

# Lecture 2: Bayesian Games

Megan Peters  
Larry Maloney



CoSMo 2018

Minneapolis, MN

Why are you so wonderful?

EON	ERA	PERIOD	EPOCH	
Phanerozoic	Cenozoic	Quaternary	Holocene	Now
			Pleistocene	Late Early
		Tertiary	Pliocene	Late Early
				Late Middle Early
			Miocene	Late Middle Early
				Late Early
			Oligocene	Late Early
				Late Middle Early
			Eocene	Late Middle Early
				Late Early
			Paleocene	Late Early
	Mesozoic	Cretaceous	Late Early	
			Late Middle Early	
		Jurassic	Late Middle Early	
			Late Middle Early	
			Late Middle Early	
		Triassic	Late Middle Early	
			Late Middle Early	
			Late Middle Early	
		Permian	Late Early	
	Paleozoic	Pennsylvanian		
		Mississippian		
		Devonian	Late Middle Early	
			Late Middle Early	
			Late Middle Early	
		Silurian	Late Early	
			Late Middle Early	
		Ordovician	Late Middle Early	
			Late Middle Early	
			Late Middle Early	
		Cambrian	D C B A	

500 million years

M F Land & D-E Nilson, (2002)  
 Animal Eyes. Oxford.

EON	ERA	PERIOD	EPOCH	
Phanerozoic	Cenozoic	Quaternary	Holocene	Now
			Pleistocene	Late Early
		Tertiary	Pliocene	Late Early
				Late Middle Early
			Miocene	Late Middle Early
				Late Early
			Oligocene	Late Early
				Late Middle Early
			Eocene	Late Middle Early
				Late Early
			Paleocene	Late Early
	Mesozoic	Cretaceous	Late Early	
			Late Middle Early	
		Jurassic	Late Middle Early	
			Late Middle Early	
			Late Middle Early	
		Triassic	Late Middle Early	
			Late Middle Early	
			Late Early	
		Permian	Late Early	
	Paleozoic	Pennsylvanian		
		Mississippian		
		Devonian	Late Middle Early	
			Late Early	
			Late Early	
		Silurian	Late Early	
			Late Middle Early	
		Ordovician	Late Middle Early	
			D C B A	
		Cambrian		

# Cambrian explosion



anomalocaridid

eyes

M F Land & D-E Nilson, (2002)  
**Animal Eyes.** Oxford.

EON	ERA	PERIOD	EPOCH	
Phanerozoic	Cenozoic	Quaternary	Holocene	Now
			Pleistocene	Late Early
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			Late Middle Early	
			Late Middle Early	
		Triassic	Late Middle Early	
			Late Middle Early	
			Late Middle Early	
		Permian	Late Early	
			Late Early	
	Paleozoic	Pennsylvanian		
		Mississippian		
		Devonian	Late Middle Early	
			Late Middle Early	
			Late Middle Early	
		Silurian	Late Early	
			Late Early	
		Ordovician	Late Middle Early	
			Late Middle Early	
			Late Middle Early	
		Cambrian	D C B A	

# Cambrian explosion

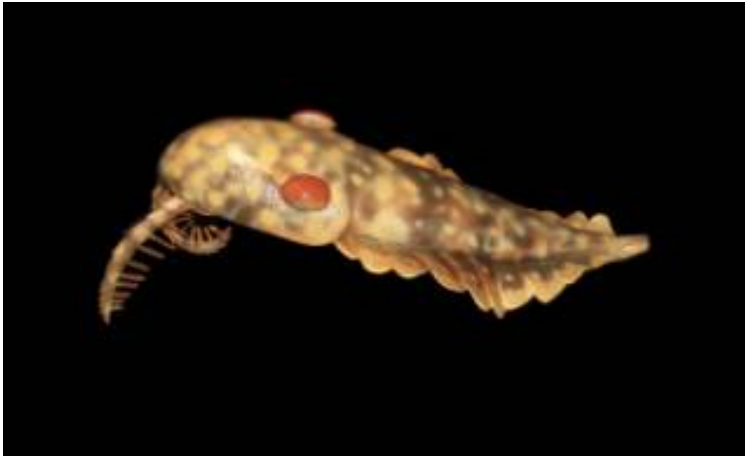


anomalocaridid

eyes  
skeletons

M F Land & D-E Nilson, (2002)  
**Animal Eyes.** Oxford.

# Cambrian explosion



eyes  
skeletons  
muscles + levers

perception  
movement

M F Land & D-E Nilson, (2002)  
**Animal Eyes.** Oxford.



# Cambrian explosion



eyes  
skeletons  
muscles + levers

perception  
movement

“arms race”

M F Land & D-E Nilson, (2002)  
**Animal Eyes.** Oxford.

# Cambrian explosion



*better eyes*

*better skeletons*

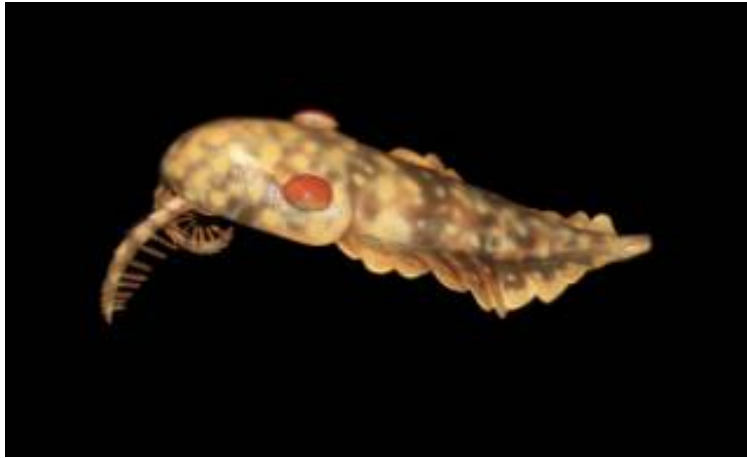
muscles + levers

“arms race”

M F Land & D-E Nilson, (2002)  
**Animal Eyes.** Oxford.



# Cambrian explosion



perception

action

*better eyes*

*better skeletons*

muscles + levers

*something missing ...*

M F Land & D-E Nilson, (2002)  
**Animal Eyes.** Oxford.

# Cambrian explosion

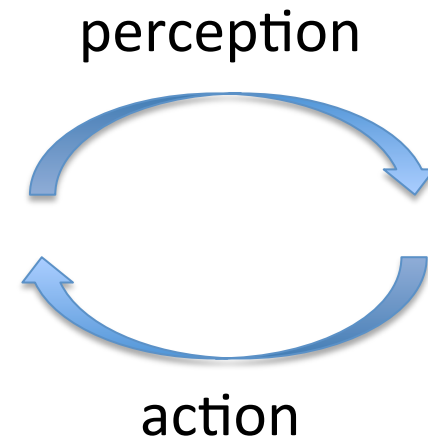


*better eyes*

*better skeletons*

muscles + levers

*better action selection*

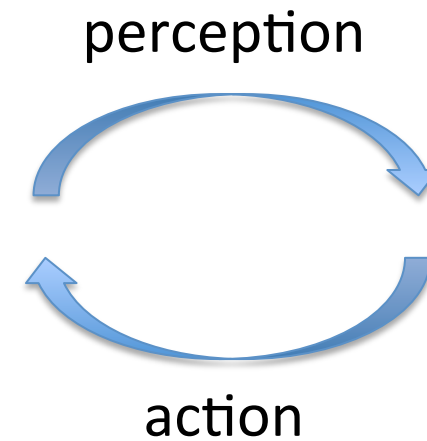


M F Land & D-E Nilson, (2002)  
**Animal Eyes.** Oxford.

action selection

movement planning

***decision***

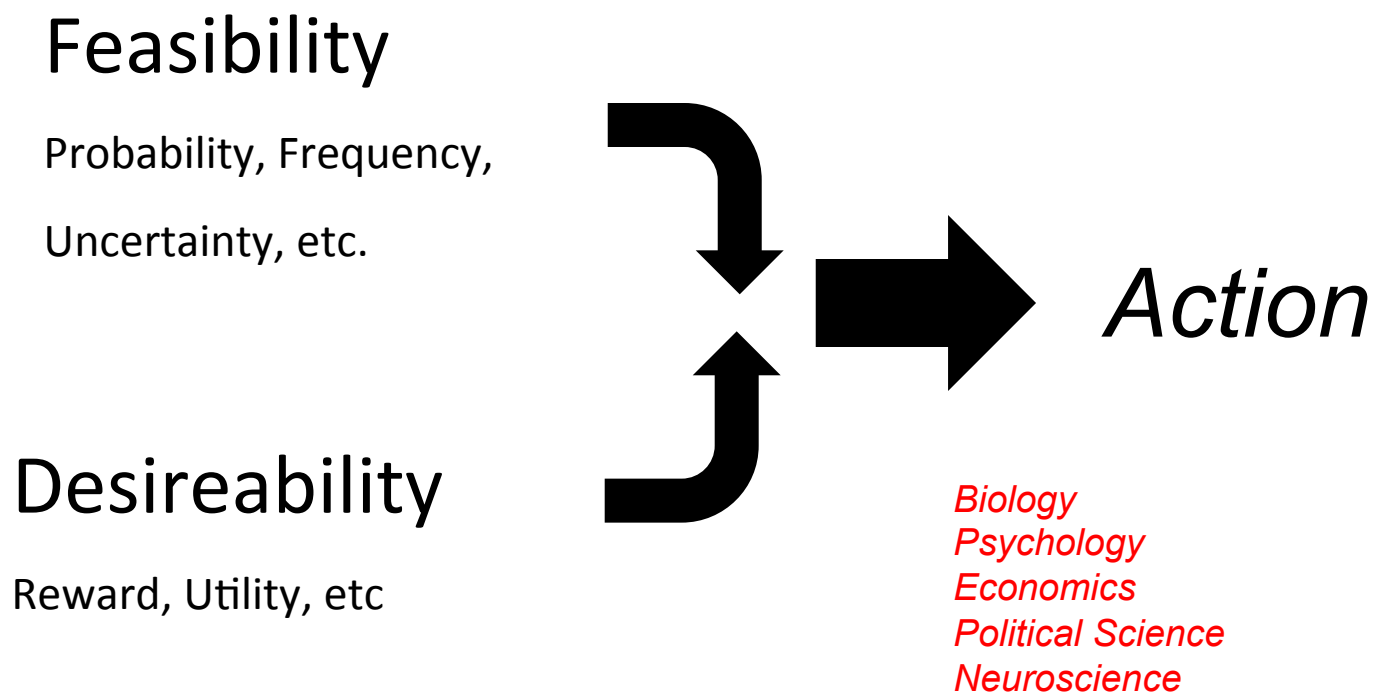


500 million years later ...

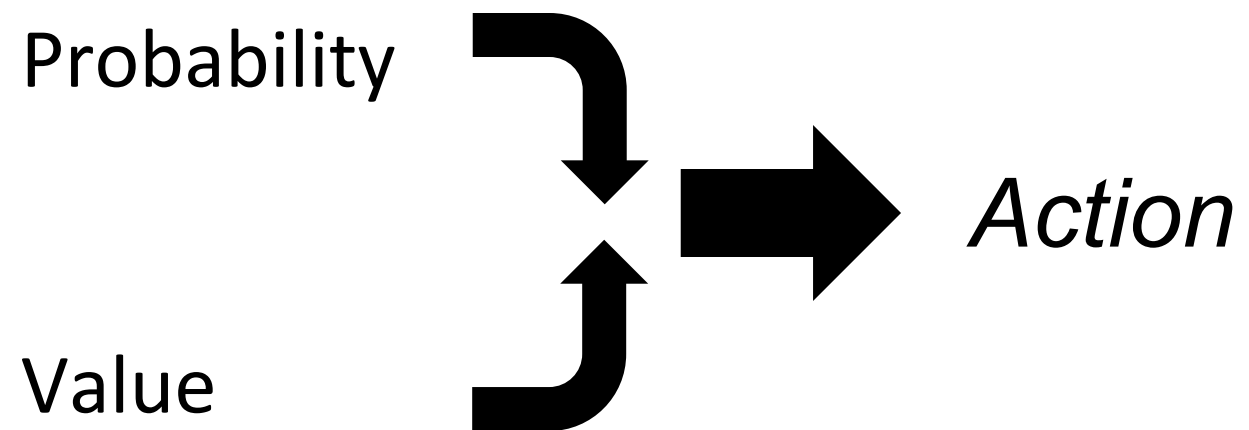
Every life is a series of decisions ....



# Anatomy of a Decision



# Stochastic Form





# Statistical Decision Theory



*Abraham Wald*



*John von Neumann*



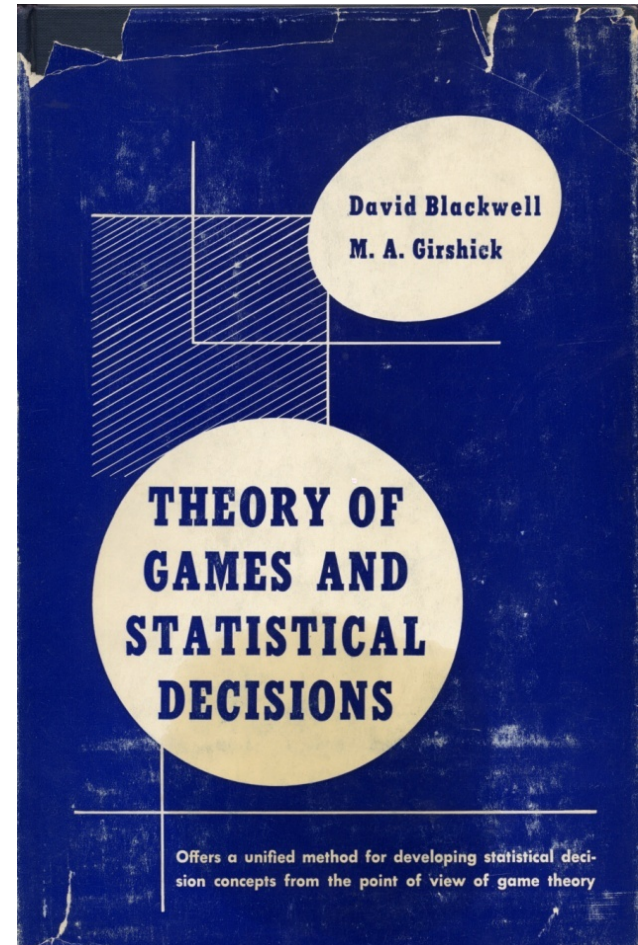
*David Blackwell*



*Oskar Morgenstern*



*M. A. Girschick*



1954

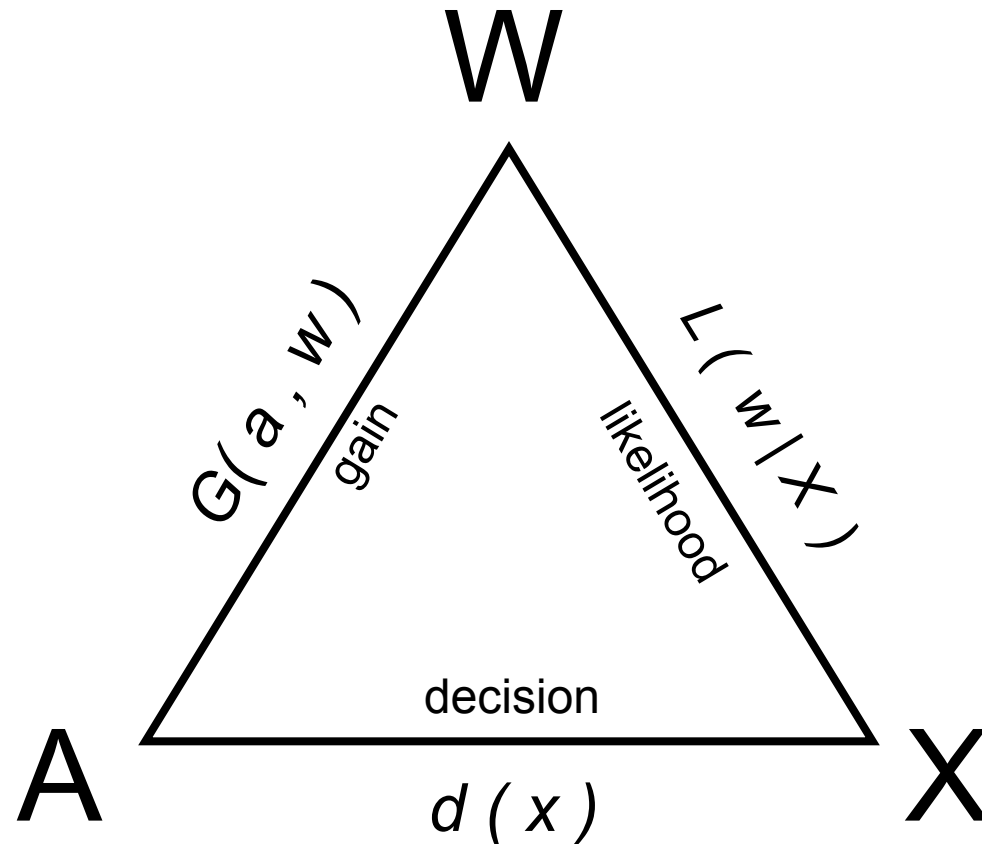
# *Three Elements of SDT*

$W = \{w_1, w_2, \dots, w_m\}$       *possible states of the world*

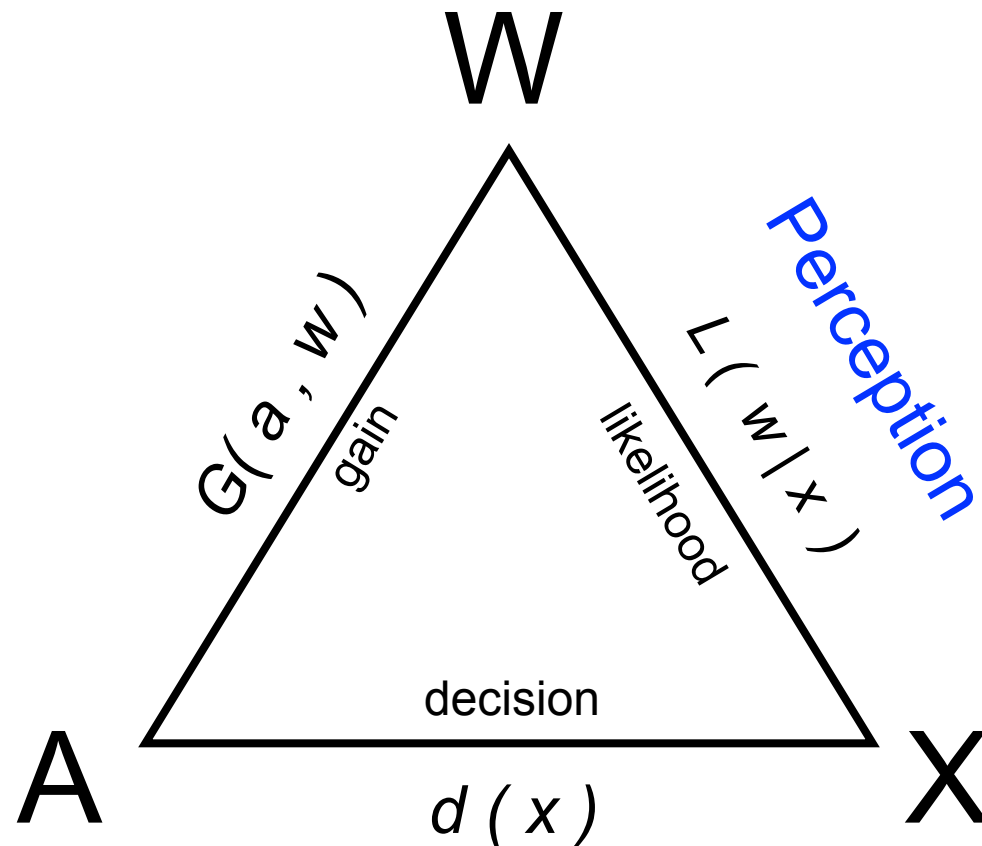
$A = \{a_1, a_2, \dots, a_p\}$       *possible actions*

$X = \{x_1, x_2, \dots, x_n\}$       *possible sensory events*

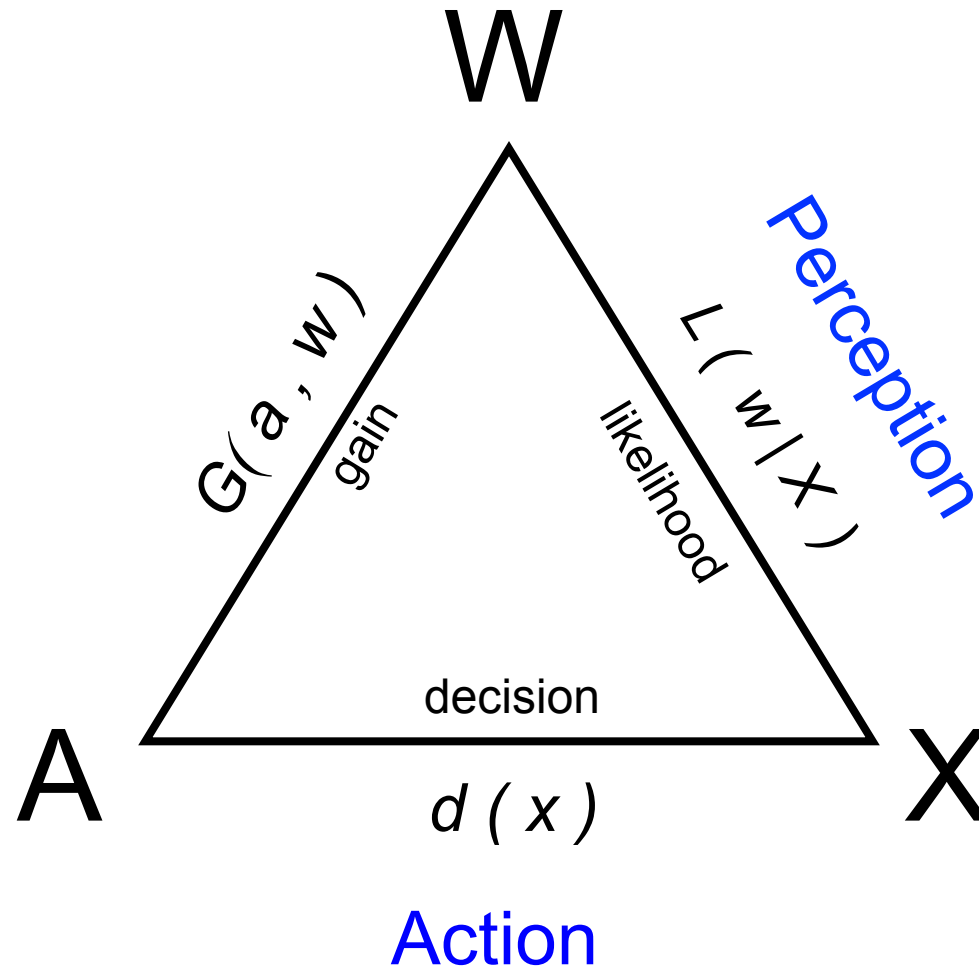
# *Three Functions of SDT*



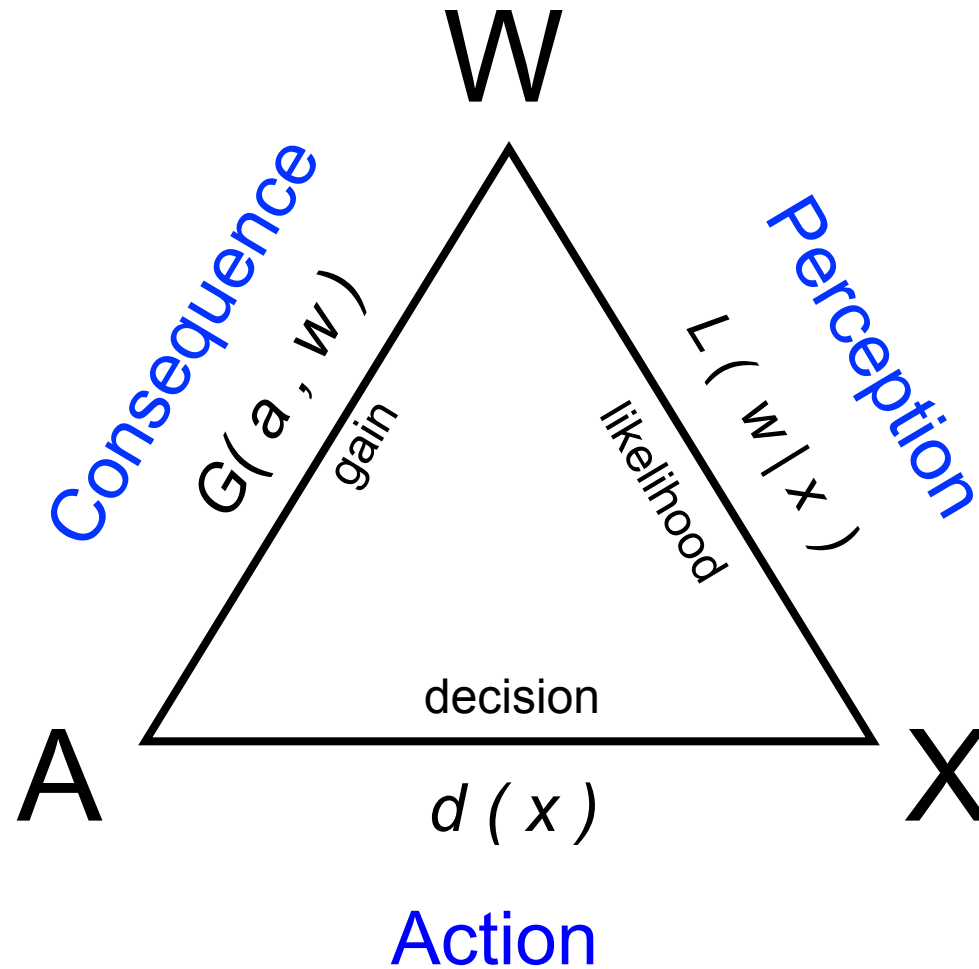
# Three Functions of SDT



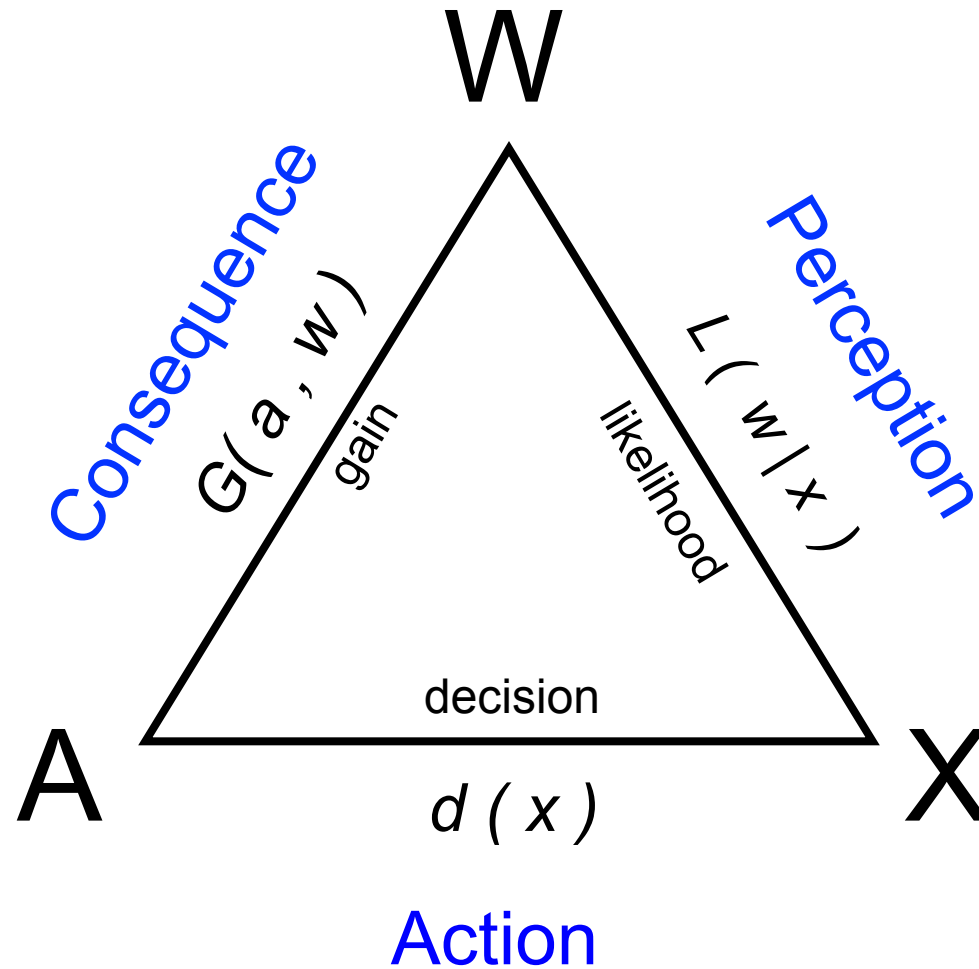
# Three Functions of SDT



# *Statistical Decision Theory*

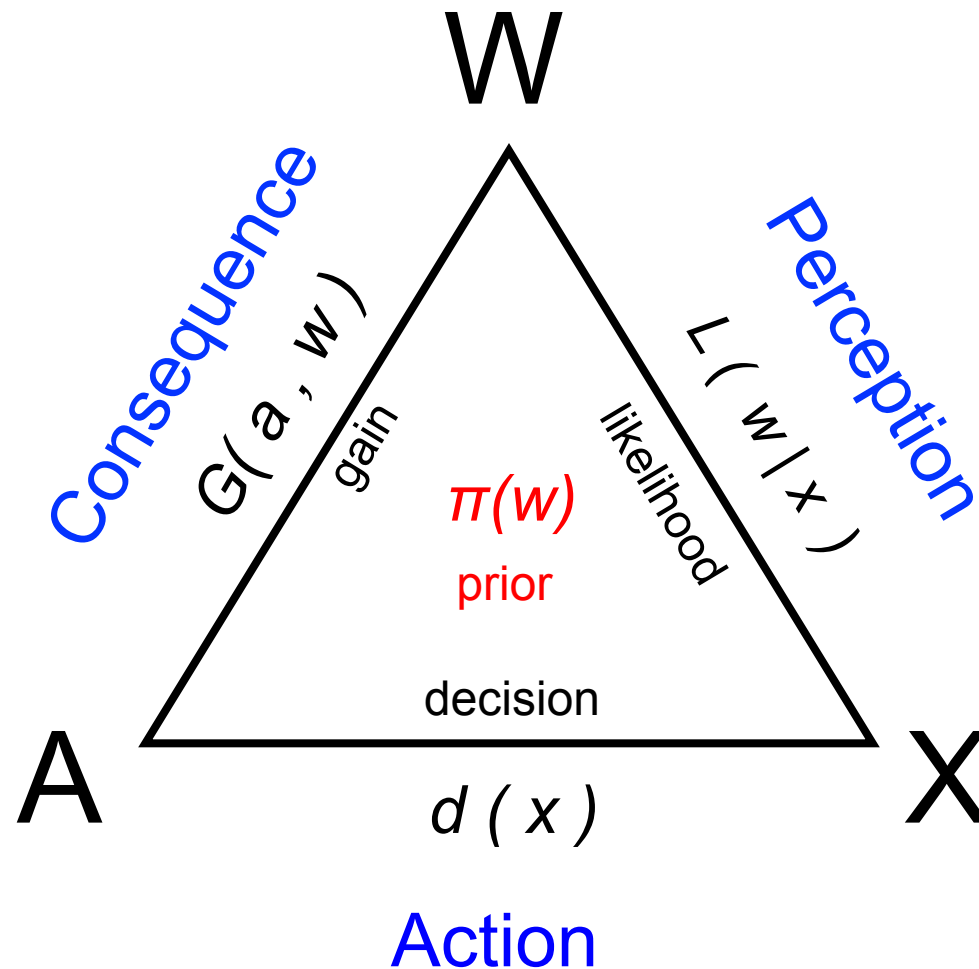


# *Statistical Decision Theory*





# Bayesian Decision Theory



# *A Bayesian Problem*

$G(a, w)$

*Gain*

$\pi(w)$

*Prior*

$L(w | x)$

*Likelihood*

$d : X \rightarrow A$

## **Linking hypothesis**

Maximize expected gain by  
choice of  $d(\cdot)$

Despite changes in gain function, likelihood, prior.

# A Bayesian Game

:

speeded reaching

# A Bayesian Problem

Trommershäuser *et al.*

Vol. 20, No. 7/July 2003/J. Opt. Soc. Am. A 1419

## Statistical decision theory and the selection of rapid, goal-directed movements

Julia Trommershäuser, Laurence T. Maloney, and Michael S. Landy

*Department of Psychology and Center for Neural Science, New York University, New York, New York 10003*

Received October 1, 2002; revised manuscript received January 30, 2003; accepted February 3, 2003

We present two experiments that test the range of applicability of a movement planning model (MEGaMove) based on statistical decision theory. Subjects attempted to earn money by rapidly touching a green target region on a computer screen while avoiding nearby red penalty regions. In two experiments we varied the magnitudes of penalties, the degree of overlap of target and penalty regions, and the number of penalty regions. Overall, subjects acted so as to maximize gain in a wide variety of stimulus configurations, in good agreement with predictions of the model. © 2003 Optical Society of America

*OCIS codes:* 330.4060, 330.7310.

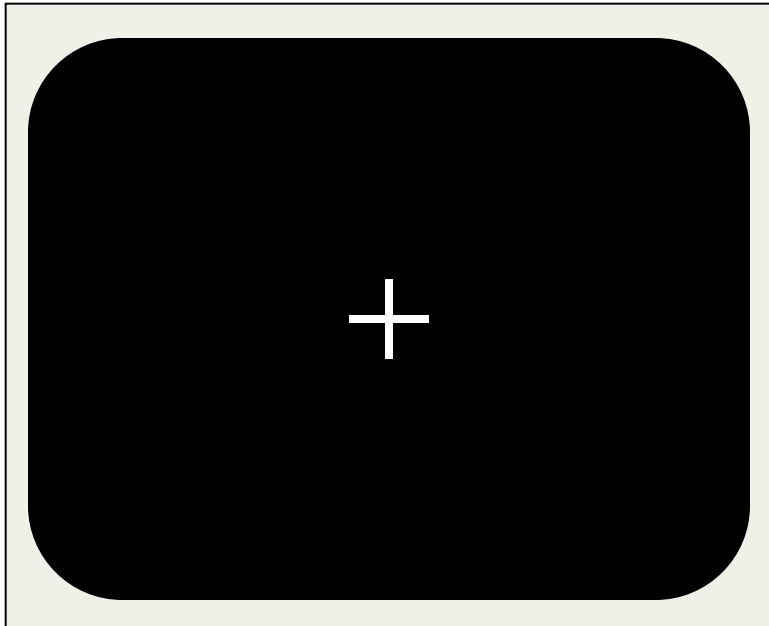
### 1. INTRODUCTION

Motor responses have consequences. In the first semifinal of the 2002 World Cup, Germany met host South Korea, and the match was decided in the 75th minute by the only goal of the match. Oliver Neuville passed the ball to Germany's play maker Michael Ballack, who scored in his

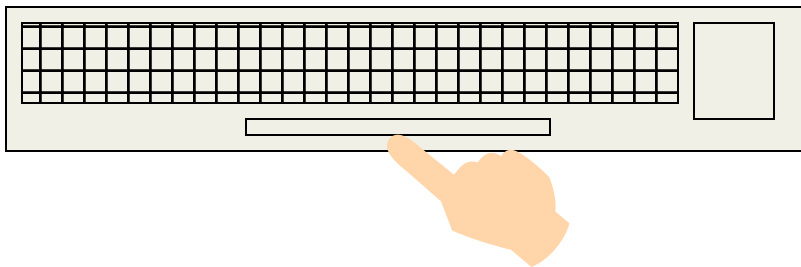
small target region, the subject cannot be certain that a movement aimed at the center of the reward region will not, instead, end up outside the reward region, possibly in the penalty region. To summarize, the subject, in each trial, must select among possible actions and must do so very rapidly. There are explicit monetary penalties associated with the outcome of the action selected, but the



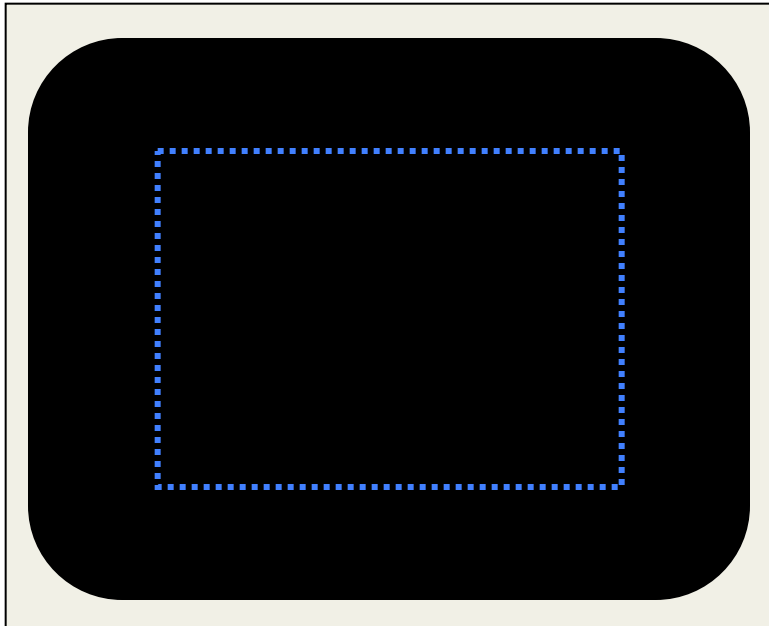
# Experimental Task



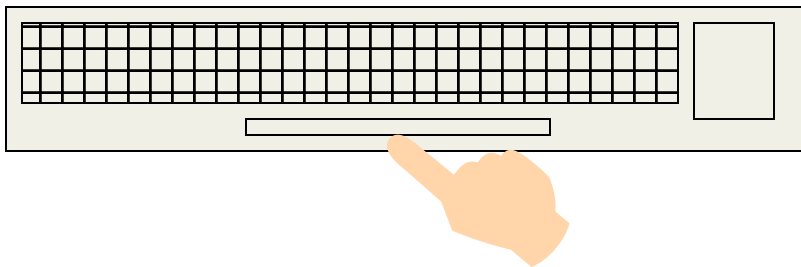
Start of trial:  
display of fixation  
cross (1.5 s)



# Experimental Task

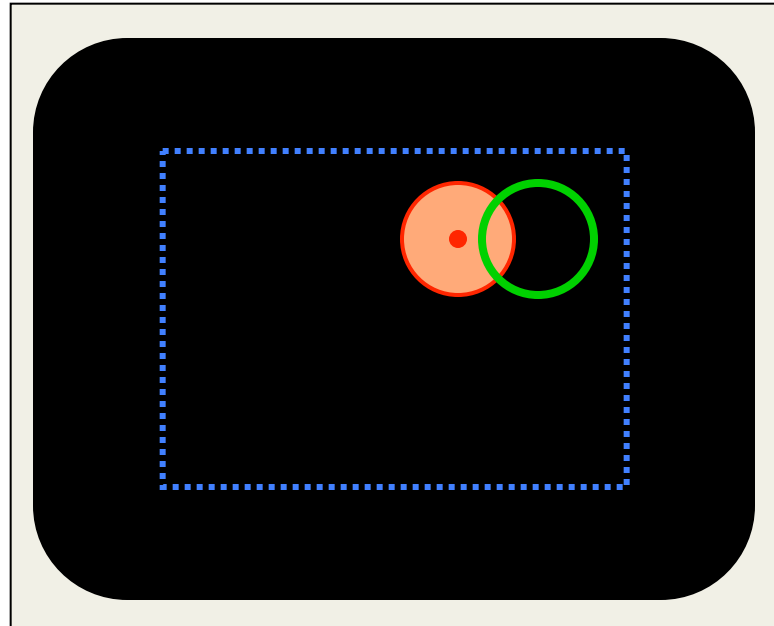


Display of response area,  
500 ms before  
target onset  
(114.2 mm x 80.6 mm)

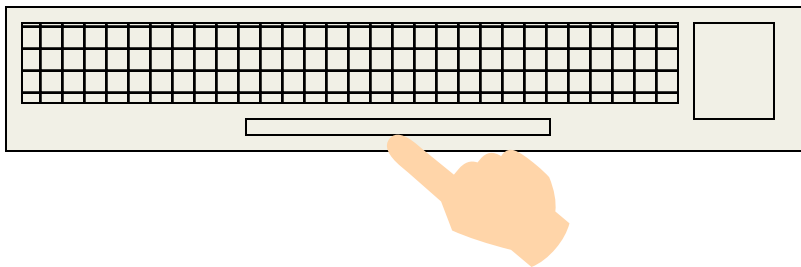




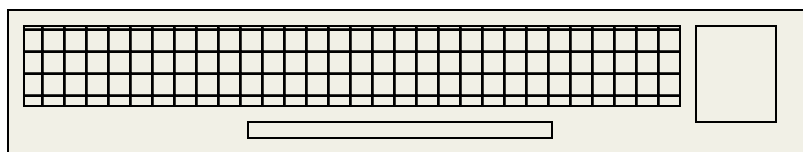
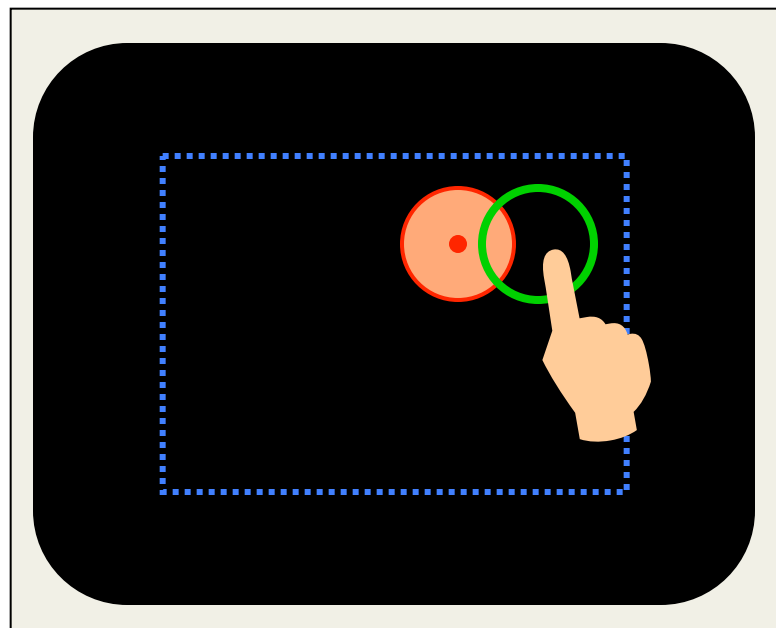
# Experimental Task



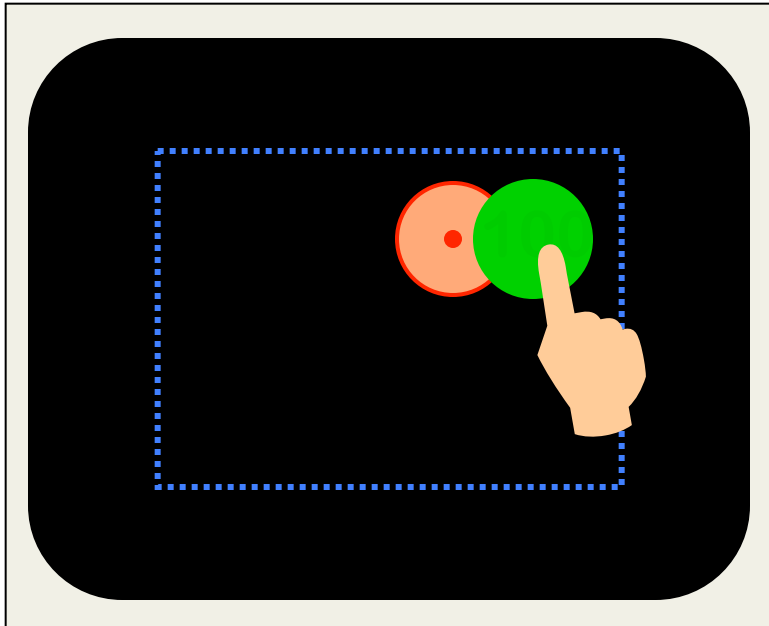
Target display (700 ms)



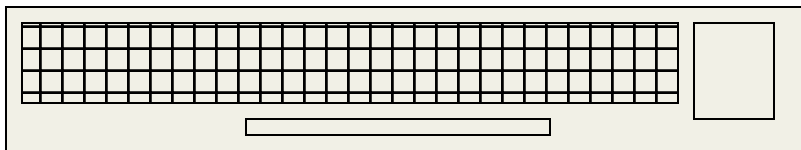
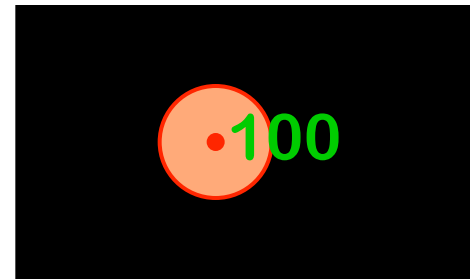
# Experimental Task



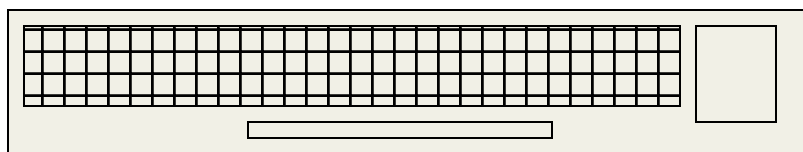
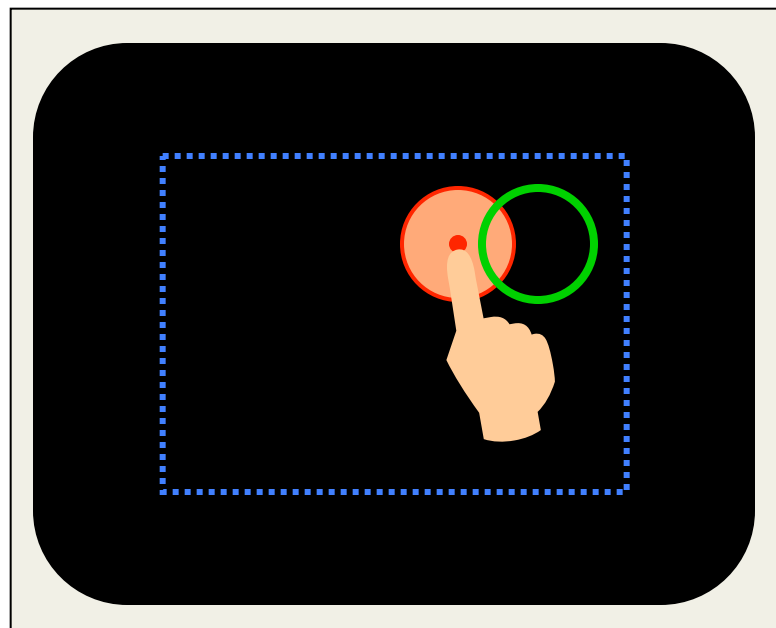
# Experimental Task



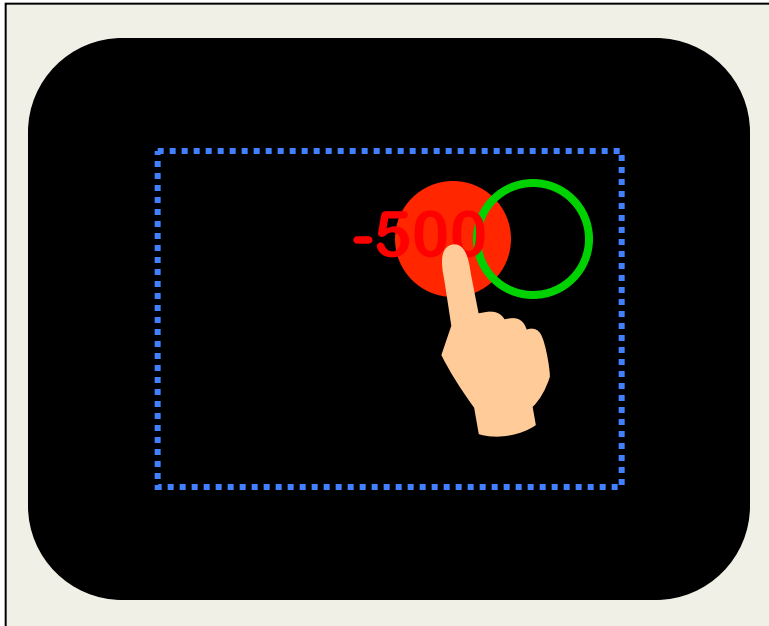
The green target is hit:  
**+100** points



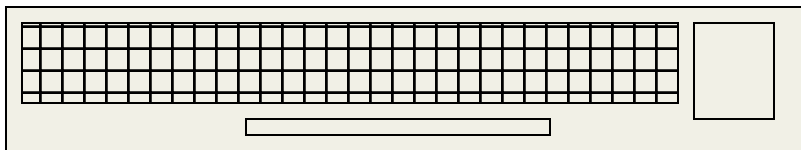
# Experimental Task



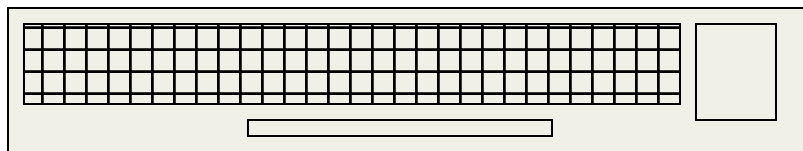
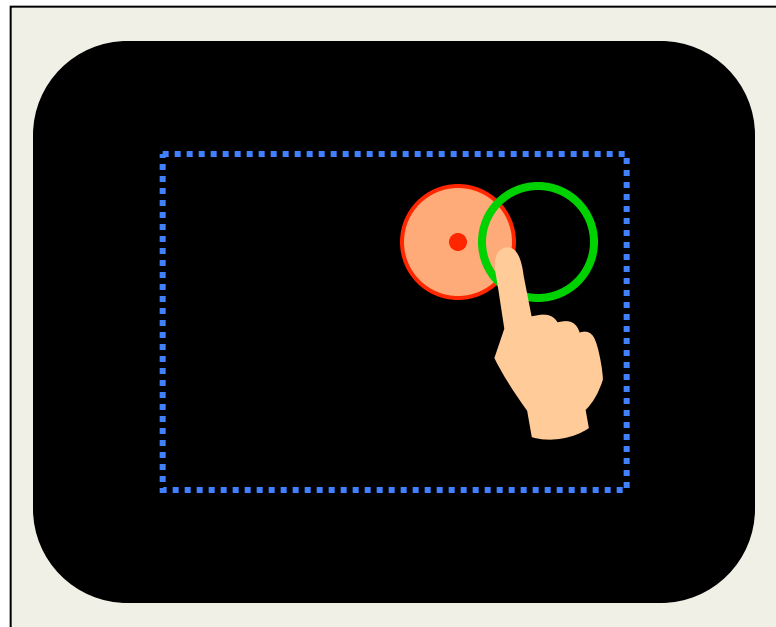
# Experimental Task



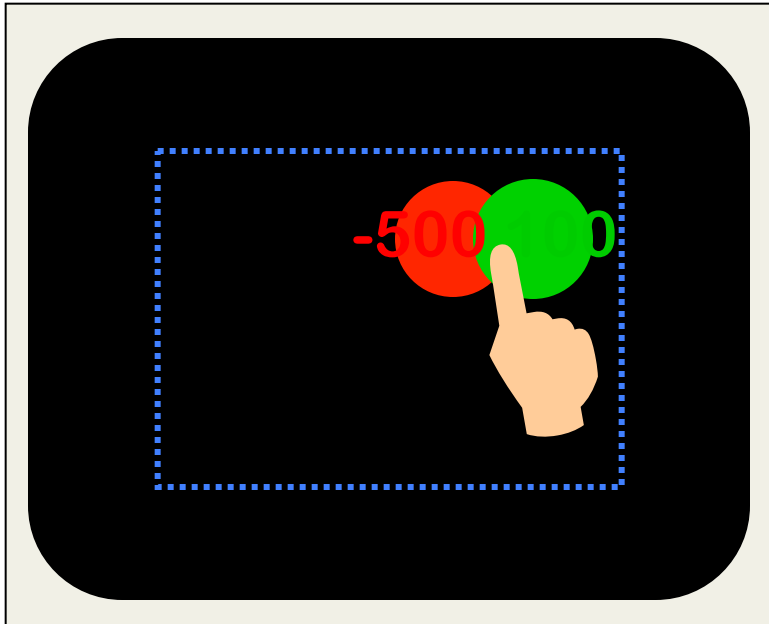
The red target is hit:  
**-500** points



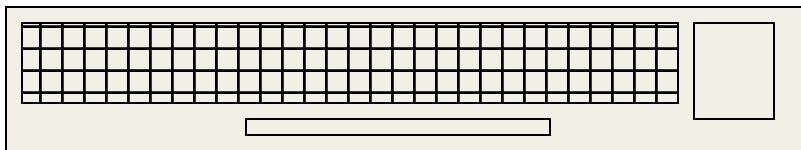
# Experimental Task



# Experimental Task

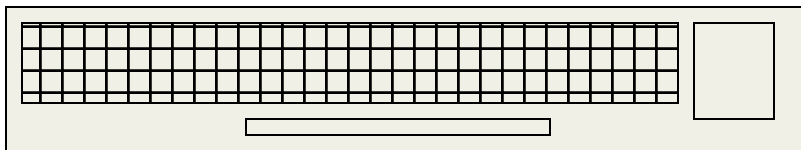
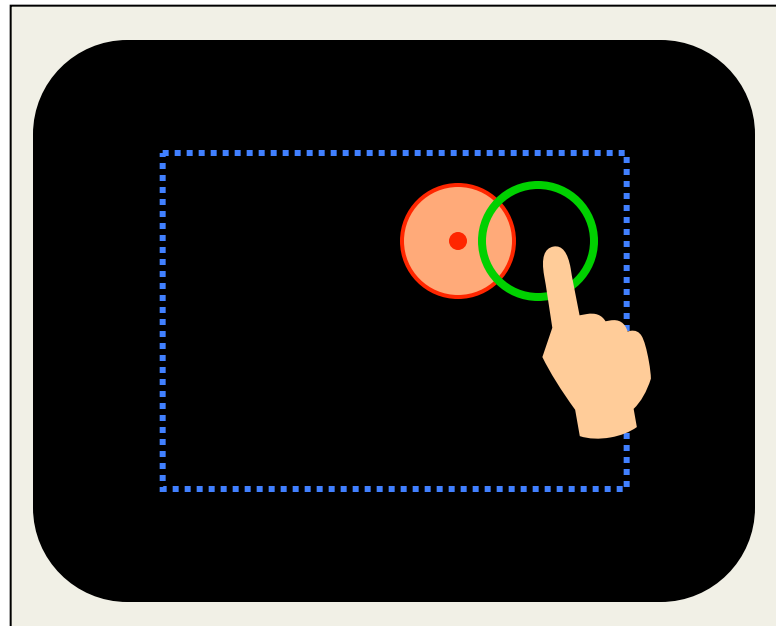


Scores add if both  
targets are hit:





# Experimental Task

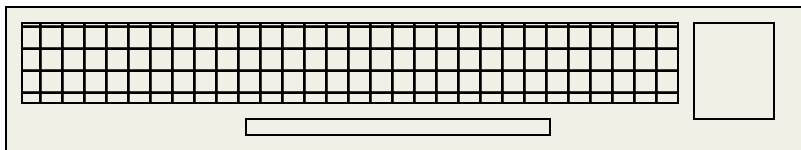


# Experimental Task

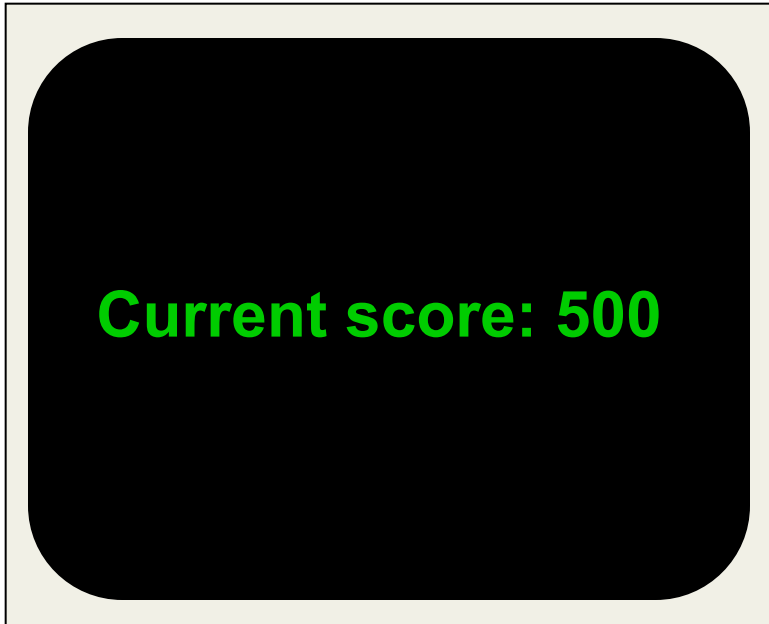


The screen is hit  
later than 700 ms  
after target display:  
-700 points.

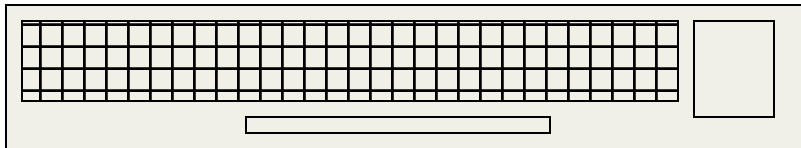
If you are on time but  
Miss the targets, 0.



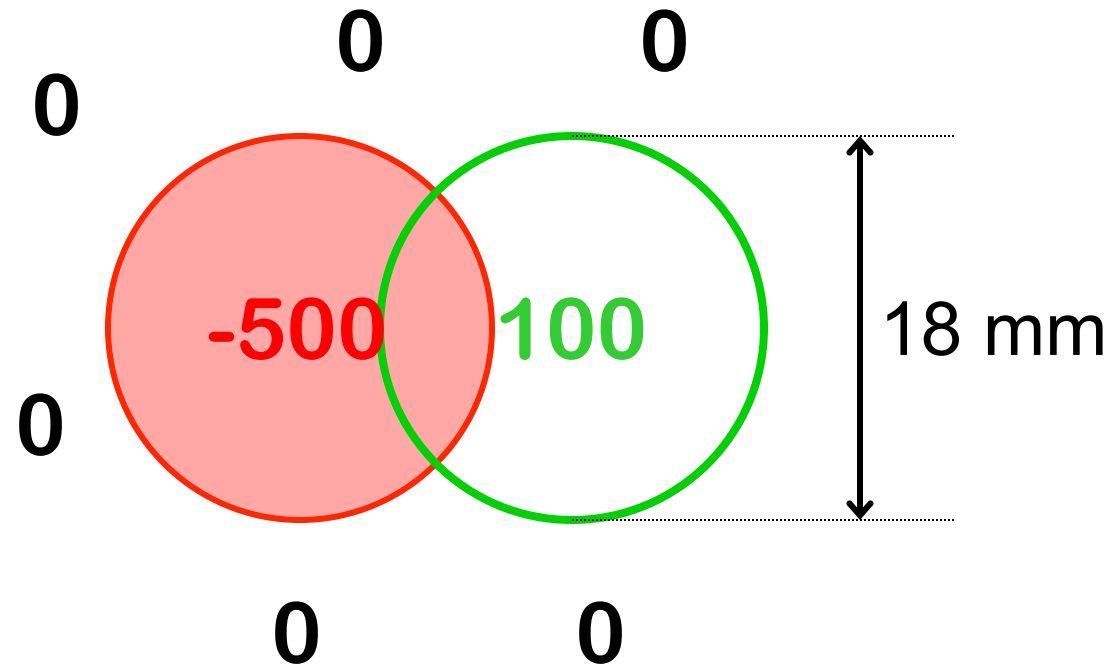
# Experimental Task



End of trial

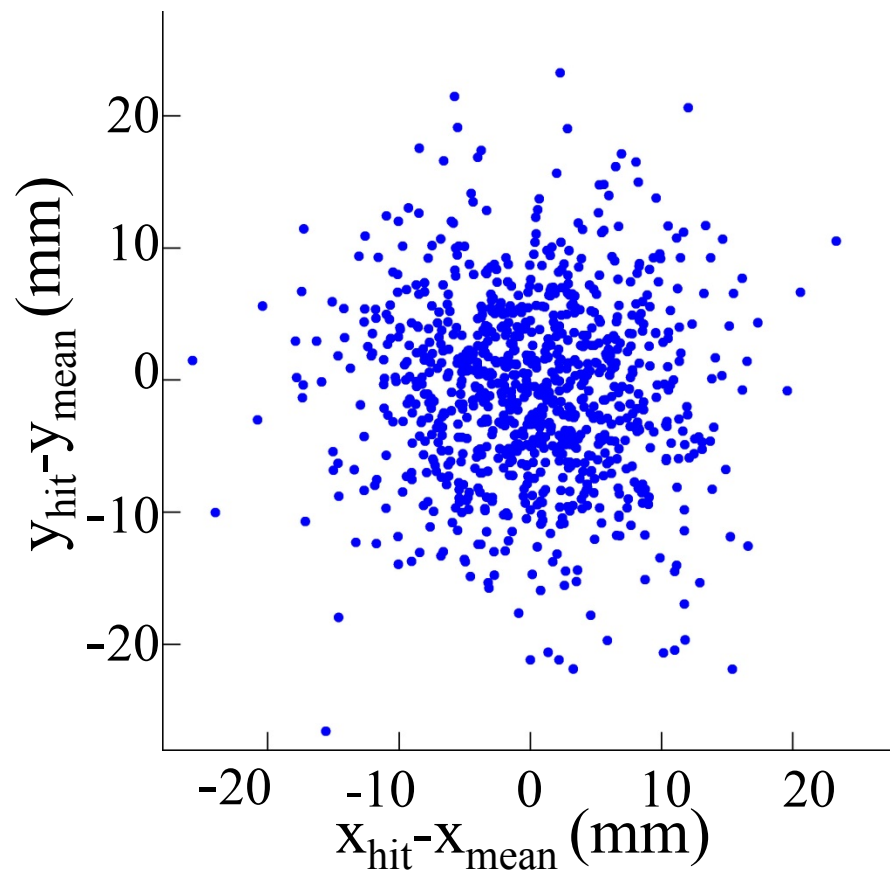


# Choice among Movement Strategies



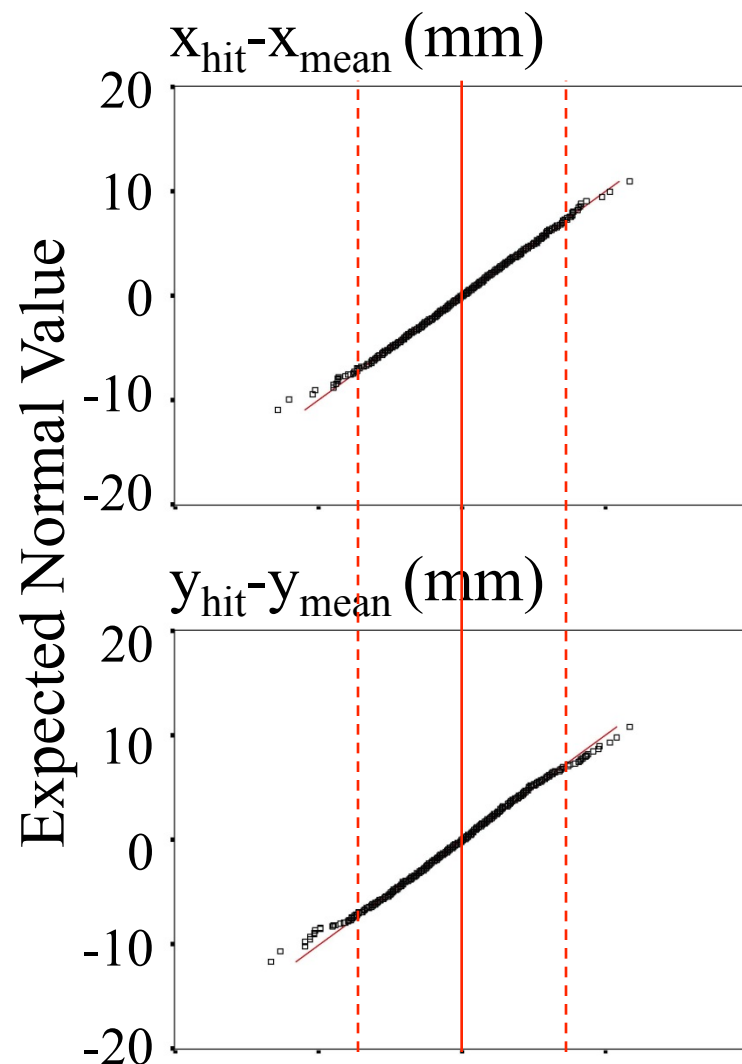
*What should Paulina do?*

## Distribution of movement end points



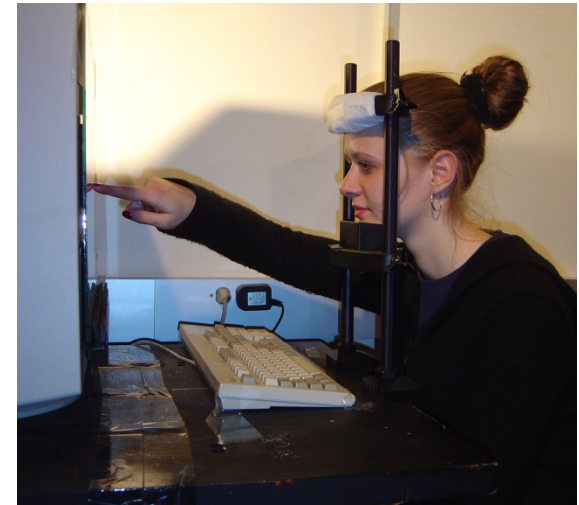
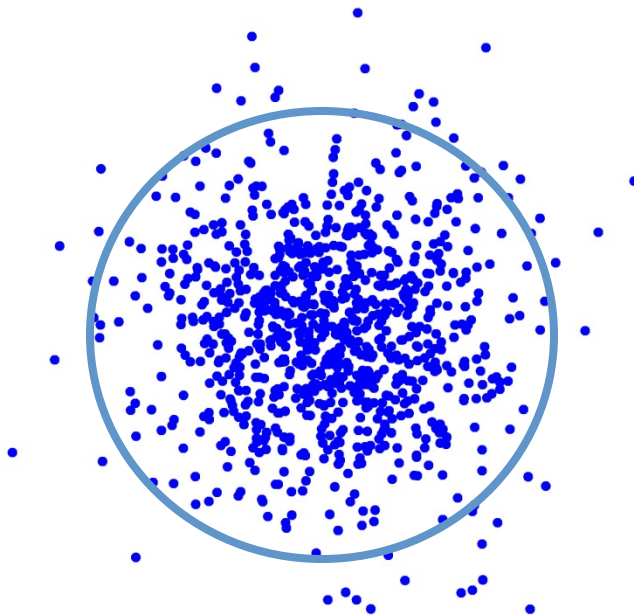
Subject S4,  $\sigma = 3.62$  mm,  
72x15 = 1080 end points

## Q-Q Plot



Observed Value

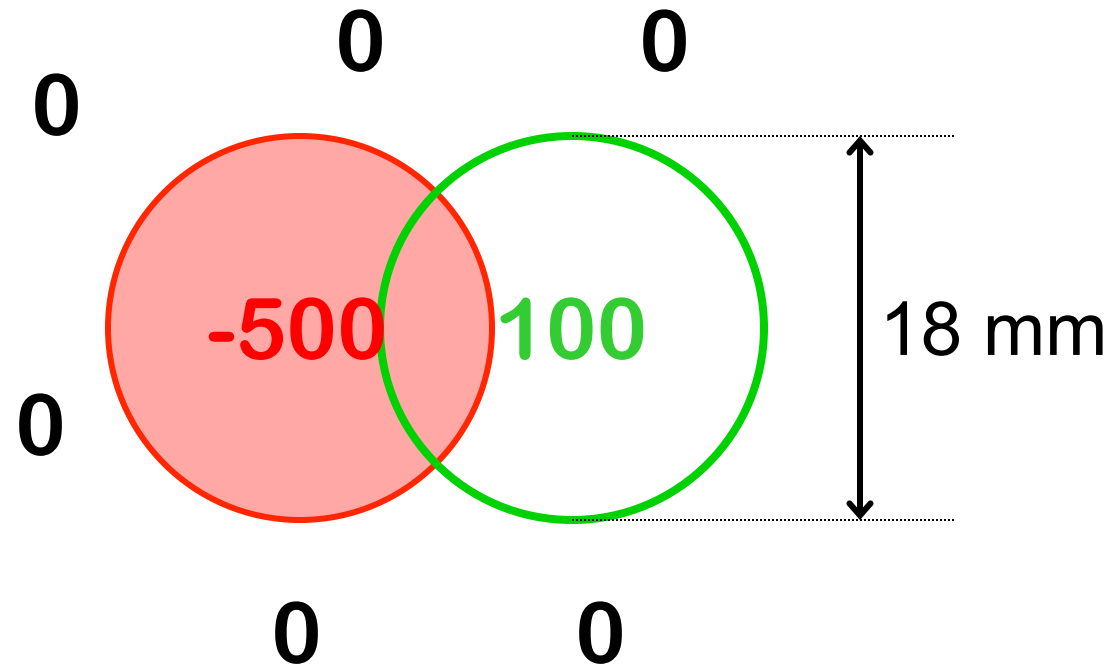
*If there were no red penalty circle ....*



*Aim for center  
Select perceptual-motor strategies that  
**minimize variance***

*Harris & Wolpert (1998)*

# Choice among Movement Strategies

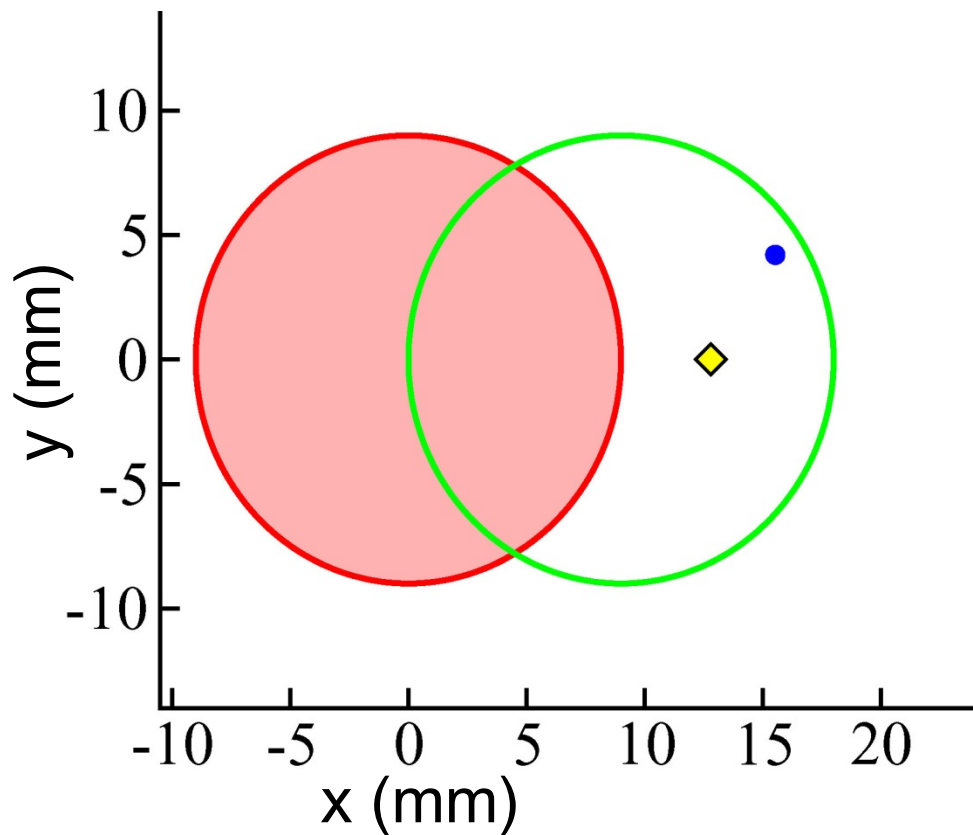


## What should Paulina do?

# Thought Experiment

● : -500

○ : 100 points (2.5 ¢)



100 points



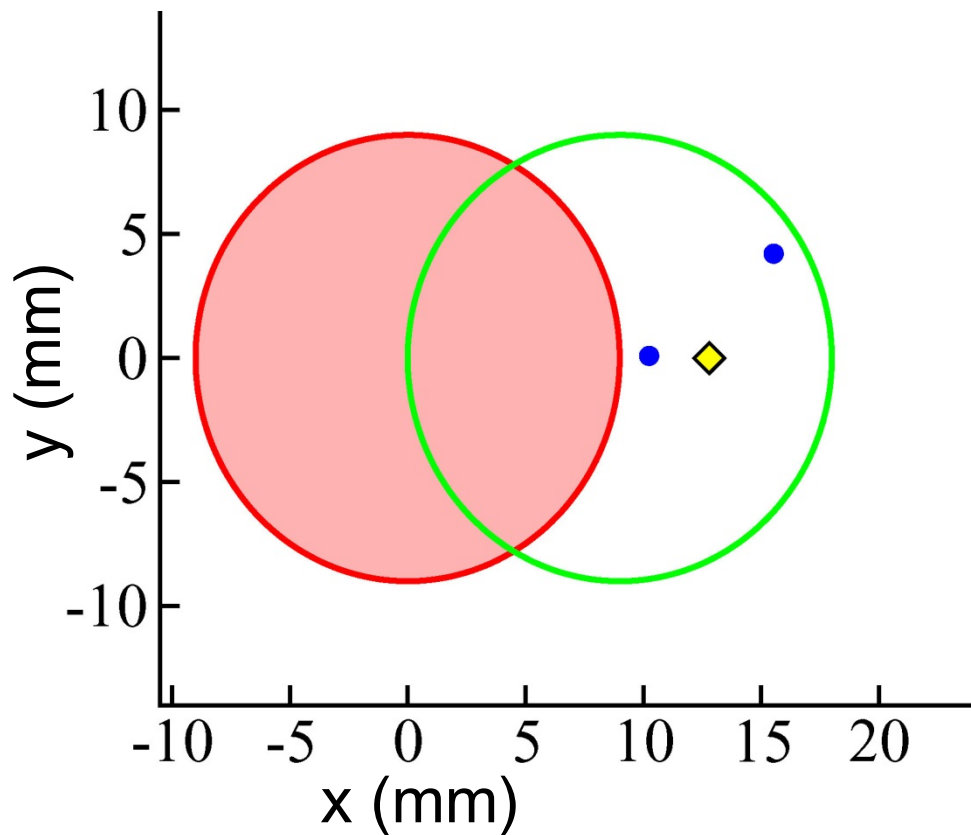
= 4.83 mm



# Thought Experiment

● : -500

○ : 100 points (2.5 ¢)



100 points  
100 points  

---

200 points

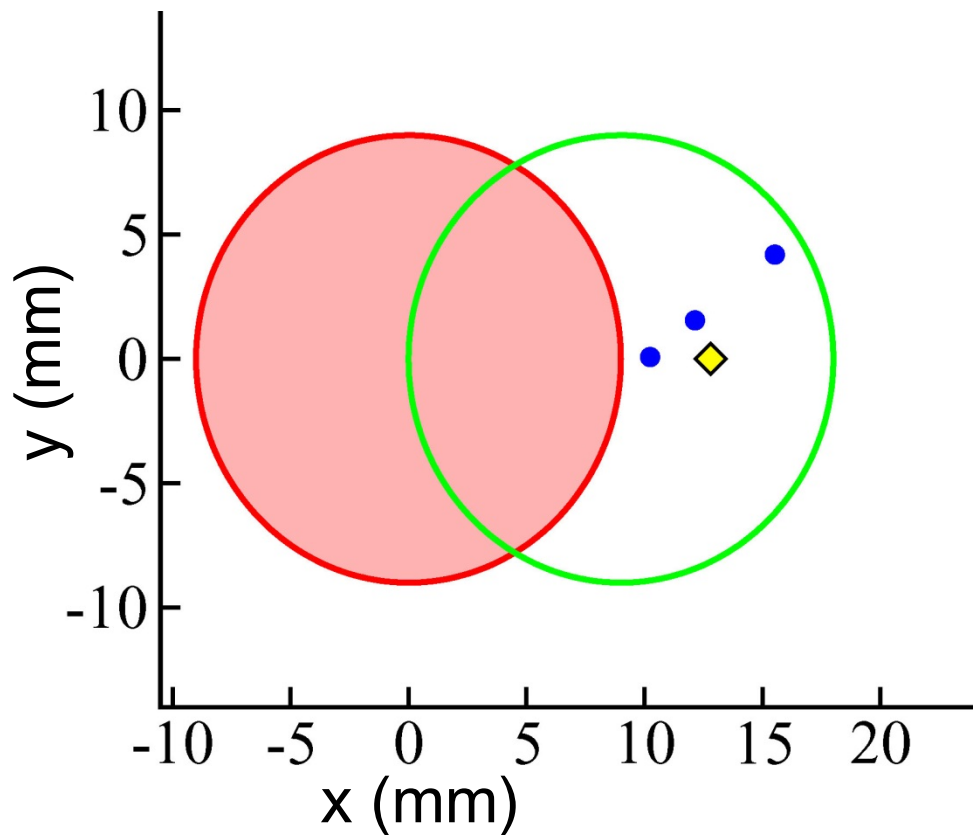


= 4.83 mm

# Thought Experiment

● : -500

○ : 100 points (2.5 ¢)



100 points  
100 points  
100 points  

---

300 points

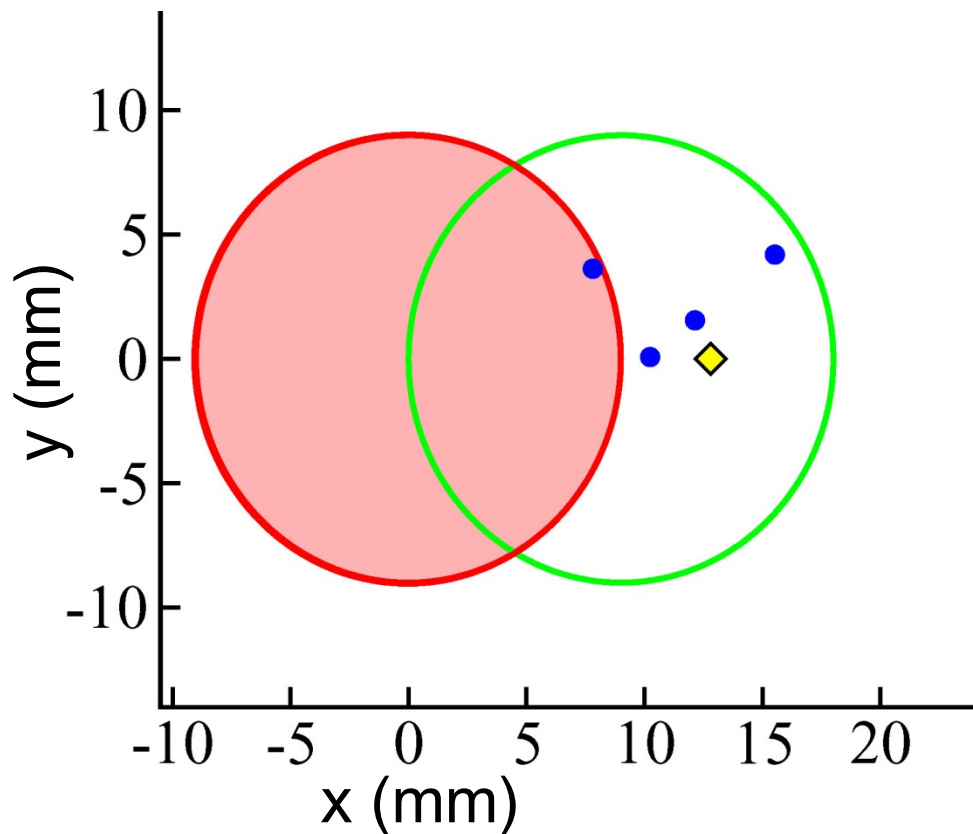


= 4.83 mm

# Thought Experiment

● : -500


○ : 100 points (2.5 ¢)



100 points  
100 points  
100 points  
-400 points  

---

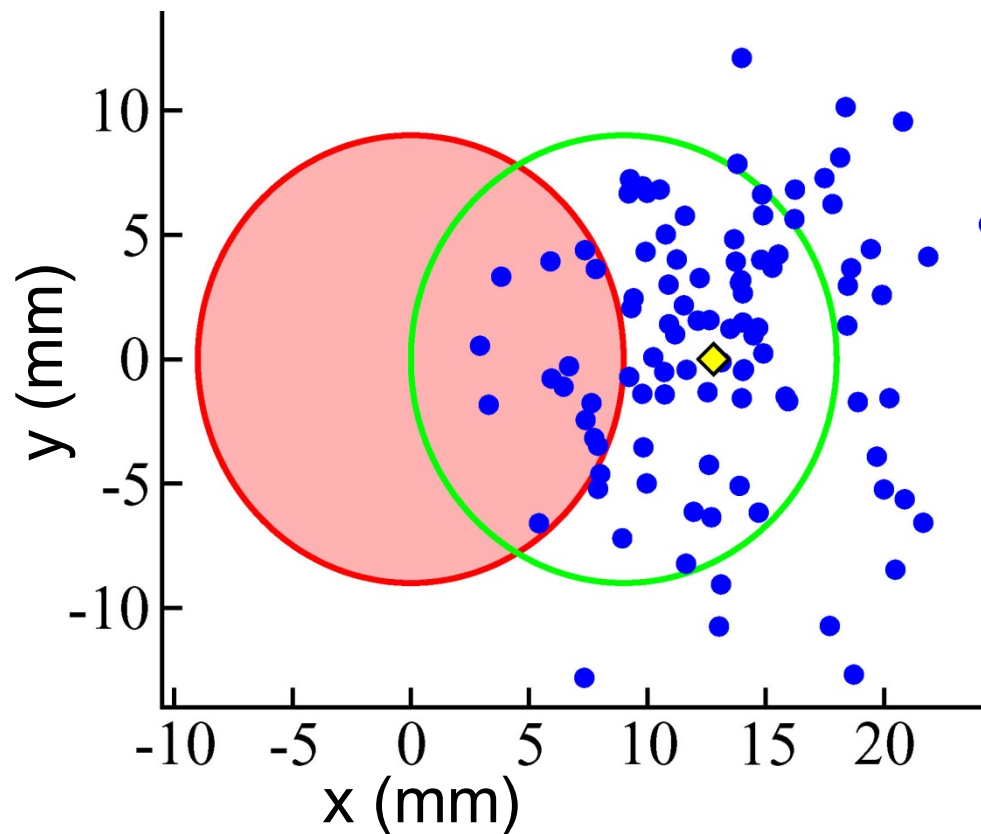
-100 points

 = 4.83 mm

# Thought Experiment


● : -500

○ : 100 points (2.5 ¢)



100 points  
100 points  
100 points  
-400 points  
⋮

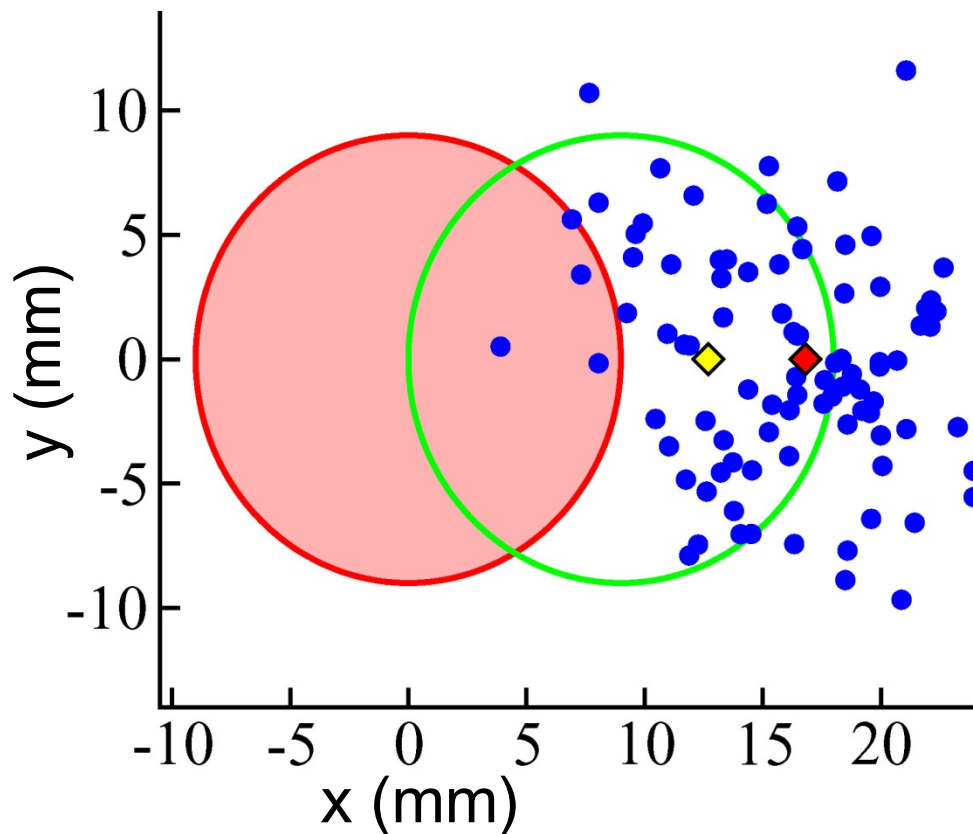
-0.3 pts. per trial

 = 4.83 mm

# Thought Experiment

● : -500

○ : 100 points (2.5 ¢)



- 0.3 pts. per trial

30.7 pts. per trial

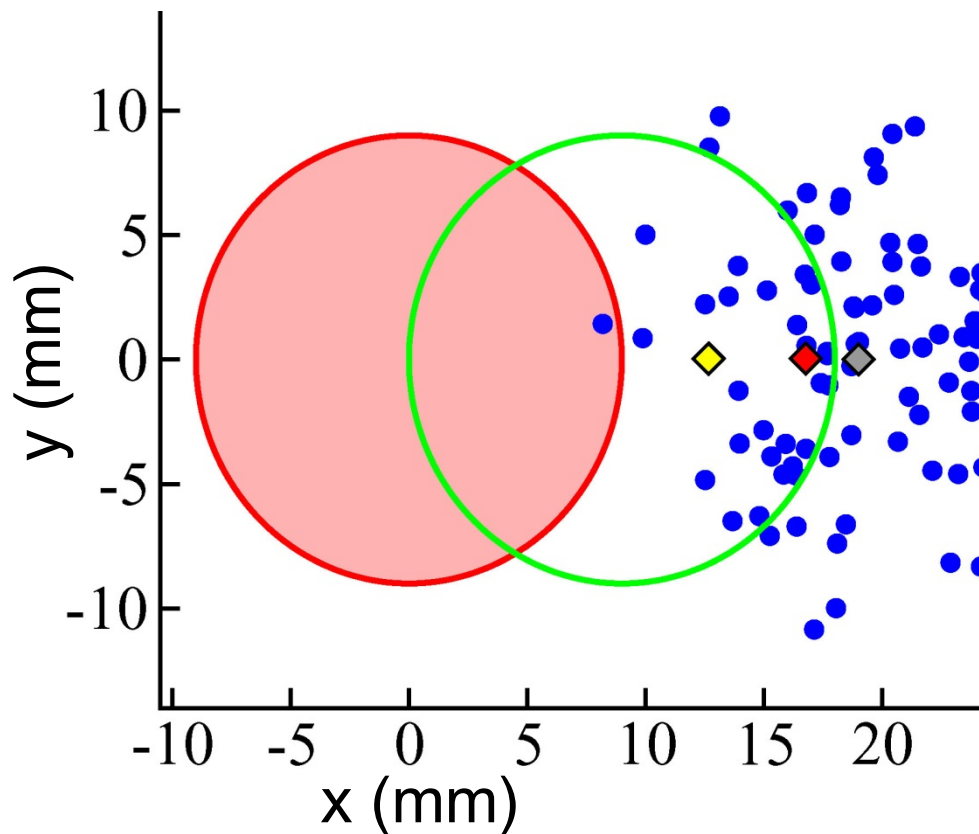


= 4.83 mm

# Thought Experiment

● : -500

○ : 100 points (2.5 ¢)



- 0.3 pts. per trial

30.7 pts. per trial

25.5 pts. per trial

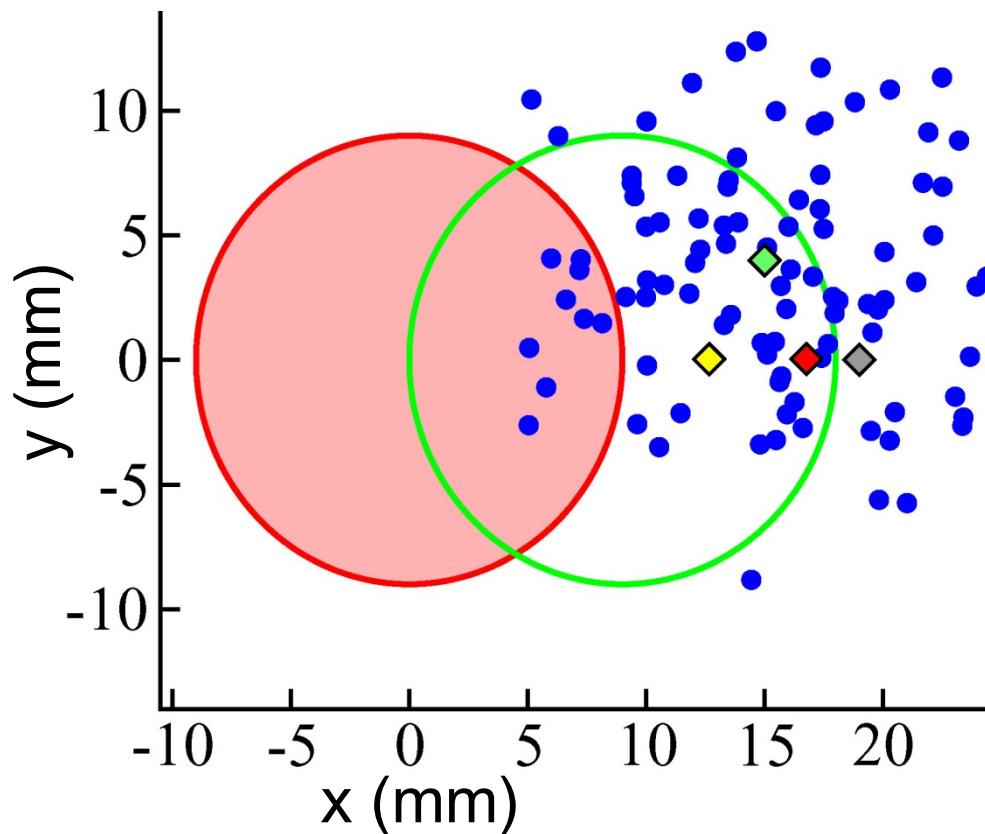


= 4.83 mm

# Thought Experiment

● : -500

○ : 100 points (2.5 ¢)



- 0.3 pts. per trial

30.7 pts. per trial

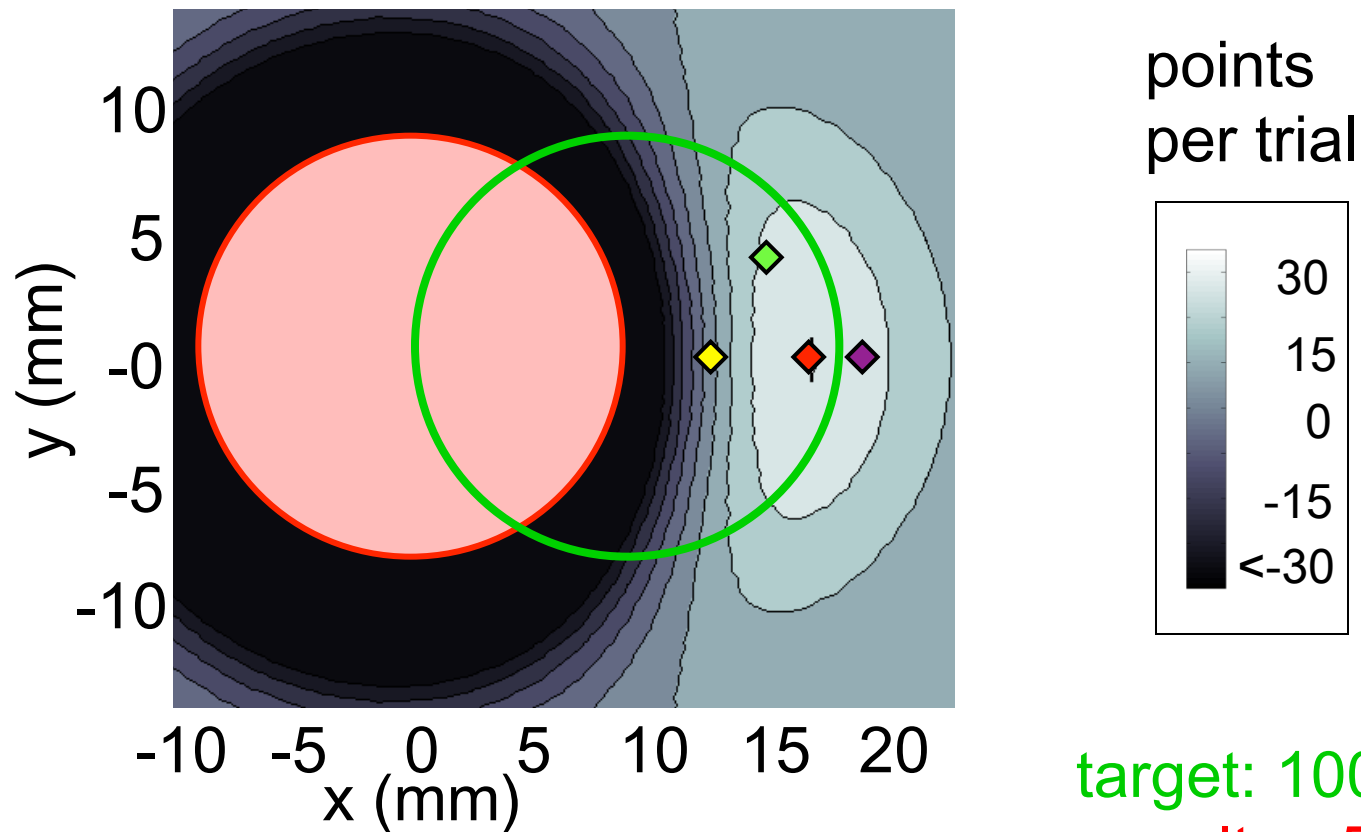
25.5 pts. per trial


22.6 pts. per trial



= 4.83 mm

Expected value as function of  
mean movement end point (x,y):

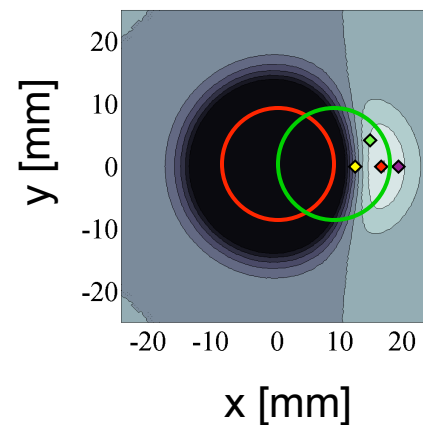
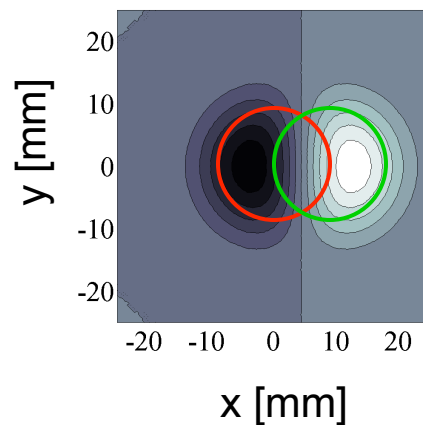
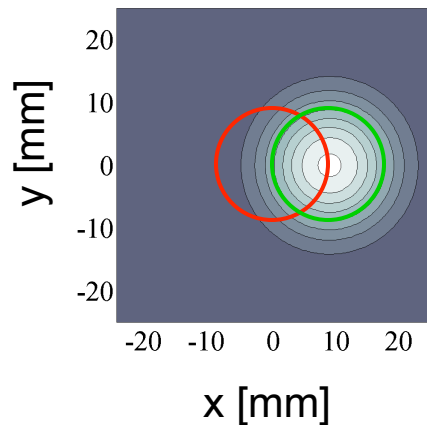
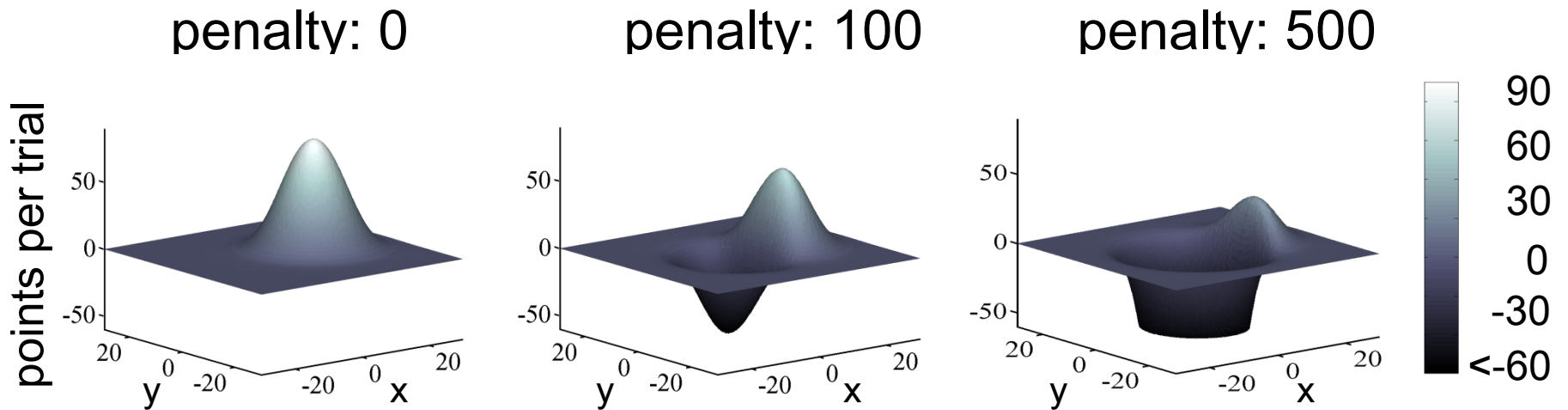


 = 4.83 mm

target: 100  
penalty: -500

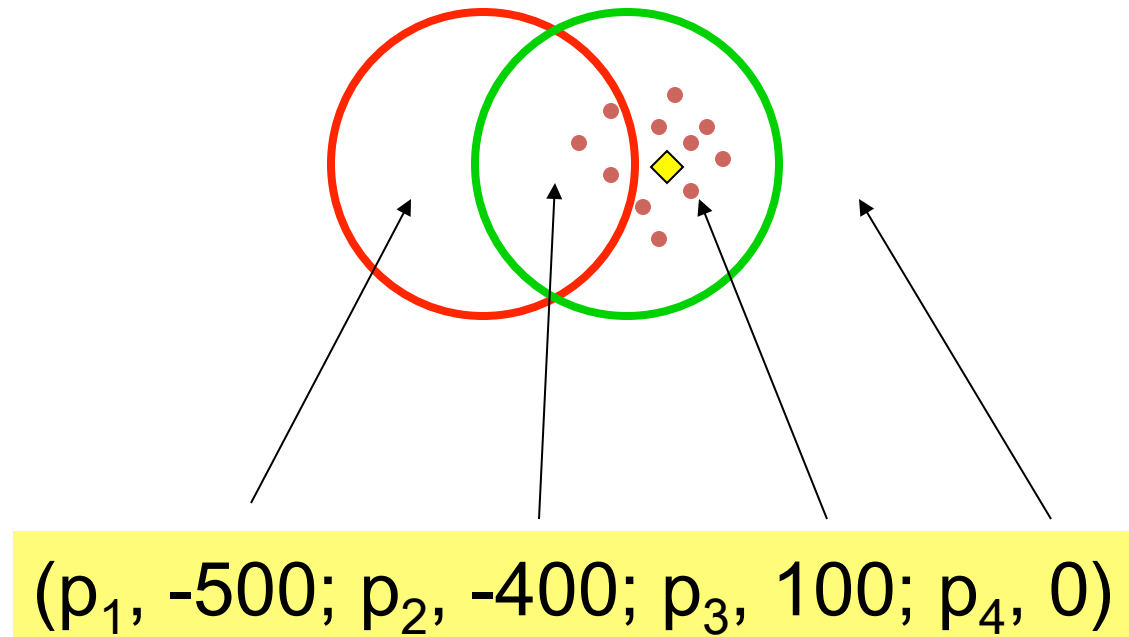


# Thought Experiment

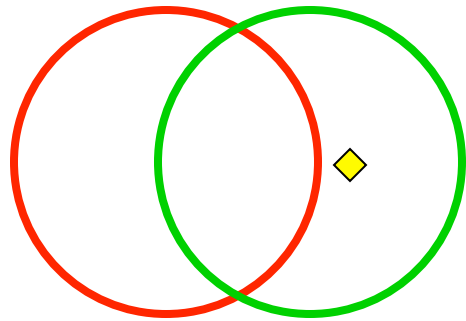



$\sigma = 4.83 \text{ mm}$

# Movement plans as lotteries



# Movement plans as lotteries



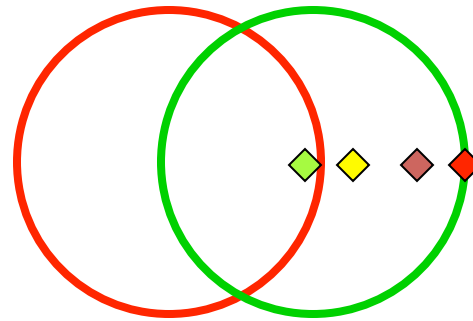
 = 4 mm


Lottery:

(1.3%, -500; 30.3%, -400; 60.9%, 100; 7.5%, 0)

# Movement plans as lotteries

Optimal aim point: lottery with MEV



 = 4 mm

(6.6%, -500; 52.3%, -400; 37.0%, 100; 4.0%, 0)

(1.3%, -500; 30.3%, -400; 60.9%, 100; 7.5%, 0)

( 0%, -500; 4.6%, -400; 62.6%, 100; 32.8%, 0)

( 0%, -500; 0.7%, -400; 37.6%, 100; 61.7%, 0)

# Experiment 1

Reaching with Asymmetric Gain/Loss

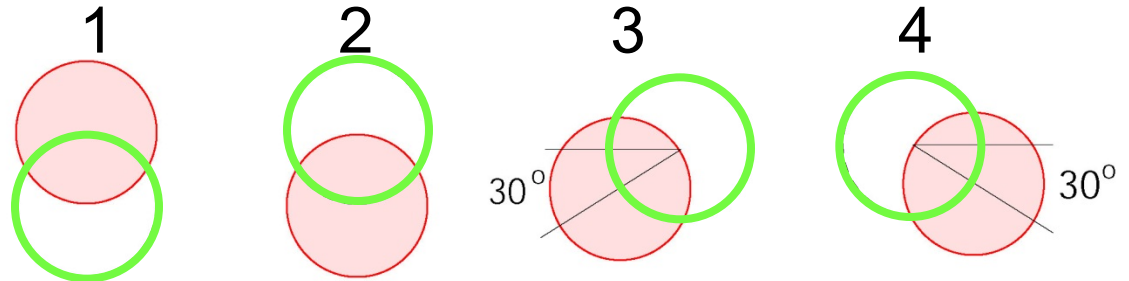


*Julia Trommershäuser*

*Trommershäuser, Maloney, Landy (2003) JOSA A*

# Test of the model: Experiment 1

4 stimulus  
configurations:  
(varied within block)



R = 9 mm

2 penalty conditions:

0 and -500 points (varied between blocks)

5 “practiced movers”

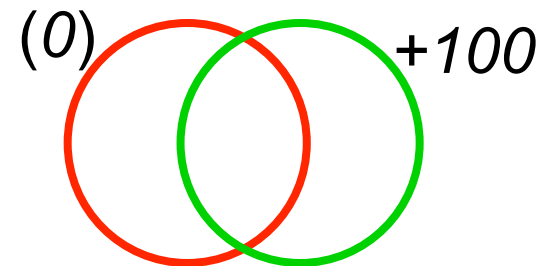
1 session of data collection: 360 trials

24 data points per condition

# General Methods: Training

*For all experiments:*

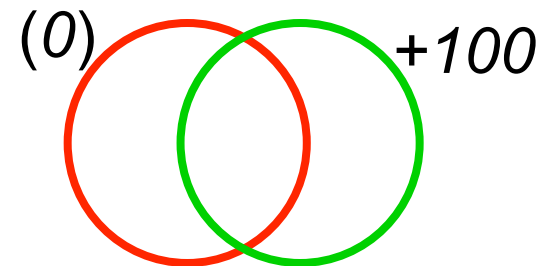
- *All subjects practice the task for 360 trials or more until their variance stabilizes.*
- *The timeout limit is gradually decreased to 700 ms during training.*
- *There are no **penalties** during training (the concept is never mentioned).*
- *We verify that each subject's movement variance has stabilized.*
- *They are told only to **make money**.*



# General Methods: Training

*For all experiments:*

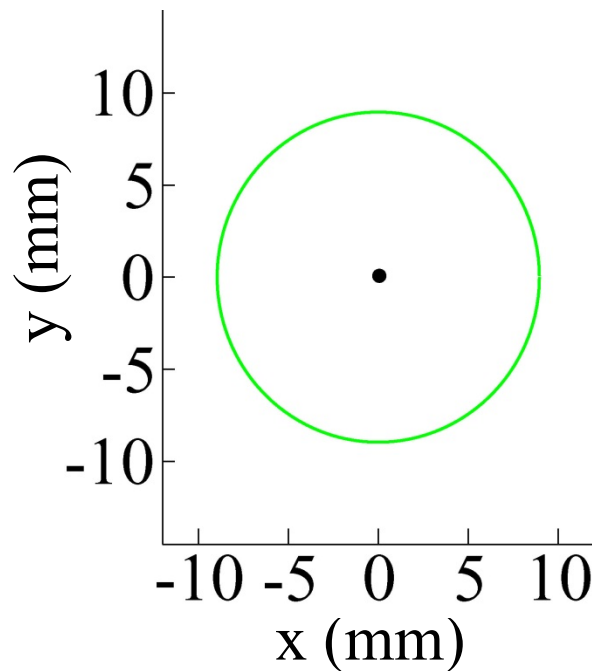
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# Results: Experiment 1

Model prediction:

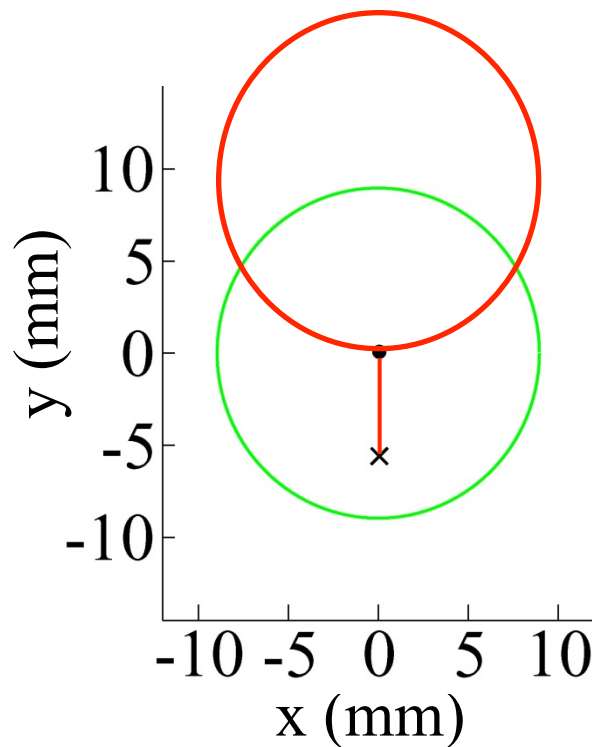


- model,  $\text{penalty} = 0$

Subject S5,  $\sigma = 2.99$  mm

# Results: Experiment 1

Model prediction: configuration 1

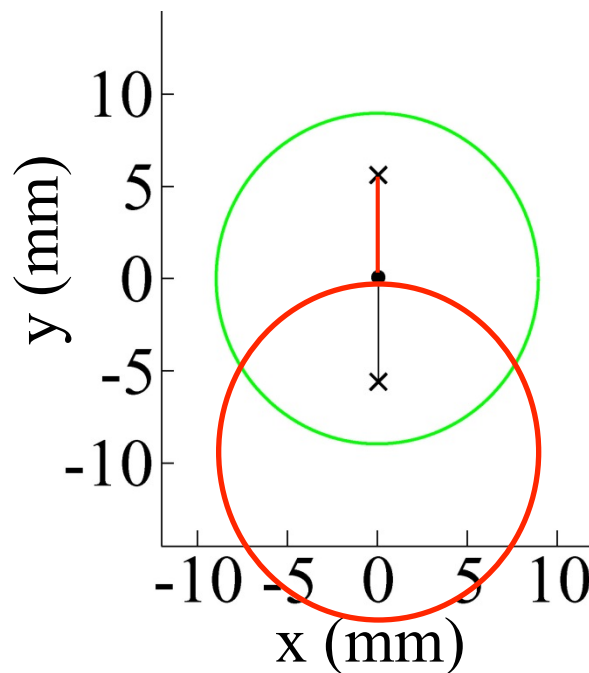


- model, **penalty = 0**
- × model, **penalty = 500**

Subject S5,  $\sigma_y = 2.99$  mm

# Results: Experiment 1

Model prediction: configuration 2

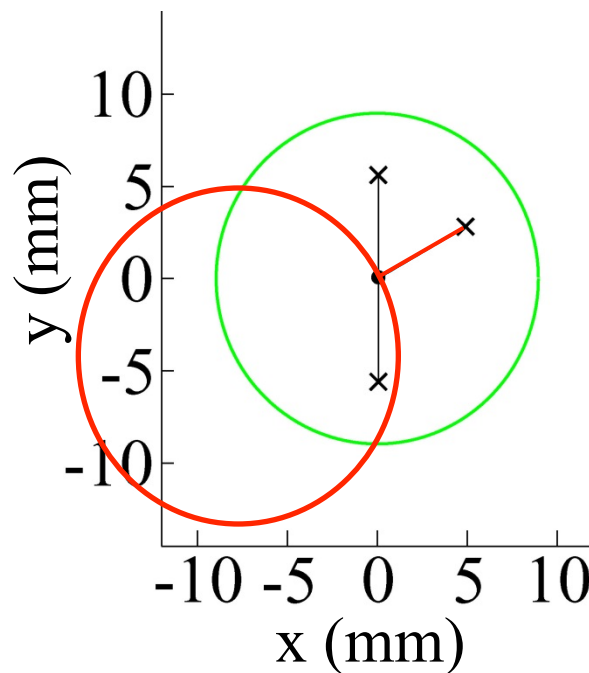


- model, **penalty = 0**
- × model, **penalty = 500**

Subject S5,  $\sigma = 2.99$  mm

# Results: Experiment 1

Model prediction: configuration 3

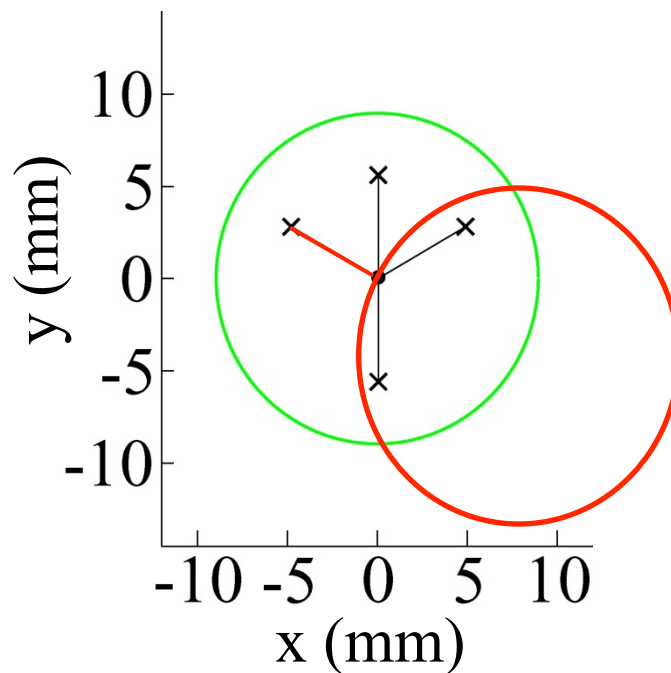


- model, **penalty = 0**
- × model, **penalty = 500**

Subject S5,  $\sigma_y = 2.99$  mm

# Results: Experiment 1

Model prediction: configuration 4

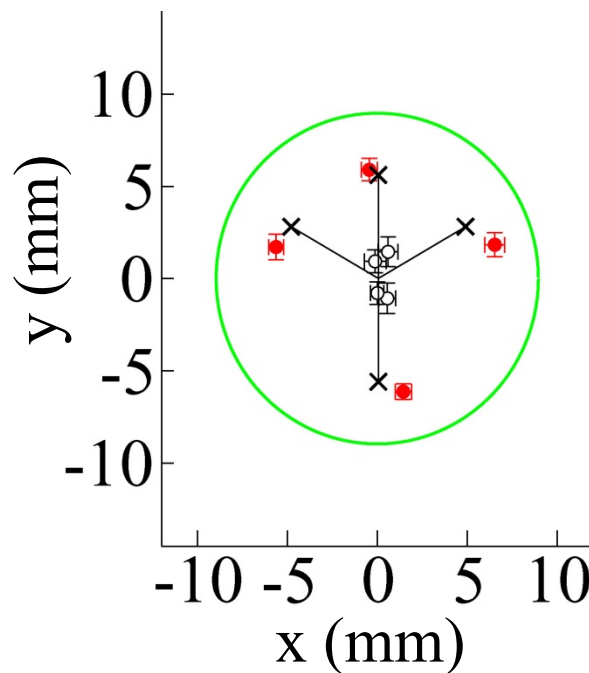


- model, **penalty = 0**
- x model, **penalty = 500**

Subject S5,  $\sigma_y = 2.99$  mm

# Results: Experiment 1

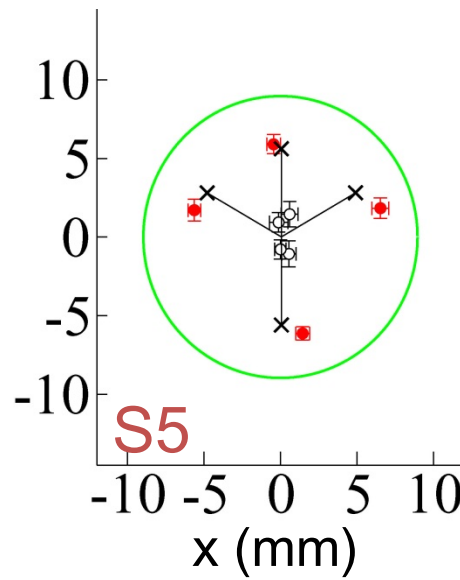
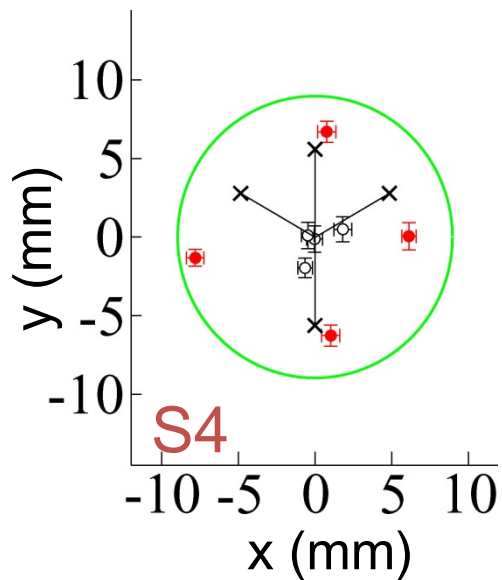
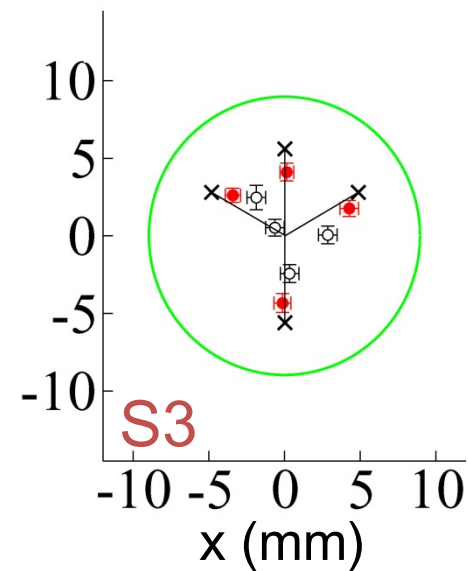
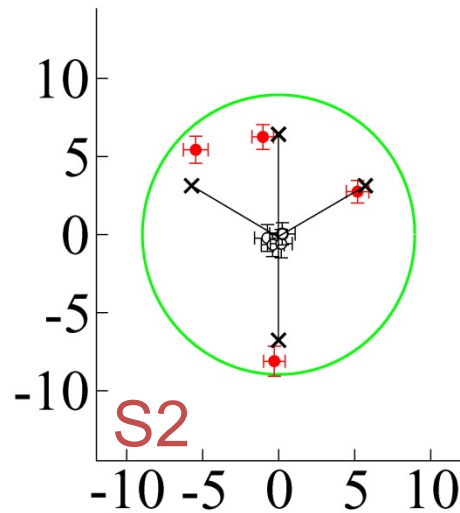
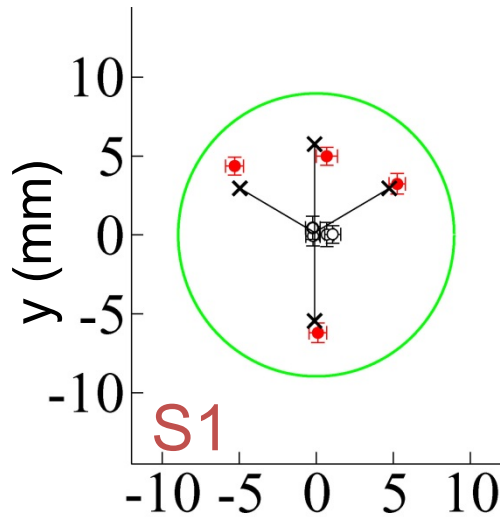
## Comparison with experiment



- exp., penalty = 0
- exp., penalty = 500
- x model, penalty = 500

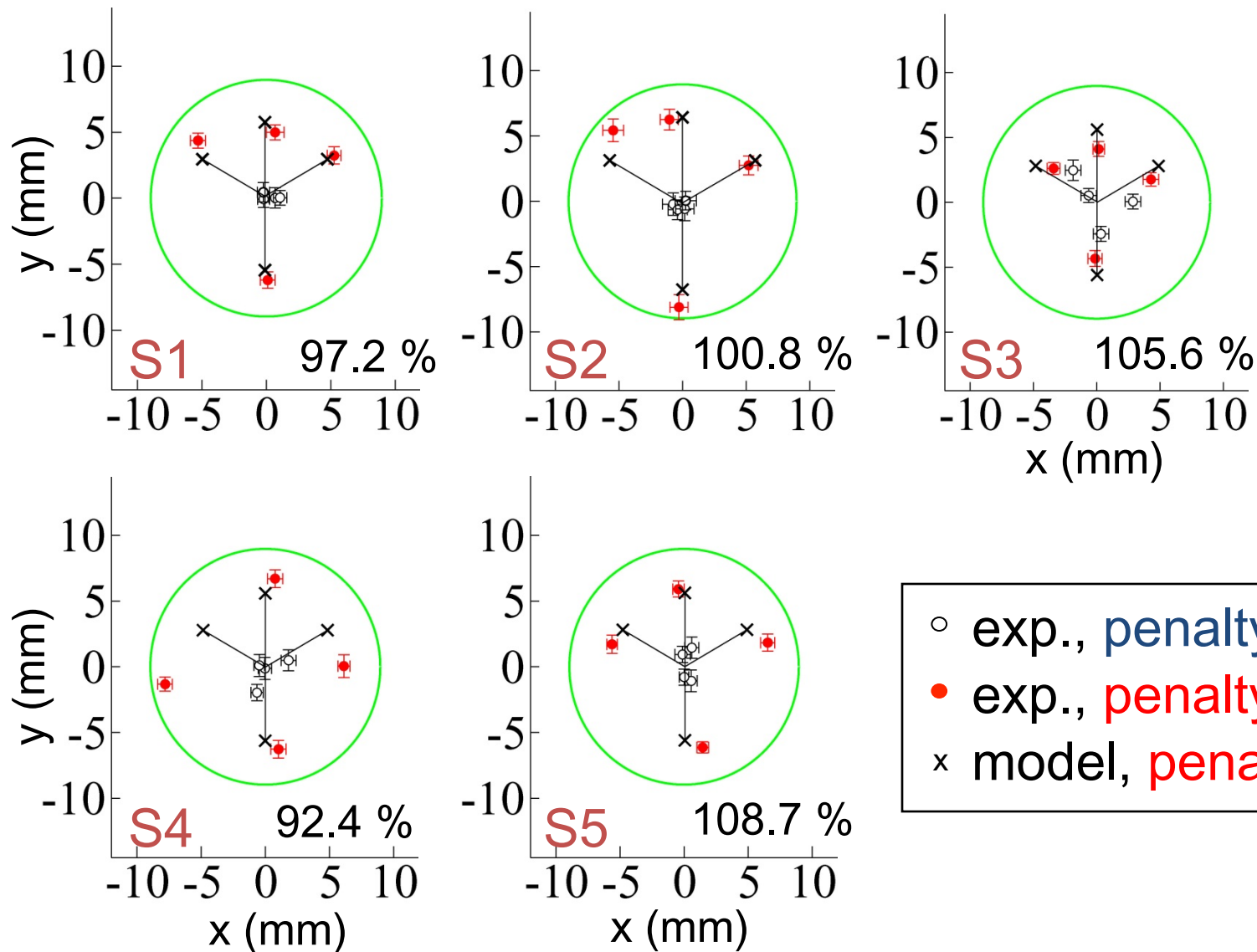
Subject S5,  $\sigma_y = 2.99$  mm

# Results: Experiment 1



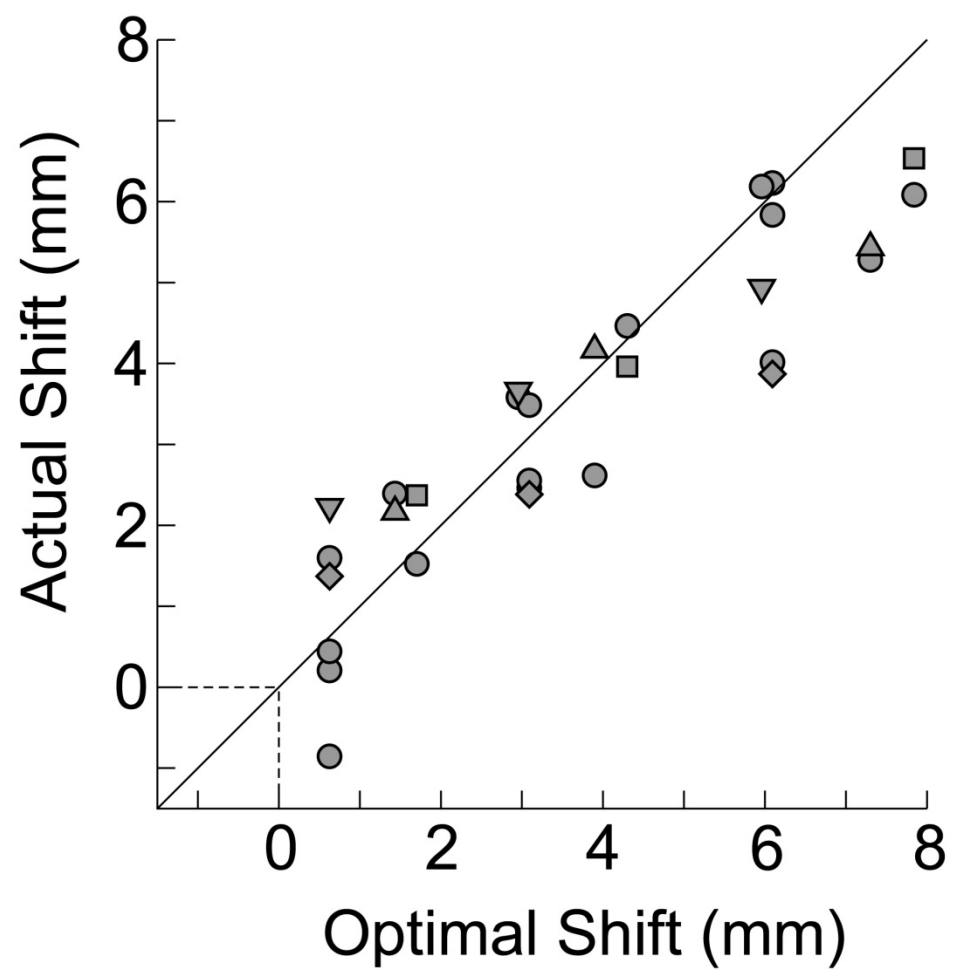
- exp., penalty = 0
- exp., penalty = 500
- x model, penalty = 500

# Results: Experiment 1

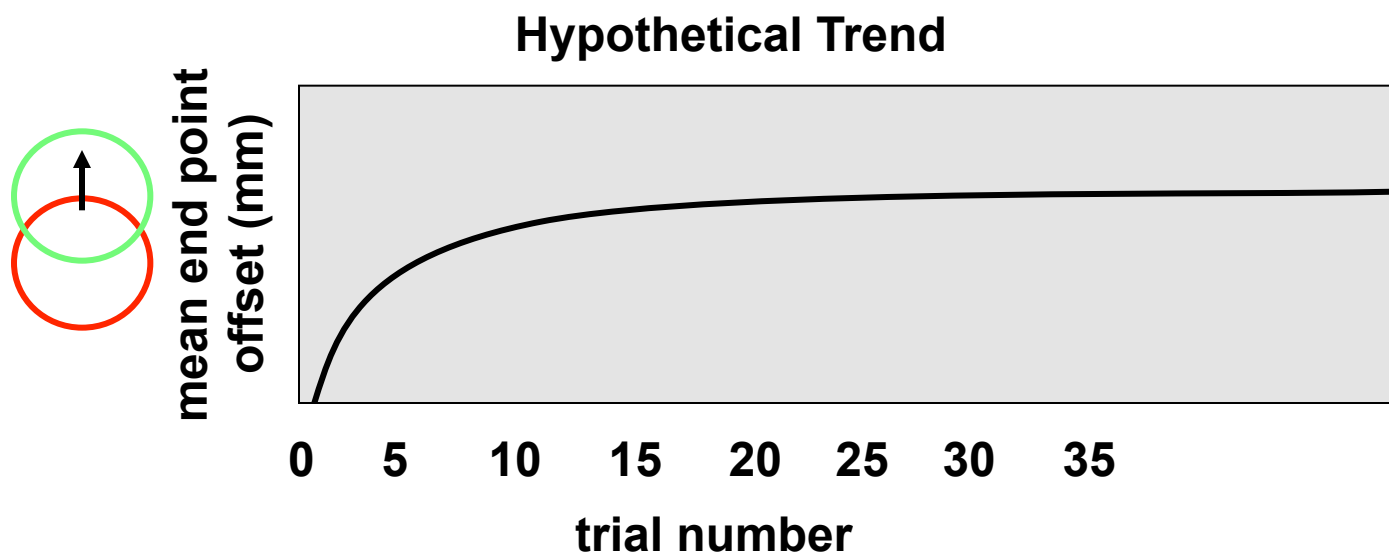




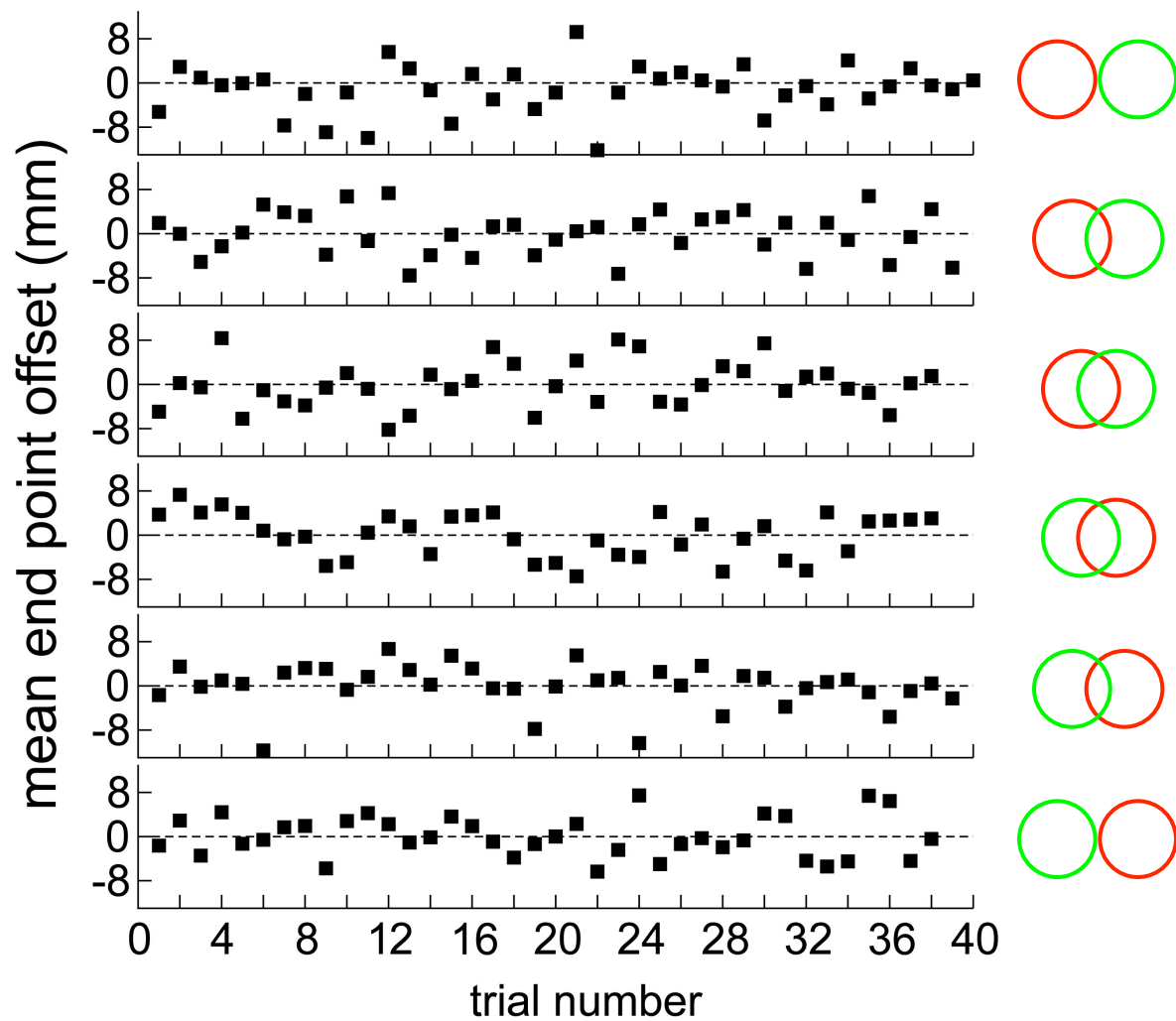
C



**b**



a



# *Movement Planning*

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*Trommershäuser, Gepshtein, Maloney, Landy & Banks, (2005),  
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*Trommershäuser, Maloney & Landy (2008), TiCS, 12(8), 291-297.*

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*and more ...*