choose between two

Cease and desist. (A) Seeley *et al.* have found that during house hunting, scouts advertising one nest site preferentially inhibit scouts advertising another site during the decision-making process. Inhibition is conveyed by a "stop" signal, given mainly to the head and thorax of a scout during the return phase of the waggle dance. (B) Stop signals from scout bees inhibit other scouts, discouraging them from advertising a potential site. These

BEHAVIOR

How Honeybees Break a Decision-Making Deadlock

During the search for a new nest site, use of an inhibitory signal enables honeybees to reach a decision.



THOMAS SCHLEGEL

SPONTANEOUS
ACTIVATION

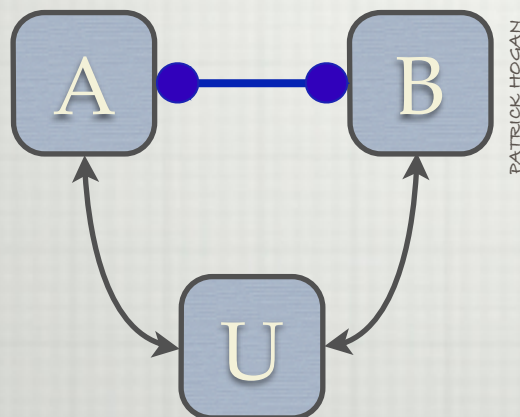
POSITIVE
FEEDBACK

DECAY

CROSS-
INHIBITION

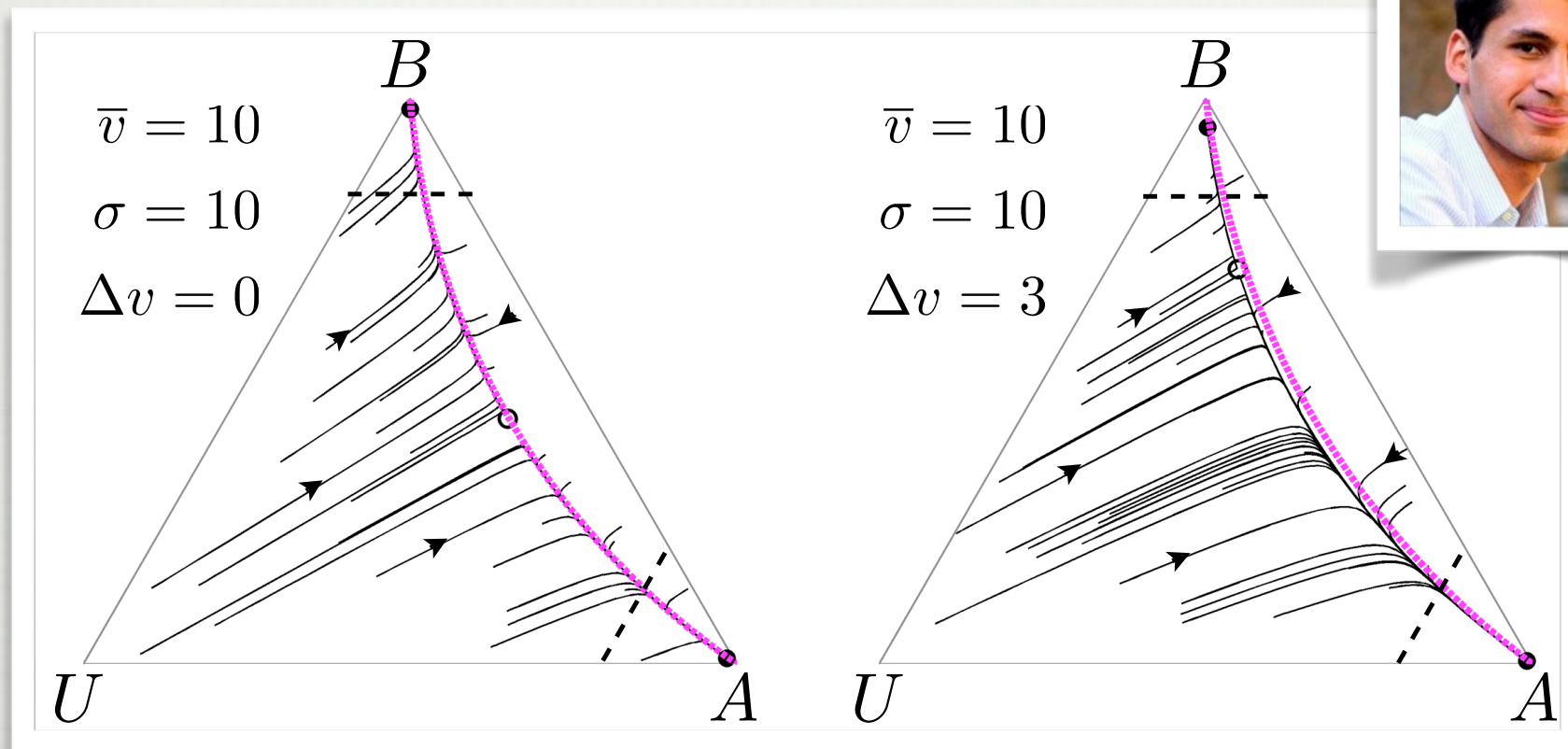
$$\begin{cases} dy_A := (y_U \gamma_A - y_A (\alpha_A - y_U \rho_A + y_B \sigma_B)) dt \\ \quad + k \sqrt{y_U^2 + y_A^2 + y_U y_A} dW_A \\ dy_B := (y_U \gamma_B - y_B (\alpha_B - y_U \rho_B + y_A \sigma_A)) dt \\ \quad + k \sqrt{y_U^2 + y_B^2 + y_U y_B} dW_B \end{cases}$$

PAIS ET AL. (2013)



ADDITIVE WHITE NOISE

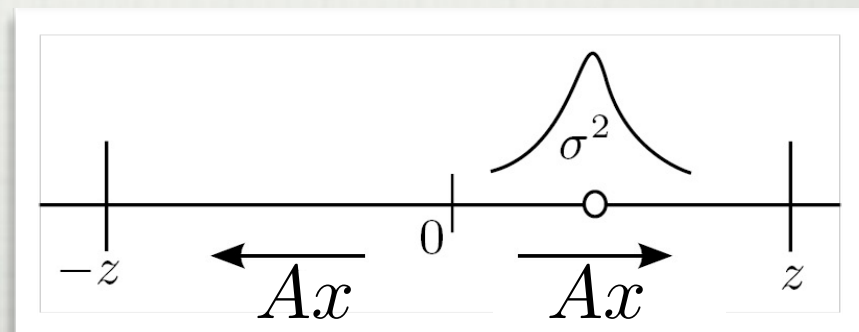
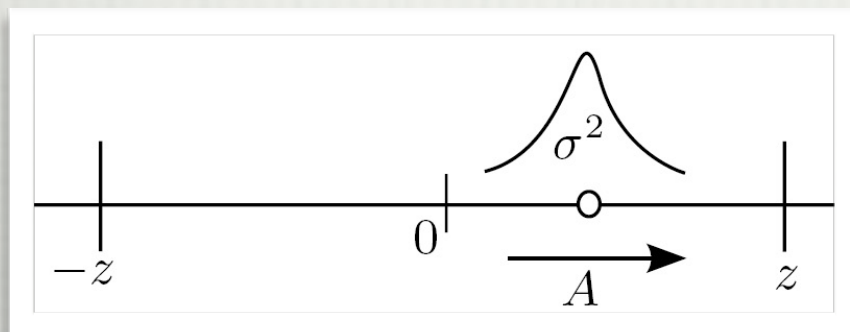
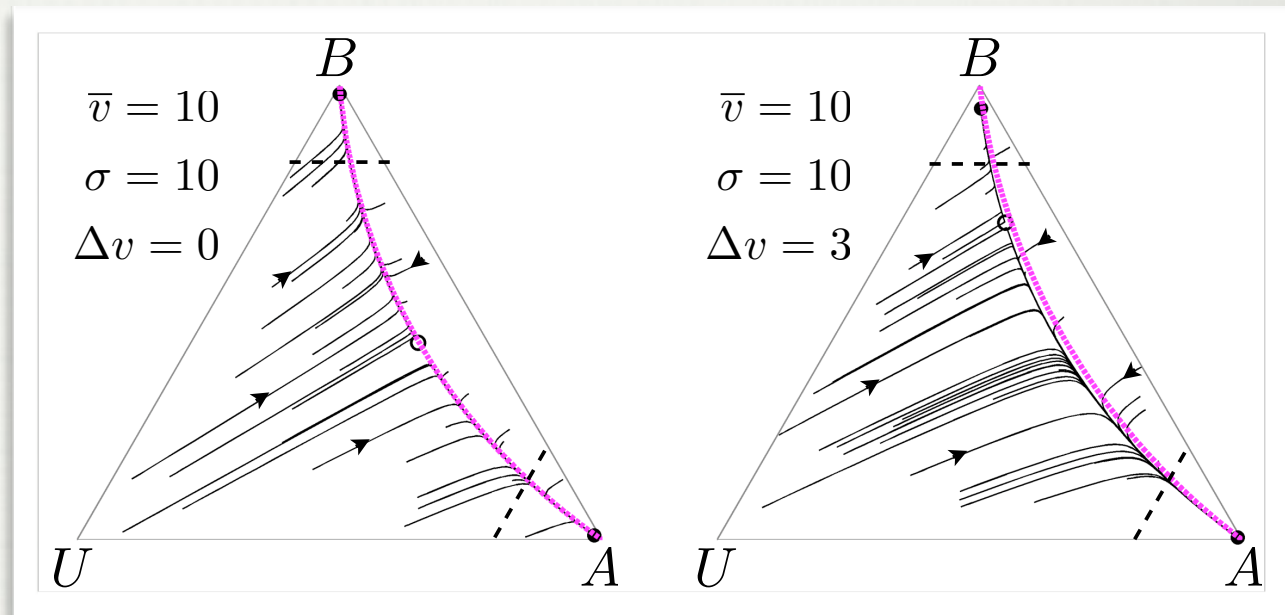
SEPARATION OF TIMESCALES



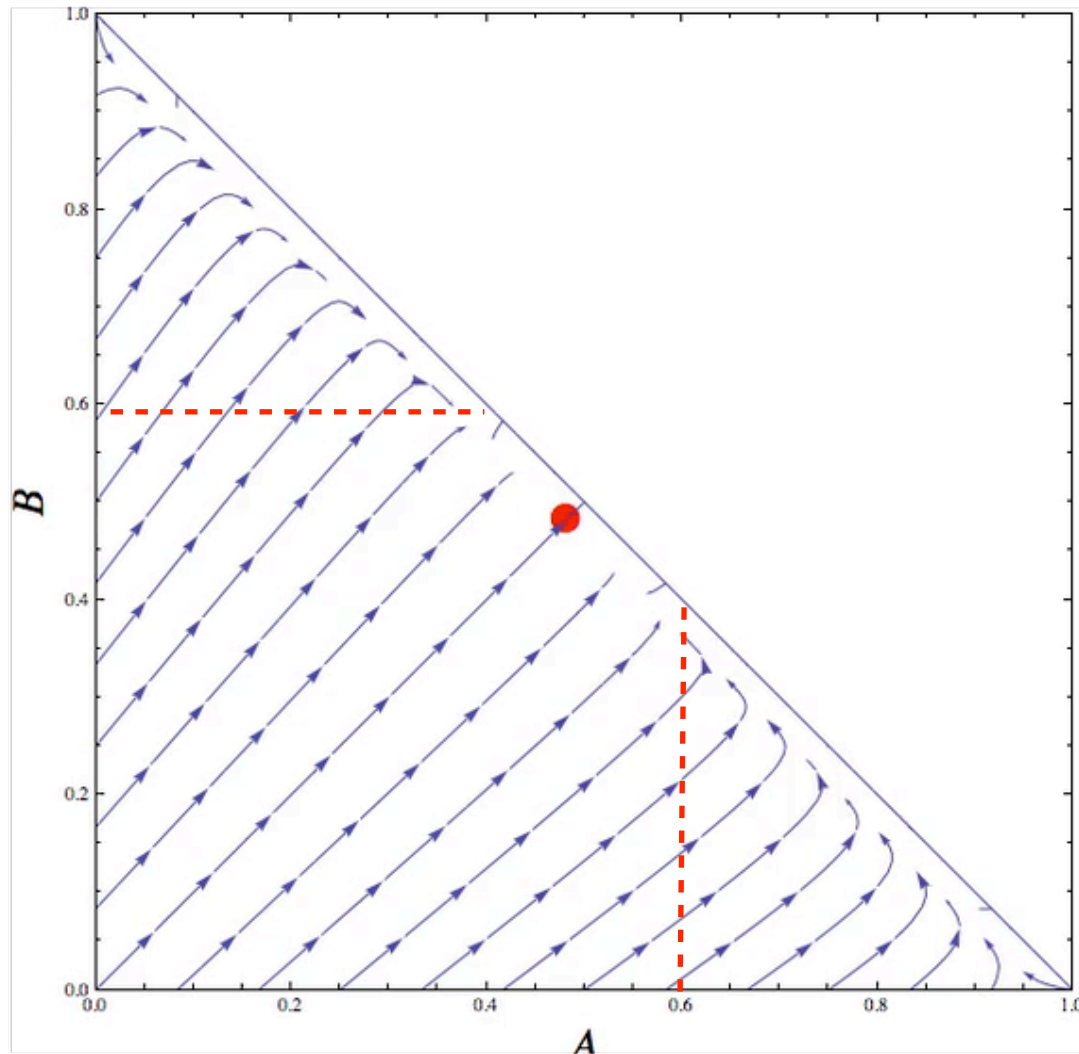
PAIS ET AL., 2013

\bar{v} LARGE, $\frac{\Delta v}{\bar{v}}$ SMALL

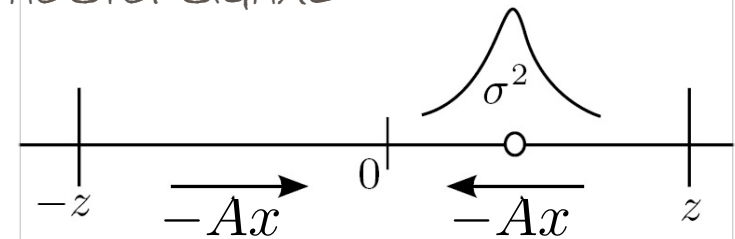
SEPARATION OF TIMESCALES



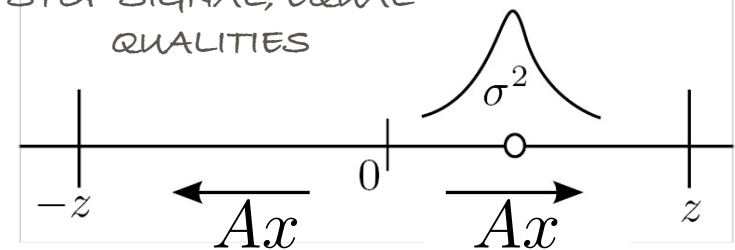
STOP-SIGNAL: INCREASING SIGNAL, EQUAL QUALITIES



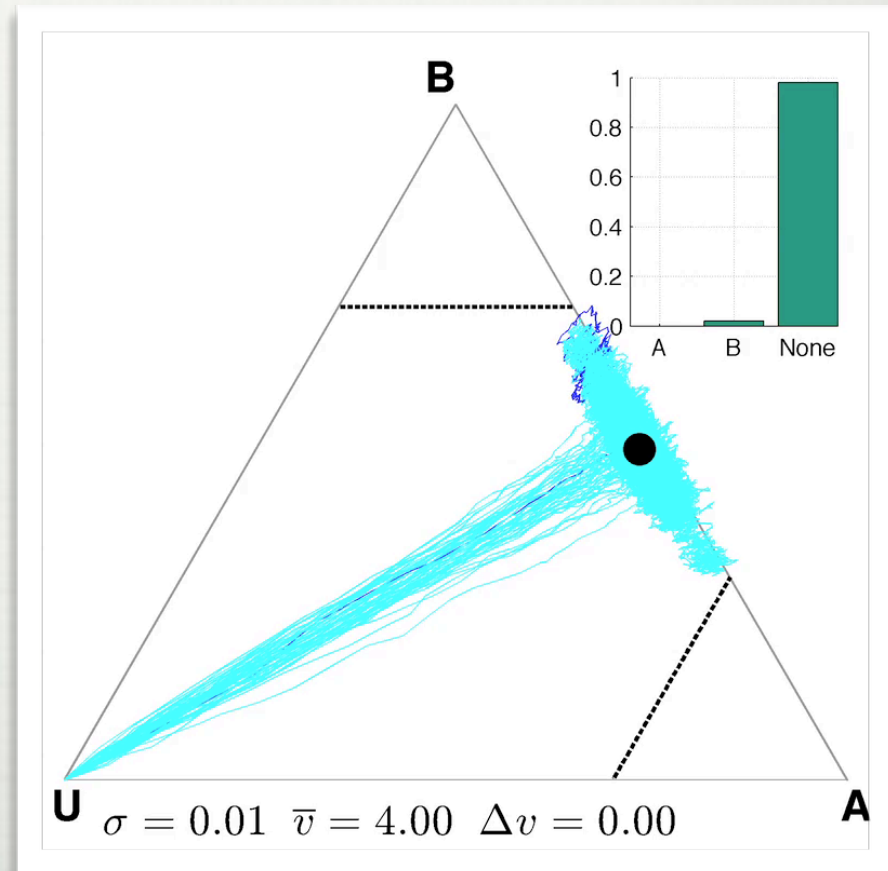
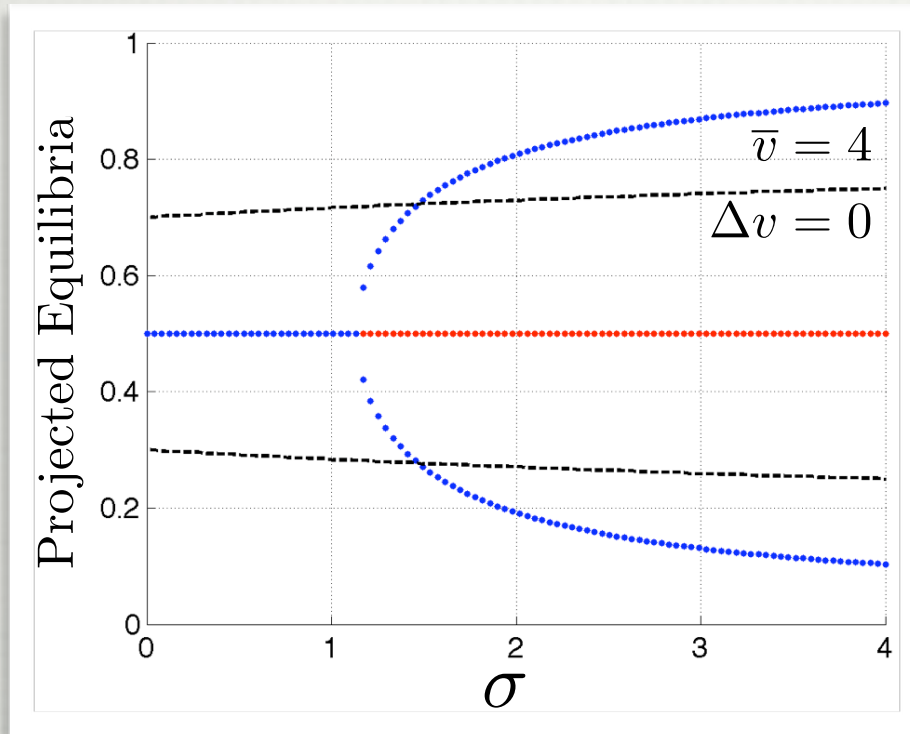
NO STOP-SIGNAL



STOP-SIGNAL, EQUAL
QUALITIES

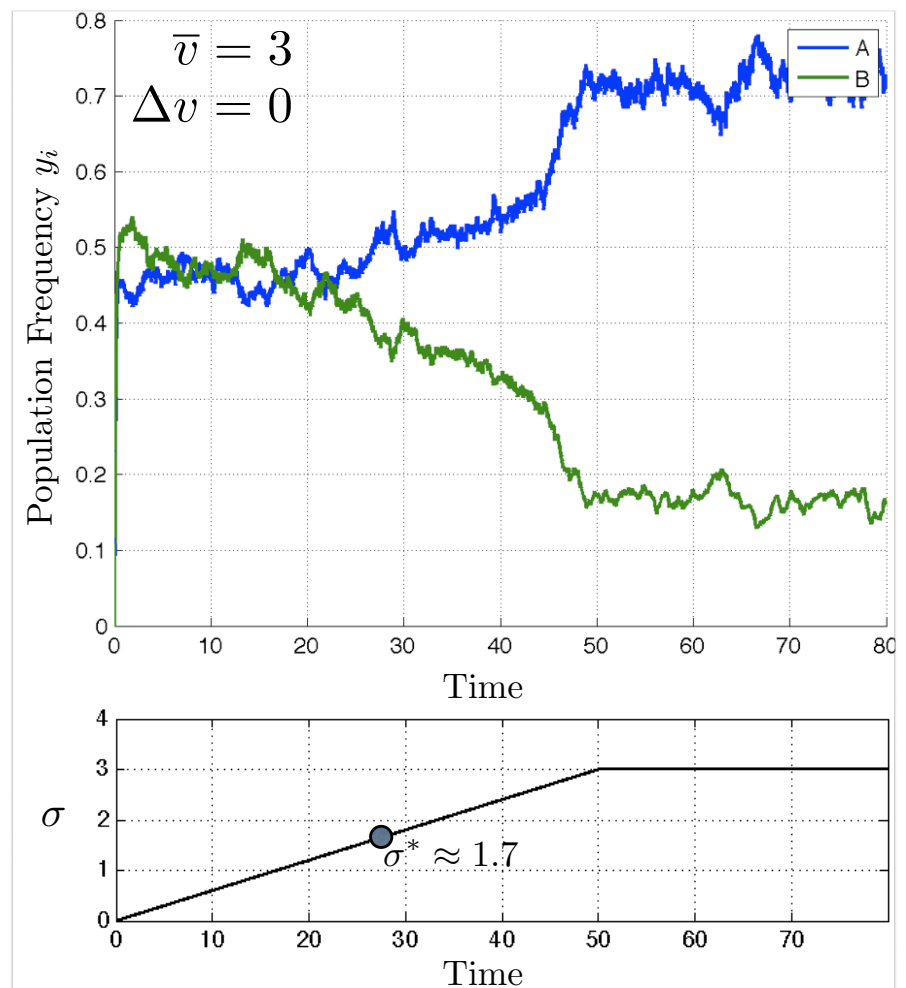
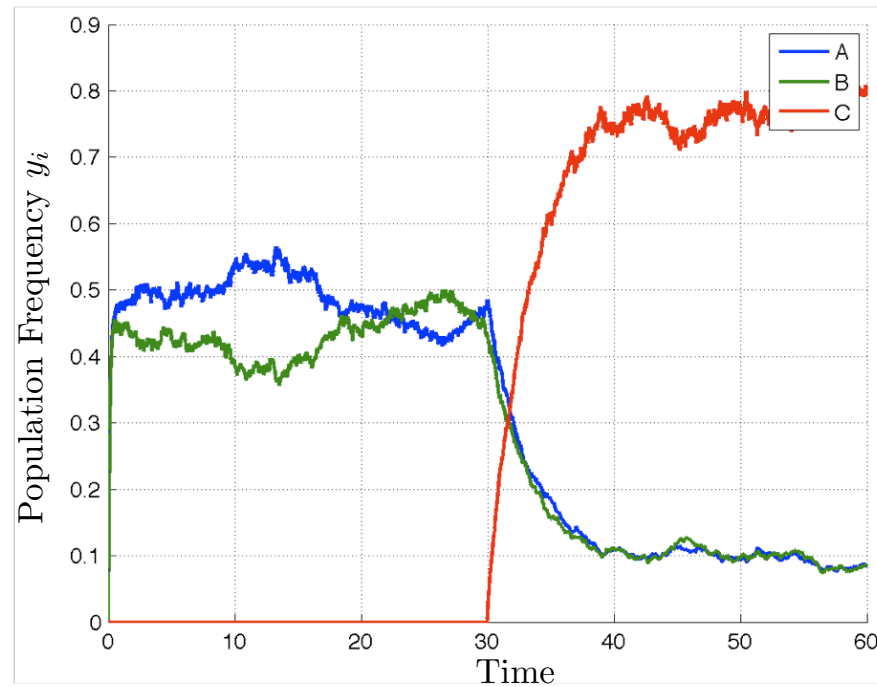


DEADLOCK BREAKING

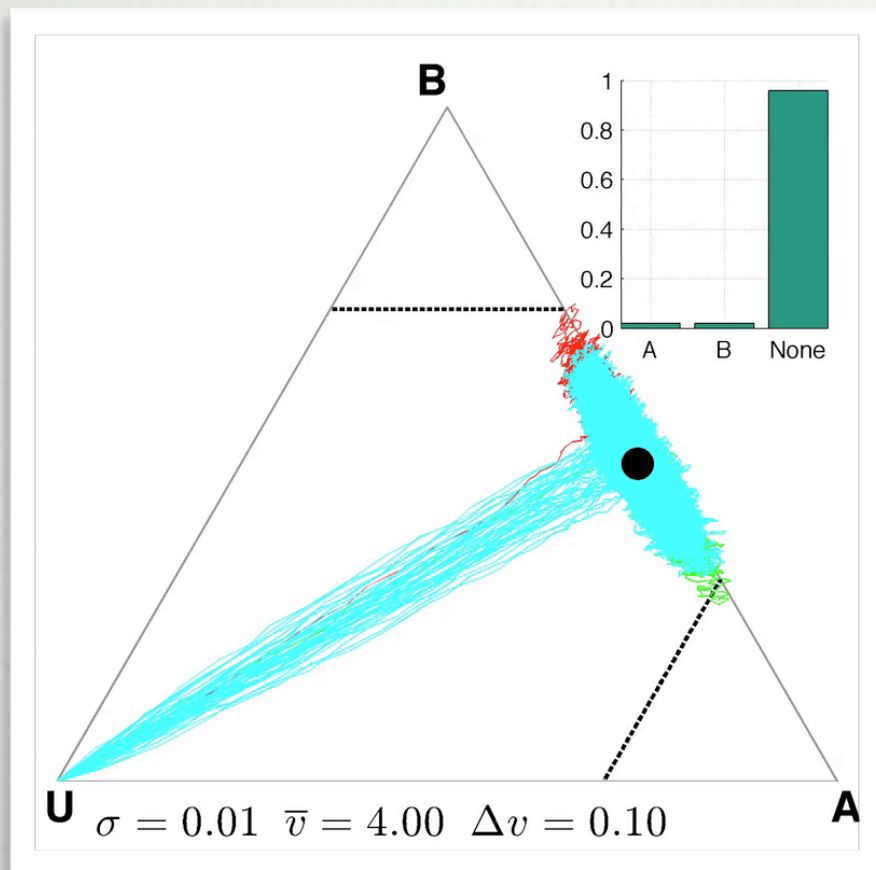


VALUE-SENSITIVE COLLECTIVE DECISION-MAKING

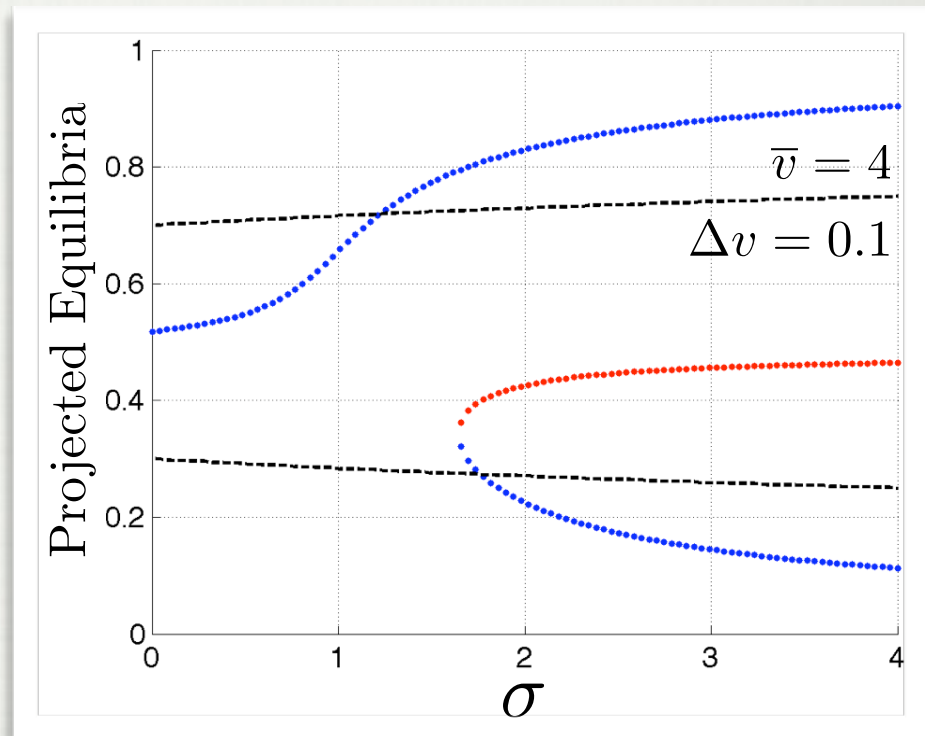
PAIS ET AL. (2013)



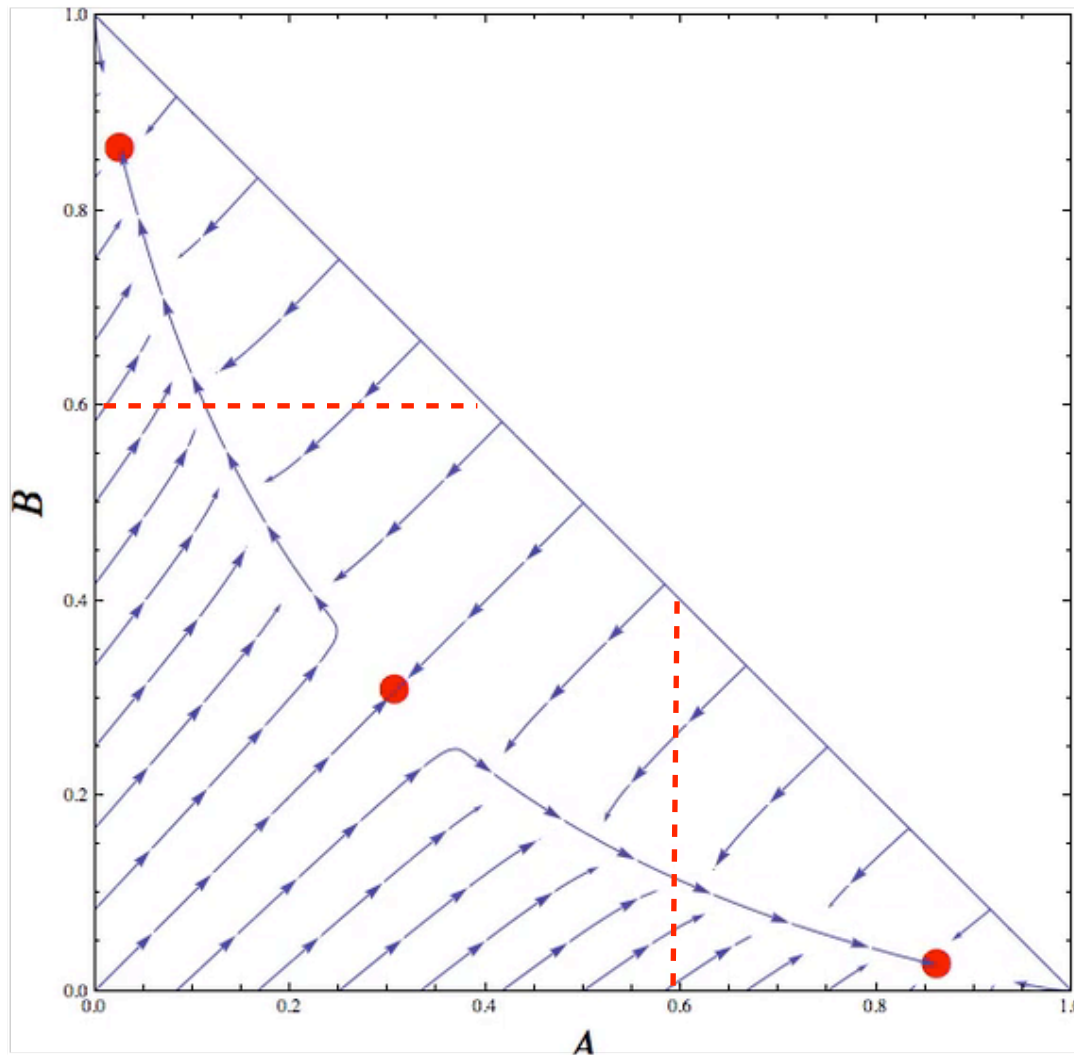
AMPLIFICATION



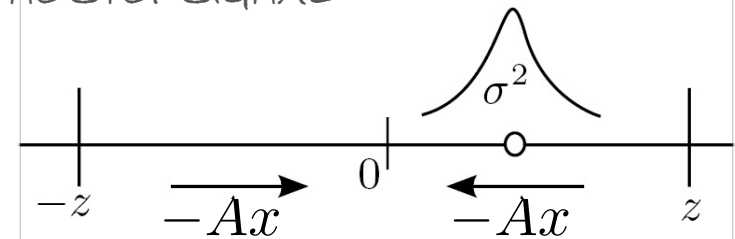
PAIS ET AL., 2013



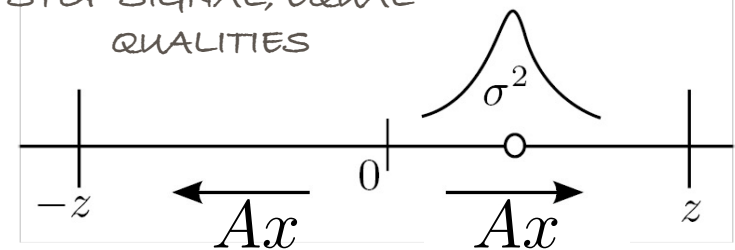
STOP-SIGNAL: INCREASING QUALITY DIFFERENCES



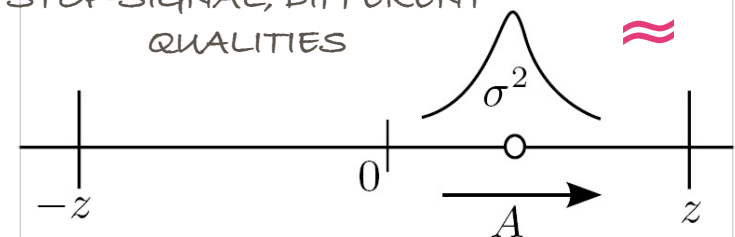
NO STOP-SIGNAL



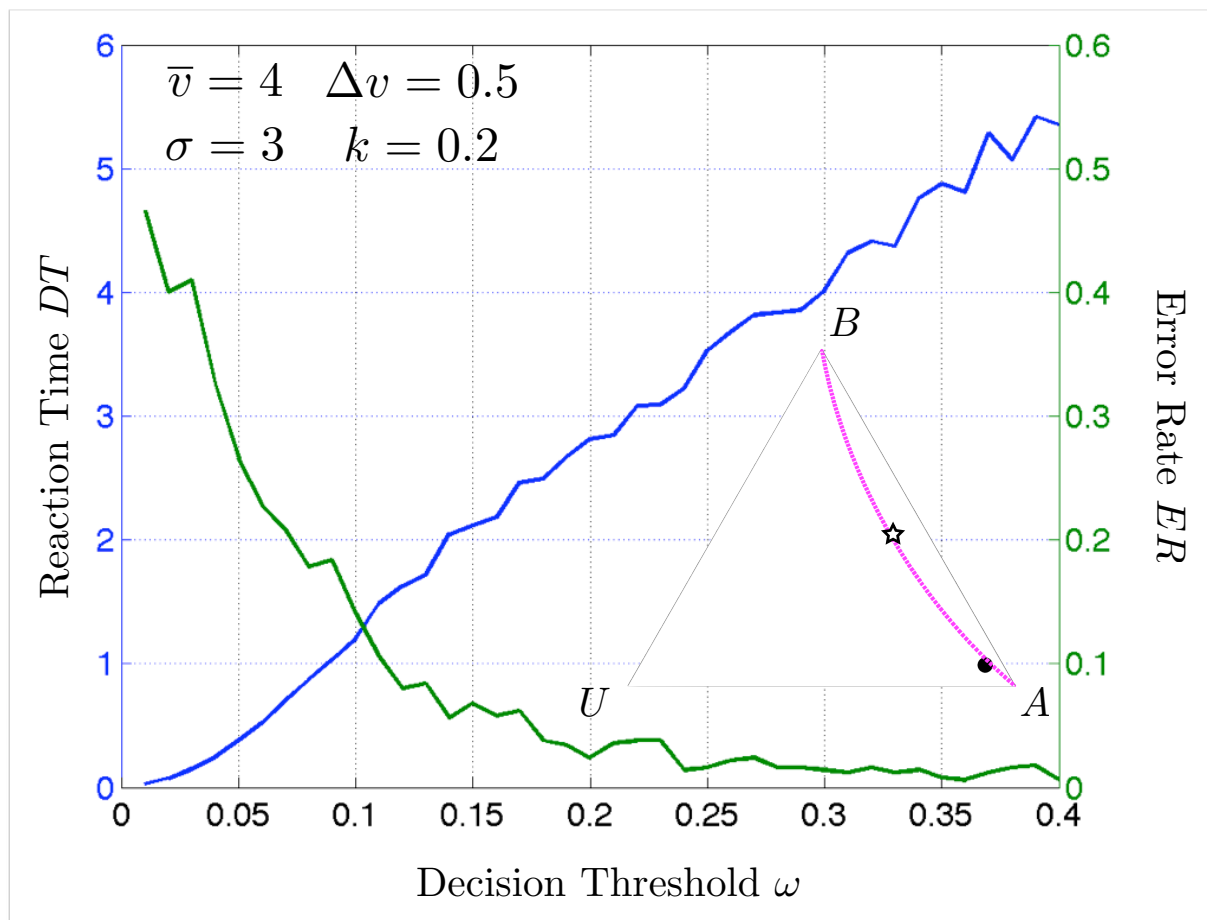
STOP-SIGNAL, EQUAL QUALITIES



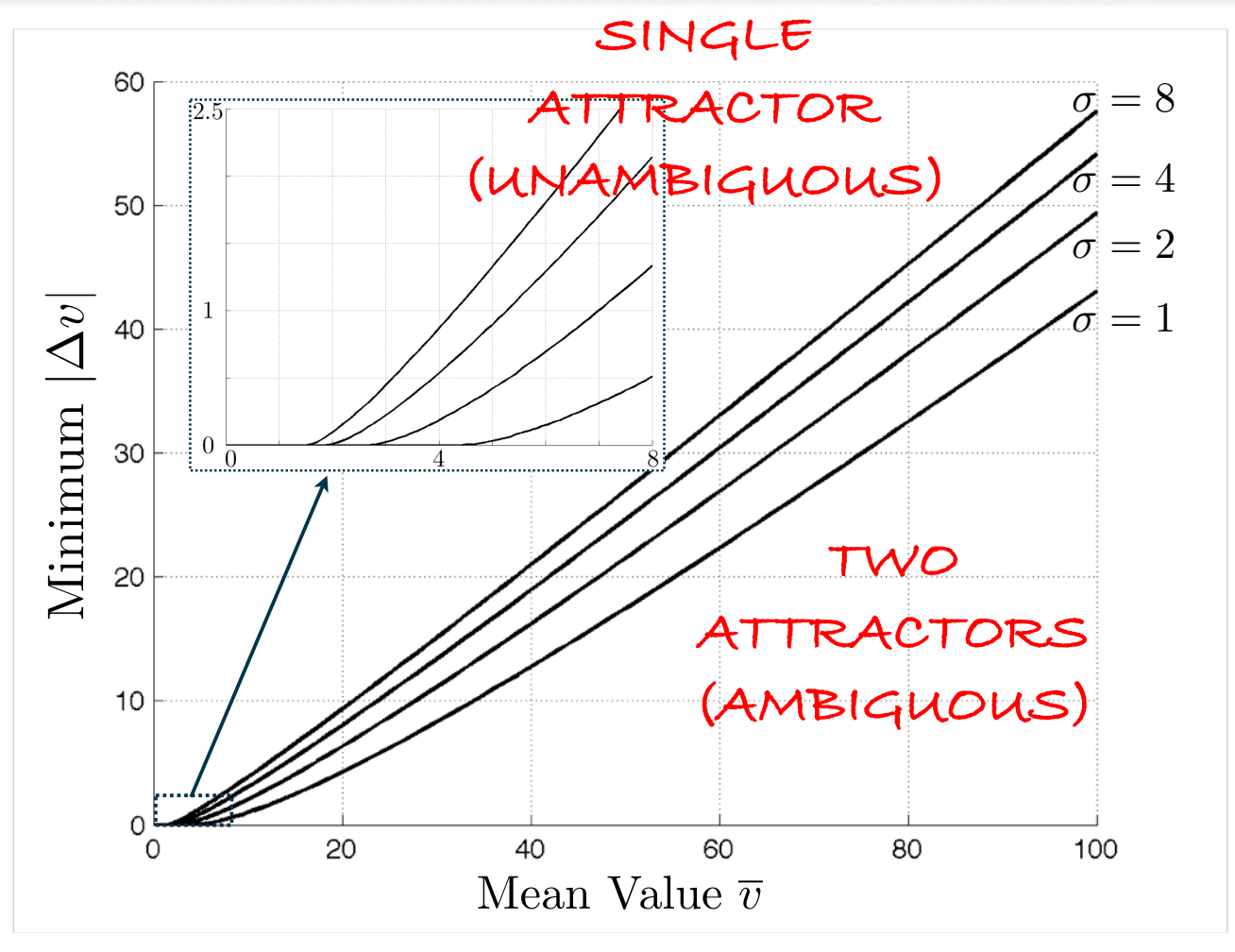
STOP-SIGNAL, DIFFERENT QUALITIES



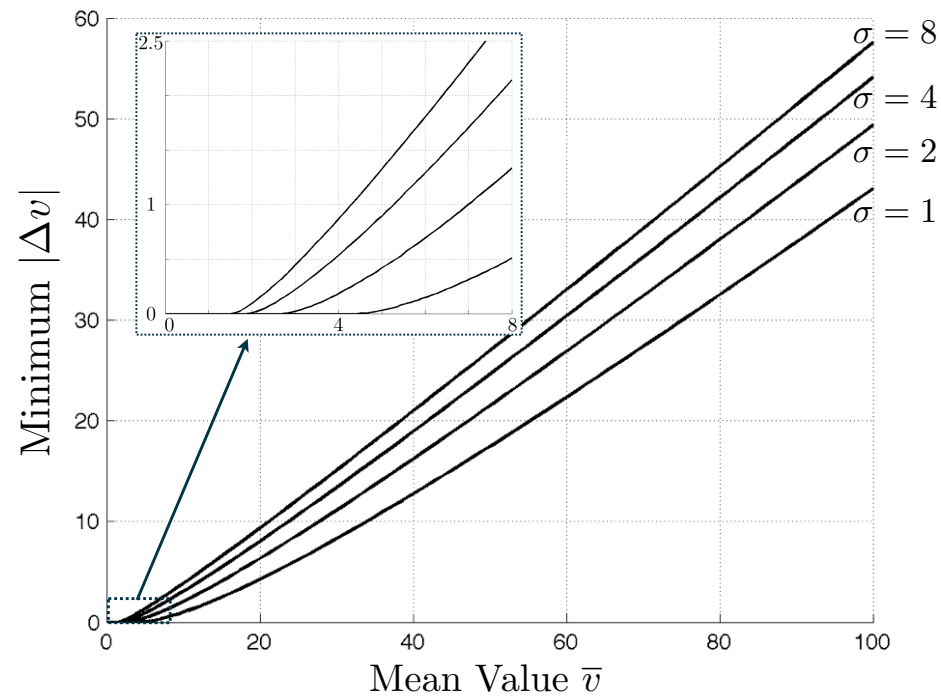
SPEED-ACCURACY TRADE-OFFS



JUST-NOTICEABLE DIFFERENCES



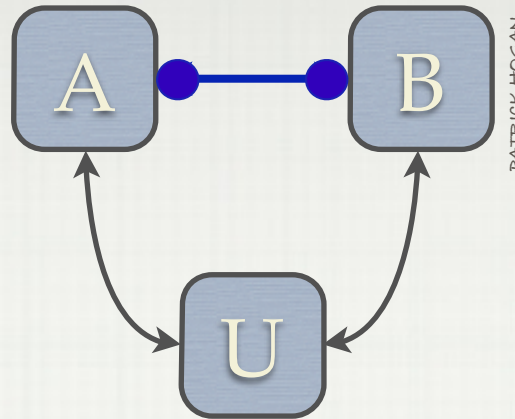
JUST-NOTICEABLE DIFFERENCES



$$\frac{\Delta v}{\bar{v}} = K$$

$f(\sigma)$

WEBER'S LAW



PATRICK HOGAN

ARTICLES

Salience driven value integration explains decision biases and preference reversal

Patrick Hogan^a and Marius Usher^{d,1}

^aDepartment of Experimental

Frontiers in Neuroscience

Visual fixation of value information

Ian Krajbich¹, Car

Most organisms face a process between the or about the role of fixations guide the explain complex

There is a growing brain makes decisions options under motivated an

rights reserved.

When Natural Selection Should Optimise Speed-Accuracy Trade-offs

Angelo Pirrone^{1,2}, Tom Stafford¹ and James A. R. Marshall^{2,3,*}

¹Department of Psychology, University of Sheffield, Sheffield, UNITED KINGDOM
²Kroto Research Institute, University of Sheffield, Sheffield, UNITED KINGDOM
³Department of Computer Science, University of Sheffield, Sheffield, UNITED KINGDOM

Correspondence*:
 James A. R. Marshall
 Behavioural and Evolutionary Theory Lab,
 Science, University of Sheffield,
 UNITED KINGDOM

Opinion
 20 March 2014

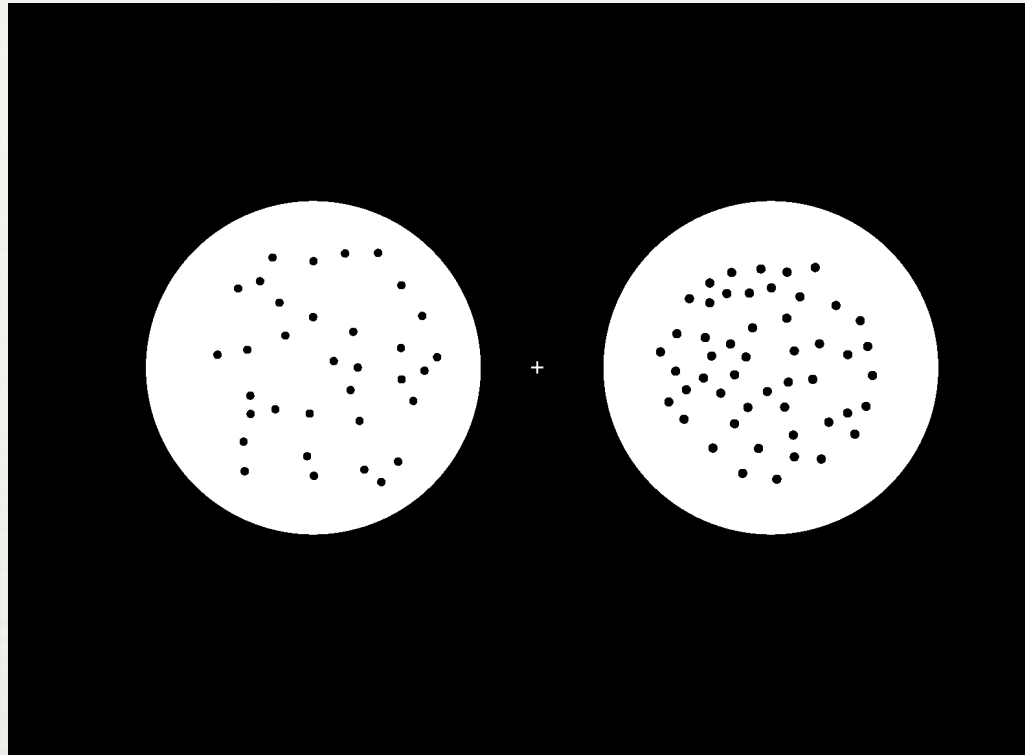


PSYCHOLOGICAL AND
 COGNITIVE SCIENCES

BEHAVIOURAL PREDICTIONS

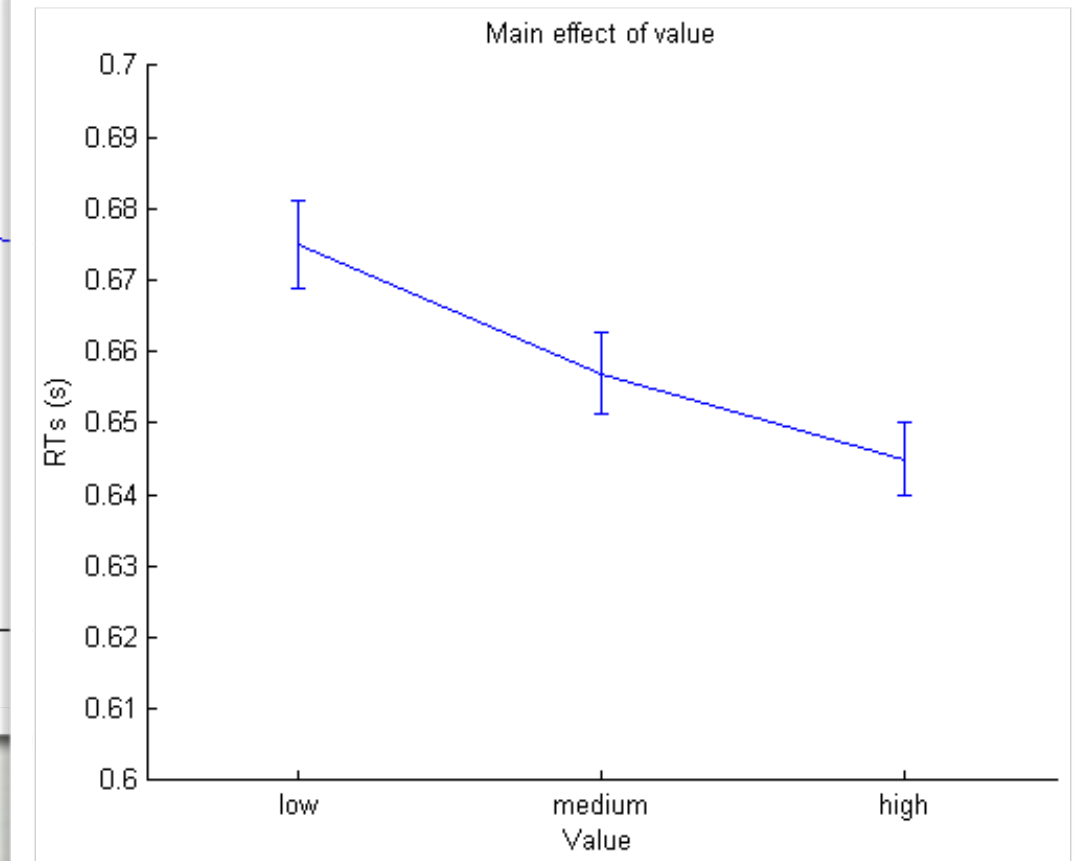
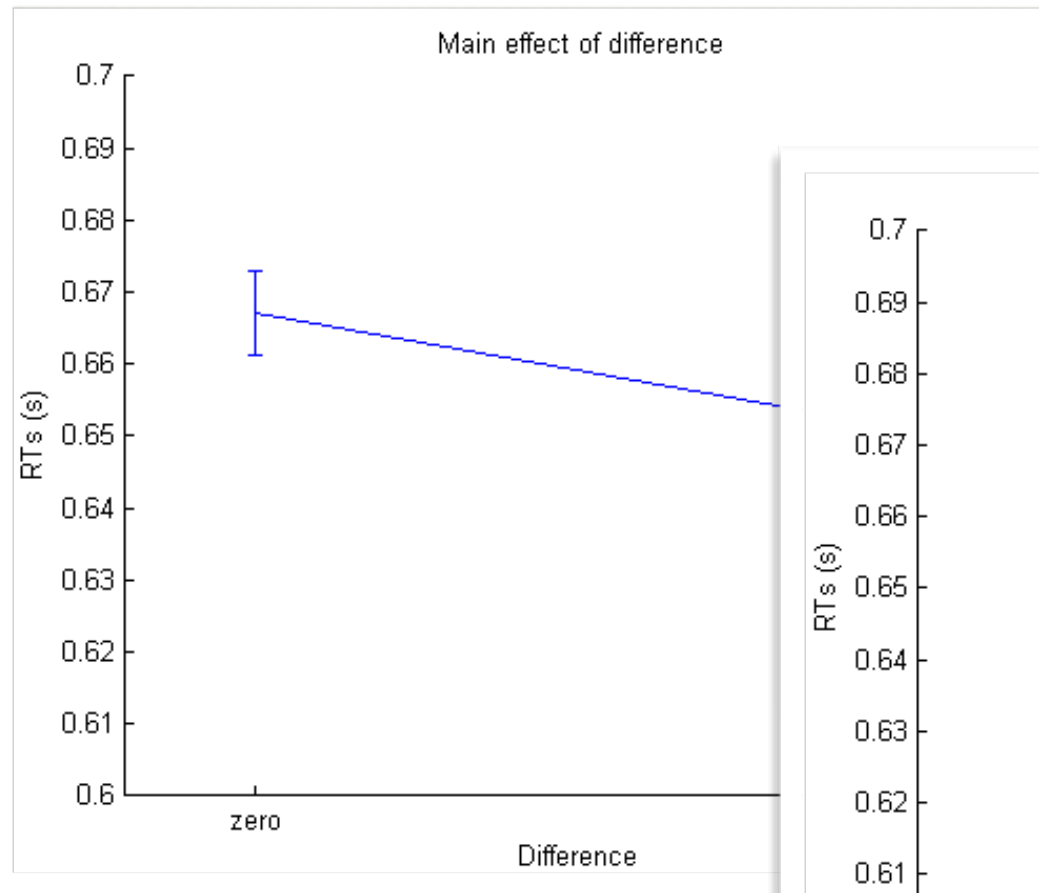
- ☐ DECISION-MAKERS IMPLEMENTING (SOMETHING LIKE) THIS MECHANISM SHOULD EXHIBIT:
 - ☐ WEBER'S LAW (STANDARD OBSERVATION)
 - ☐ SPEED-ACCURACY TRADE-OFFS (STANDARD OBS.)
 - ☐ REACTION-TIME SENSITIVITY TO ABSOLUTE VALUE
 - ☐ DECISION-MEMORY BASED ON PREVIOUS STIMULI VALUES

VALUE-SENSITIVITY IN HUMANS

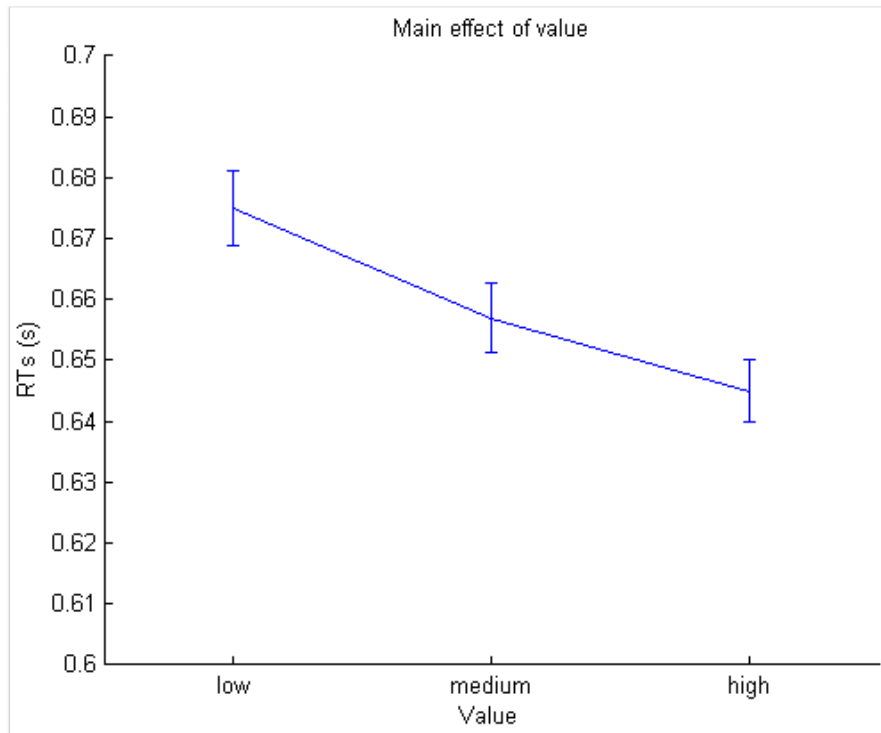


- ☐ TWO REWARD INSTRUCTIONS:
 - ☐ PERCEPTUAL: FIXED REWARD FOR CORRECT CHOICE, ZERO FOR INCORRECT
 - ☐ VALUE: REWARD PROPORTIONAL TO NUMBER OF DOTS

VALUE-SENSITIVITY IN HUMANS



VALUE-SENSITIVITY IN HUMANS



☐ INCONSISTENT WITH:

☐ SIMPLE DRIFT-DIFFUSION MODEL

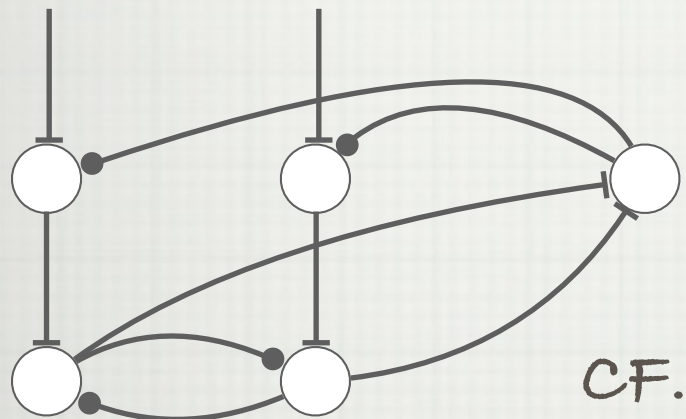
☐ CONSISTENT WITH:

☐ STOP-SIGNAL MODEL

☐ ACCUMULATOR MODELS

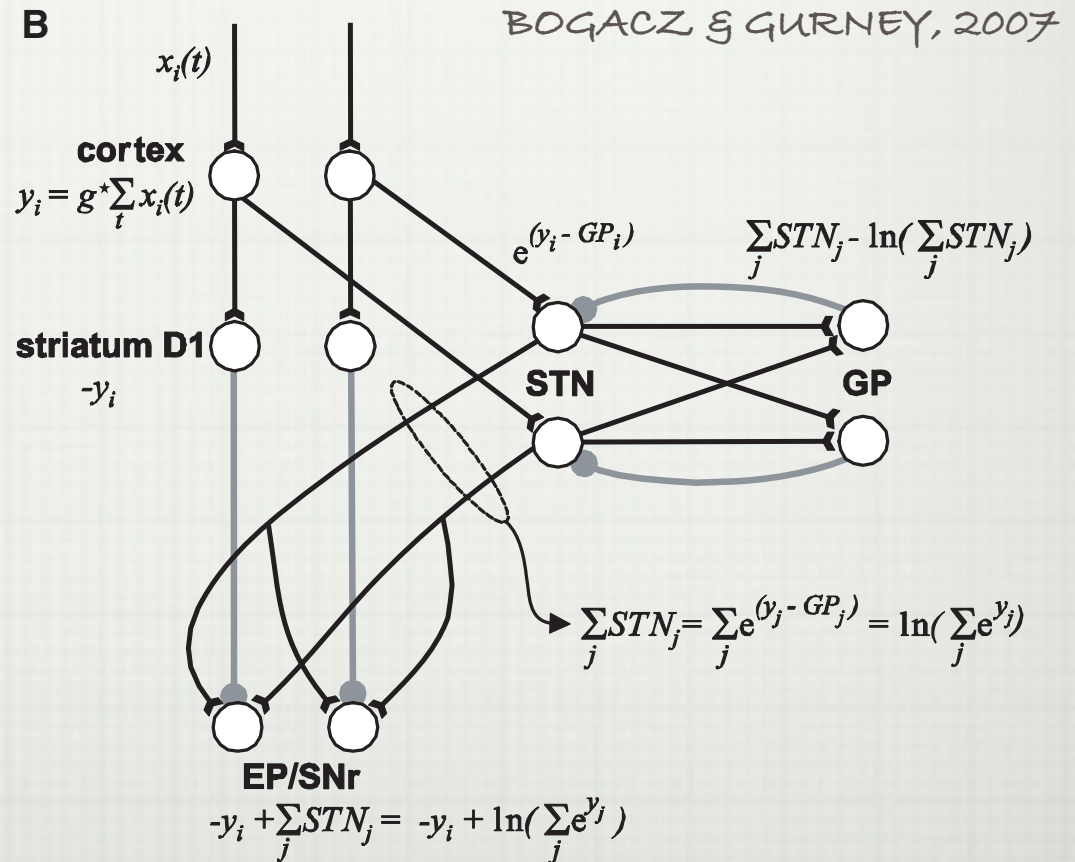
☐ RACE/SEQUENTIAL CHOICE MODELS (E.G. KACELNIK ET AL.)

LOOKING FOR IMPLEMENTATIONS OF VALUE-SENSITIVE DECISION-MAKING



CF.

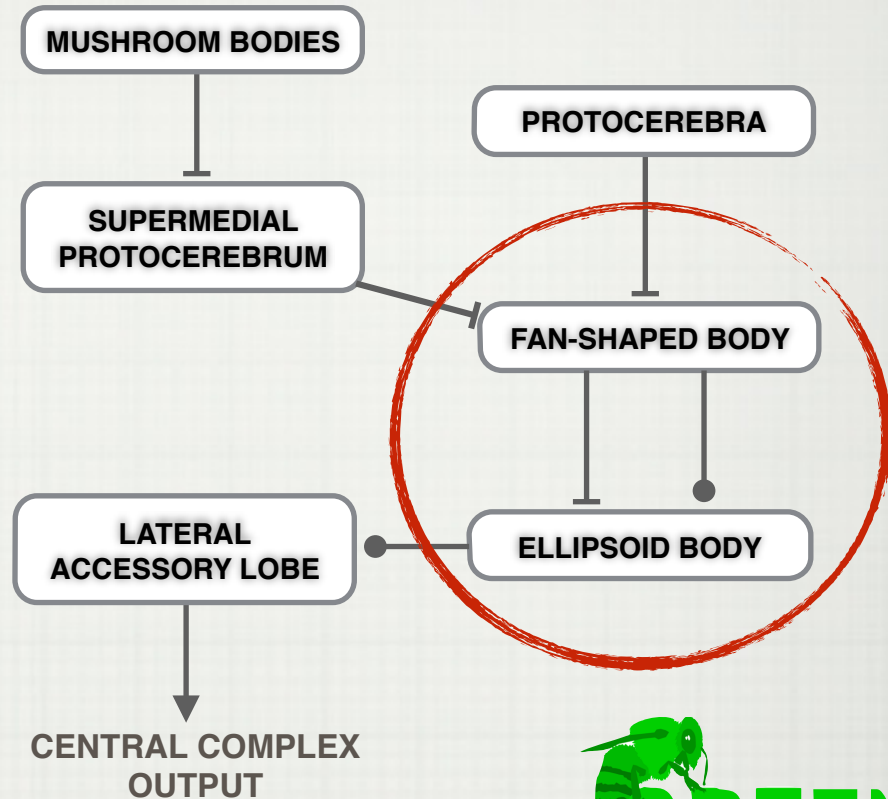
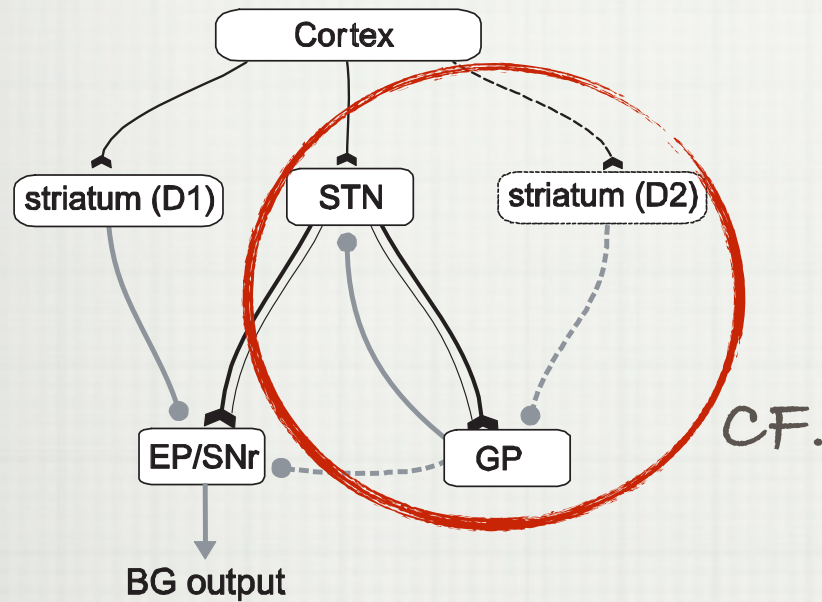
—| EXCITATION
—• INHIBITION



VERTEBRATE

ARTHROPOD

A



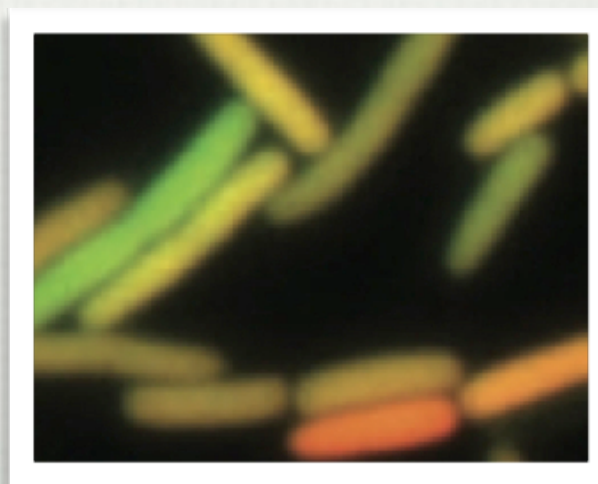
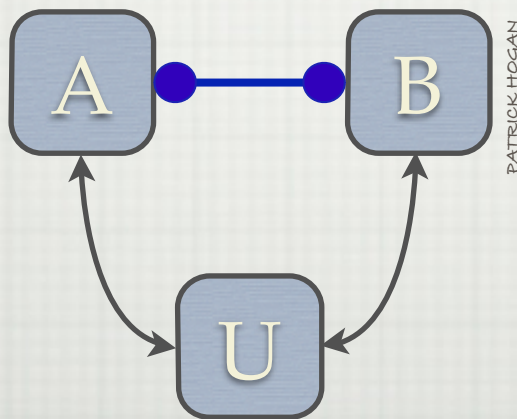
BOGACZ & GURNEY, 2007

STRAUSFELD & HIRTH, 2013

LOOKING FOR IMPLEMENTATIONS OF VALUE-SENSITIVE DECISION-MAKING

$$\begin{cases} dy_A &:= (y_U \gamma_A - y_A(\alpha_A - y_U \rho_A + y_B \sigma_B)) dt \\ &+ k \sqrt{y_U^2 + y_A^2 + y_U^2 y_A^2} dW_A \\ dy_B &:= (y_U \gamma_B - y_B(\alpha_B - y_U \rho_B + y_A \sigma_A)) dt \\ &+ k \sqrt{y_U^2 + y_B^2 + y_U^2 y_B^2} dW_B \end{cases}$$

PAIS ET AL. (2013)



PAYBACK FOR ENGINEERING?



- ☐ AS WELL AS DESIRABLE DECISION PROPERTIES, STOP-SIGNAL MODEL IMPLEMENTS ENERGY EFFICIENT DESIGN-MAKING
- ☐ SHOULD BE SUITABLE FOR CONTROLLER DESIGN FOR COLLECTIVE BEHAVIOUR



scentro

ACKNOWLEDGEMENTS & REFERENCES

- ☐ THOMAS SCHLEGEL, TOM SEELEY AND KIRK VISSCHER (EXPERIMENTS)
- ☐ PATRICK HOGAN (STOP-SIGNAL MODEL)
- ☐ DARREN PAIS AND NAOMI LEONARD (DYNAMICAL SYSTEMS ANALYSES)
- ☐ ANGELO PIRRONE AND TOM STAFFORD (PSYCHOPHYSICS EXPERIMENTS)
- ☐ ALEX COPE (GREEN BRAIN MODELS)
- ☐ SEELEY ET AL. (2012) SCIENCE 335, 108-111
 - ☐ NIVEN (2012) SCIENCE 335, 43-44
- ☐ PAIS ET AL. (2013) PLOS ONE 8: E73216
- ☐ PIRRONE ET AL. (2014) FRONTIERS IN NEUROSCIENCE 08:73. DOI: 10.3389/FNINS.2014.00073
- ☐ [HTTP://GREENBRAINPROJECT.CO.UK](http://GREENBRAINPROJECT.CO.UK)



HYSTERESIS

PAIS ET AL., 2013

