

# **Methods in AI Research**

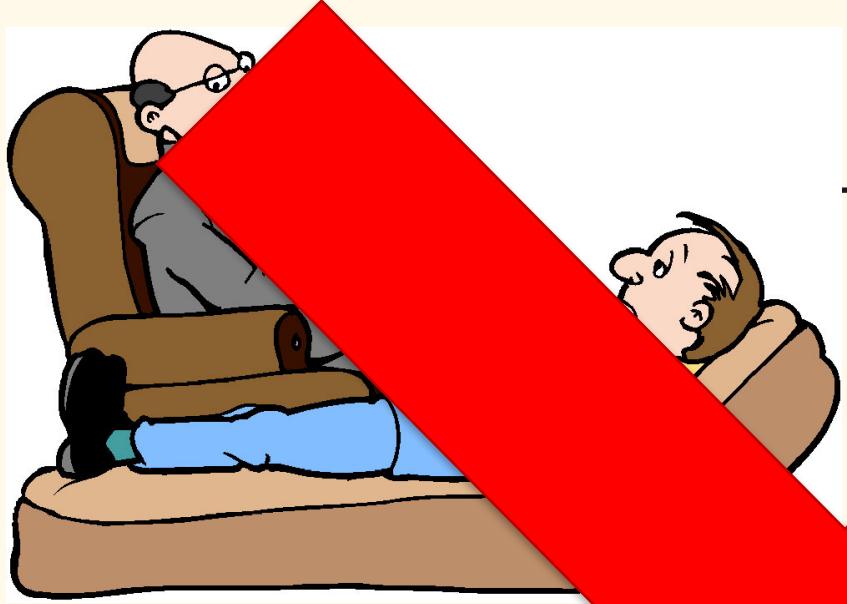
## **Cognitive Processing 1:**

### **Modeling human behavior**

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**Chris Janssen**

**c.p.janssen@uu.nl**  
**www.cpjanssen.nl**



# In spirit of Dartmouth workshop

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“The study is to proceed on the basis of the conjecture that every aspect of learning or any other feature of intelligence ***can in principle be so precisely described*** that a ***machine can*** be made to ***simulate*** it.”

Some topics:

- ✓ Language processing & -production
- ✓ Neural nets (cognitive modeling)
- ✓ Abstraction (from sensor data to concepts)
- ✓ Creativity

# Today's topics

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- **Why use models?**
  - In science
  - In practice / industry
- **What is a cognitive architecture?**
  - What are the (dis-)advantages?
  - What is contrast with cognitive model?
  - Example: multitasking in ACT-R
- **What makes a good model?**
  - Level of abstraction
- **If you want to know more...**

# Psychology: Diverse field

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Memory

Emotions

Language

Visual  
perception

Learning

Math

# Physics: Converging field

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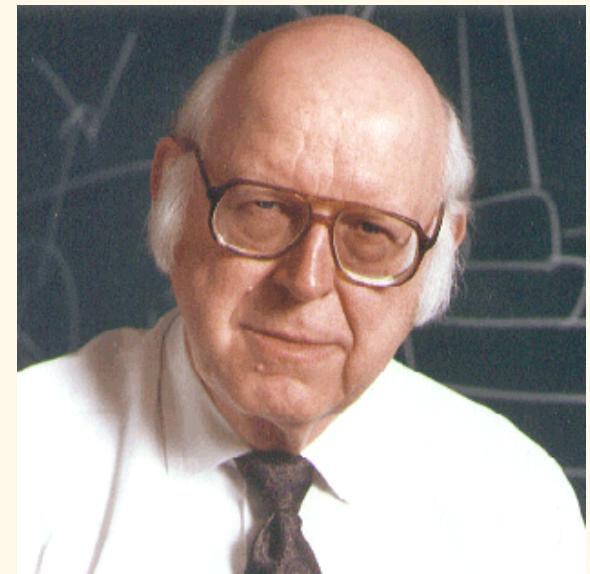


# Core idea of Cognitive Modeling (& AI)

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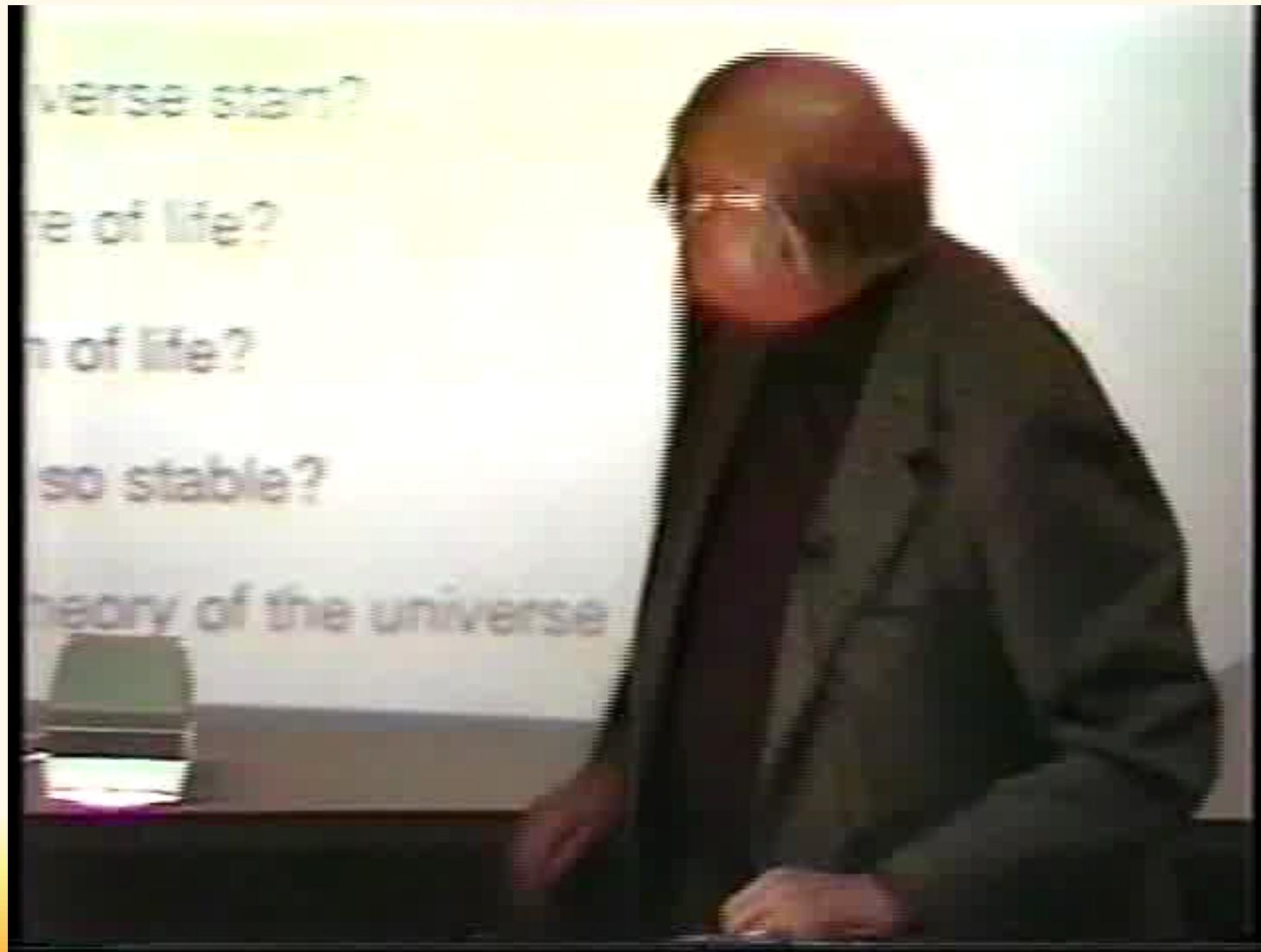
- ‘Make psychology more like physics’
- How can the mind occur in the physical universe?

→Unified Theory of Cognition  
(Allen Newell, 1990)



<http://act-r.psy.cmu.edu/misc/newellclip.mpg>

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# Why use computer models/simulations?

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1. *Formal* specification of a theory: working code
  2. Working code is *detailed*
    - Allows for scrutiny & scientific discussion
  3. Model can make *predictions*
  4. Understanding by *building*
- 
- Like physics:
    - Describe nature using equations
    - Apply equations to dataset / situation

# Understanding by building

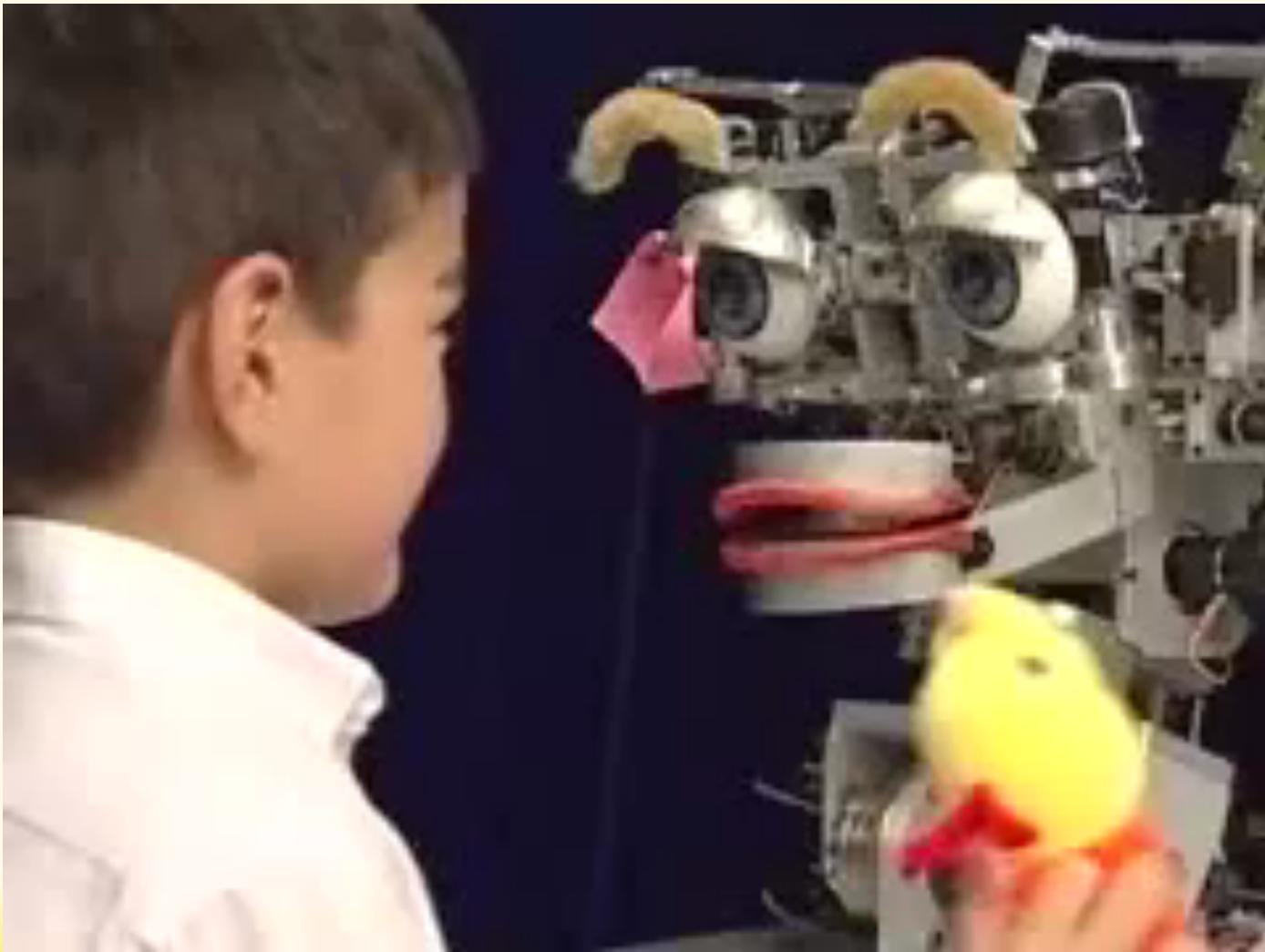
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Robocup Standard League Finals 2014

# Understanding by building

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MIT's Kismet

# Understanding by building



See: <http://cog.cs.drexel.edu/distract-r/>

# Why use computer models/simulations?

---

1. Formal specification of a theory: working code
  2. Working code is detailed
    - Allows for scrutiny & scientific validation
  3. Model can make predictions
  4. Understanding is built through validation
    - Physics:
      - Describe nature using equations
      - Apply equations to dataset / situation
- Scientific motivation**

# Why use models in practice/industry?

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- Intelligent tutor systems
- Intelligent game opponents
- Usability test
- Adaptive interface: interruptions

→ *Applied systems informed by theory*

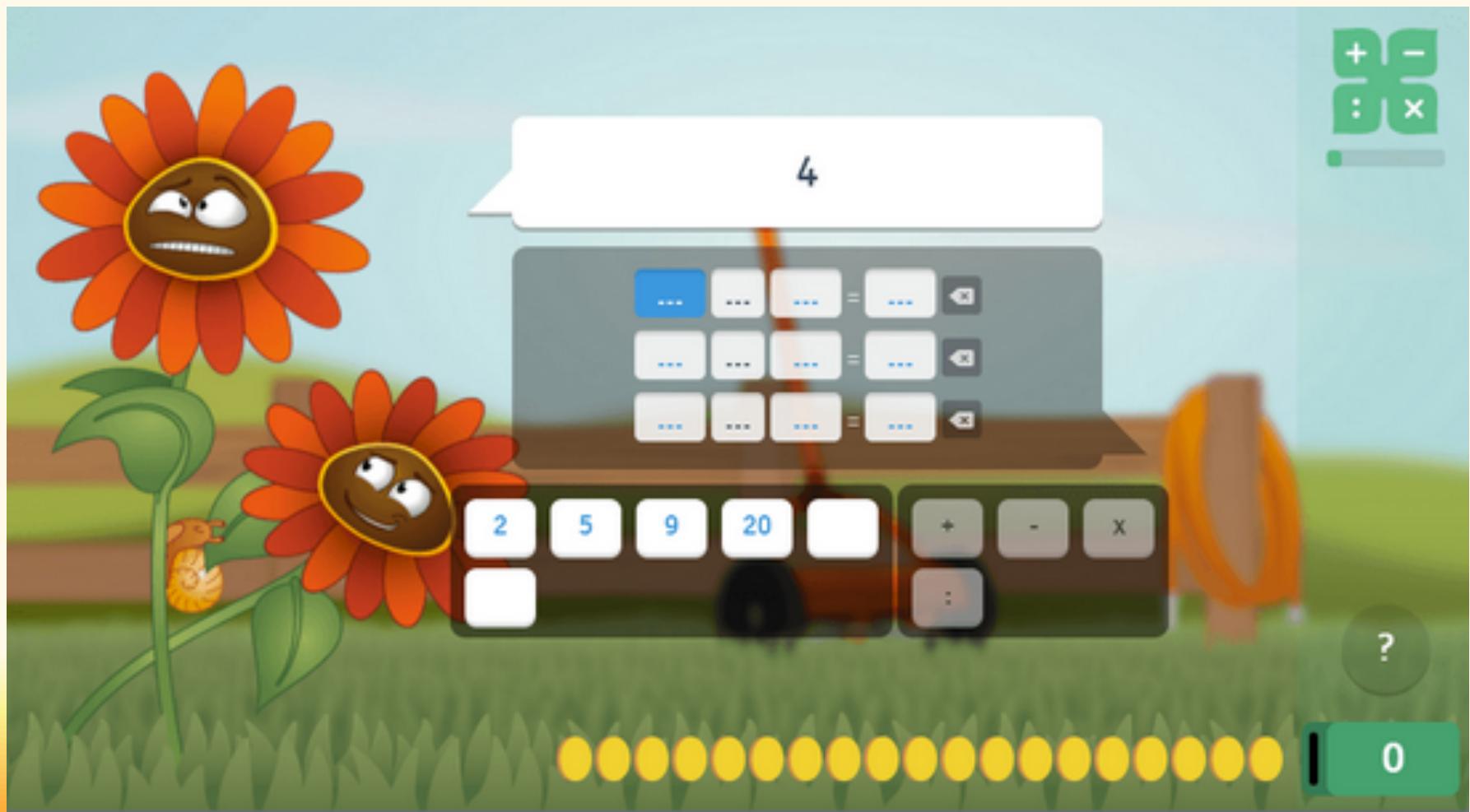
# Intelligent tutor systems

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- $2X + 7 = 27$
- $X = ?$
- Answer student:  $x = 17$
- Questions a tutor system addresses:
  - What went wrong?
  - What to practice next?
  - What instructions should be given?

# Intelligent tutor systems: State of the art

- [www.rekentuin.nl](http://www.rekentuin.nl) >1600 schools



# Intelligent tutor systems: State of the art

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**Anderson, J. R., Betts, S., Ferris, J. L., & Fincham, J. M. (2012). Tracking children's mental states while solving algebra equations. *Human brain mapping*, 33(11), 2650-2665.**

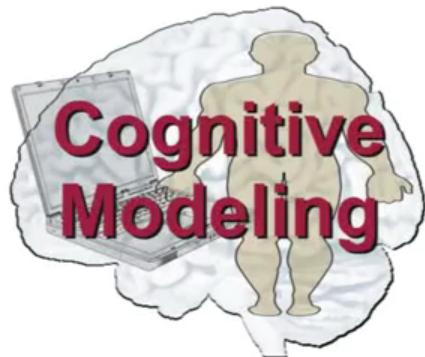
For video see: [http://act-r.psy.cmu.edu/wordpress/wp-content/uploads/2013/09/kidalg\\_mix\\_ascaled\\_text1.mp4](http://act-r.psy.cmu.edu/wordpress/wp-content/uploads/2013/09/kidalg_mix_ascaled_text1.mp4)

# Intelligent game opponents



“Beating the human” – not our goal

# Intelligent game opponents: State of the art



EBERHARD KARLS  
UNIVERSITÄT  
TÜBINGEN



## Mario AI

Learning Autonomous Behavior in Mario Bros



<https://www.youtube.com/watch?v=ApIG6KnOr2Q>

# Intelligent game opponents: State of the art



<https://www.youtube.com/watch?v=FAU2I4ZgDsw>

Siebert, Gray, Lindstedt (2017) Topics in Cognitive SCience

# Intelligent game opponents: State of the art



Sangster, Gray (2017) MathPsych; Sangster, Medonca, Gray (2016) Human Factors conference

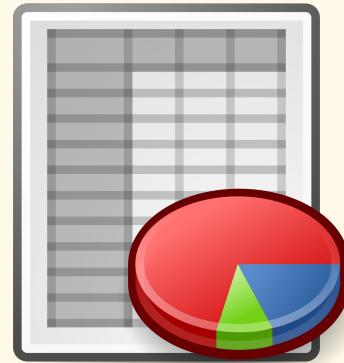
# Usability Test: Distract-R



See: <http://cog.cs.drexel.edu/distract-r/>

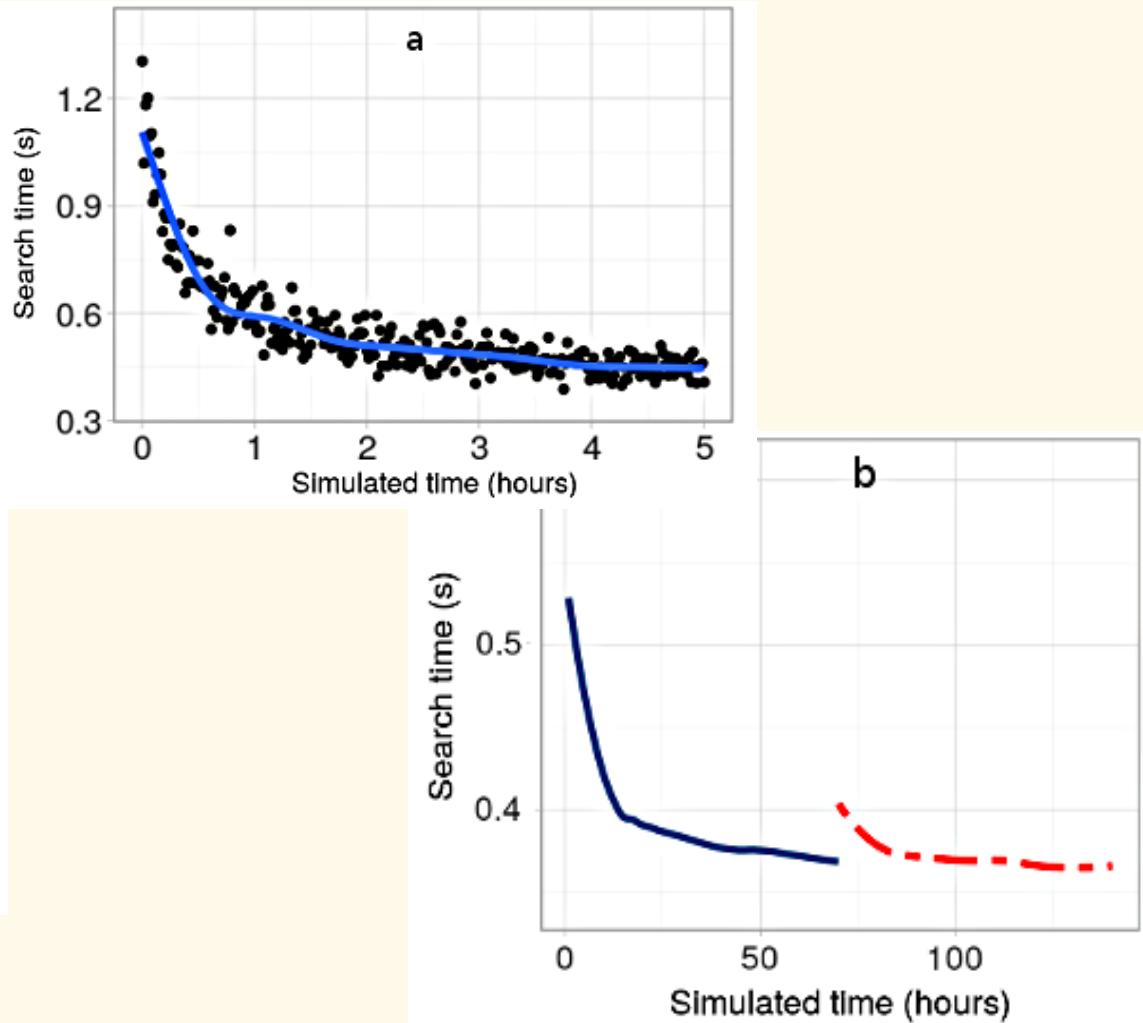
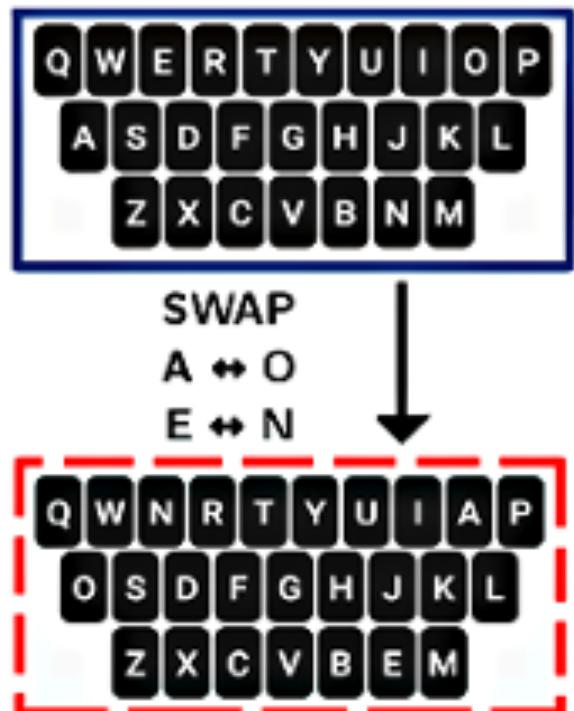
# Adaptive interface: interruptions

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Iqbal, S. T., & Bailey, B. P. (2010). Oasis: A framework for linking notification delivery to the perceptual structure of goal-directed tasks. *ACM Transactions on Computer-Human Interaction (TOCHI)*, 17(4), 15:1–28.

# Software & interface (re-) design



<https://users.aalto.fi/~jokinej10/visual-search/>  
Jokinen et al (2017; CHI)

# Why user models for practice?

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- 1. Have human-like experiences**
  - Game opponent -> requires understanding first
- 2. Make appropriate, (adaptive) decisions due to understanding of user:**
  - Feedback (tutor)
  - Difficulty (game)
  - Usability
  - Adaptive interface (interruptions)
  - Software & interface (re-) design

# Today's topics

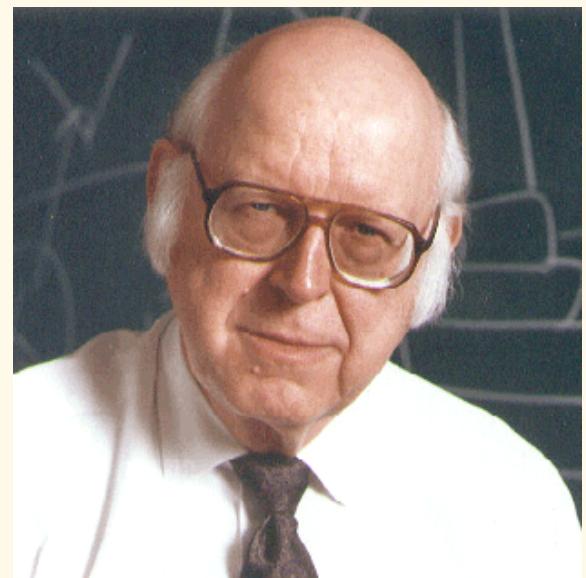
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- Why use models?
  - In science
  - In practice / industry
- What is a cognitive architecture?
  - What are the (dis-)advantages?
  - What is contrast with cognitive model?
  - Example: multitasking in ACT-R
- What makes a good model?
  - Level of abstraction
- If you want to know more...

# More about the basic science...

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→Unified Theory of Cognition  
(Allen Newell, 1990)



# **Unified Theory of Cognition**

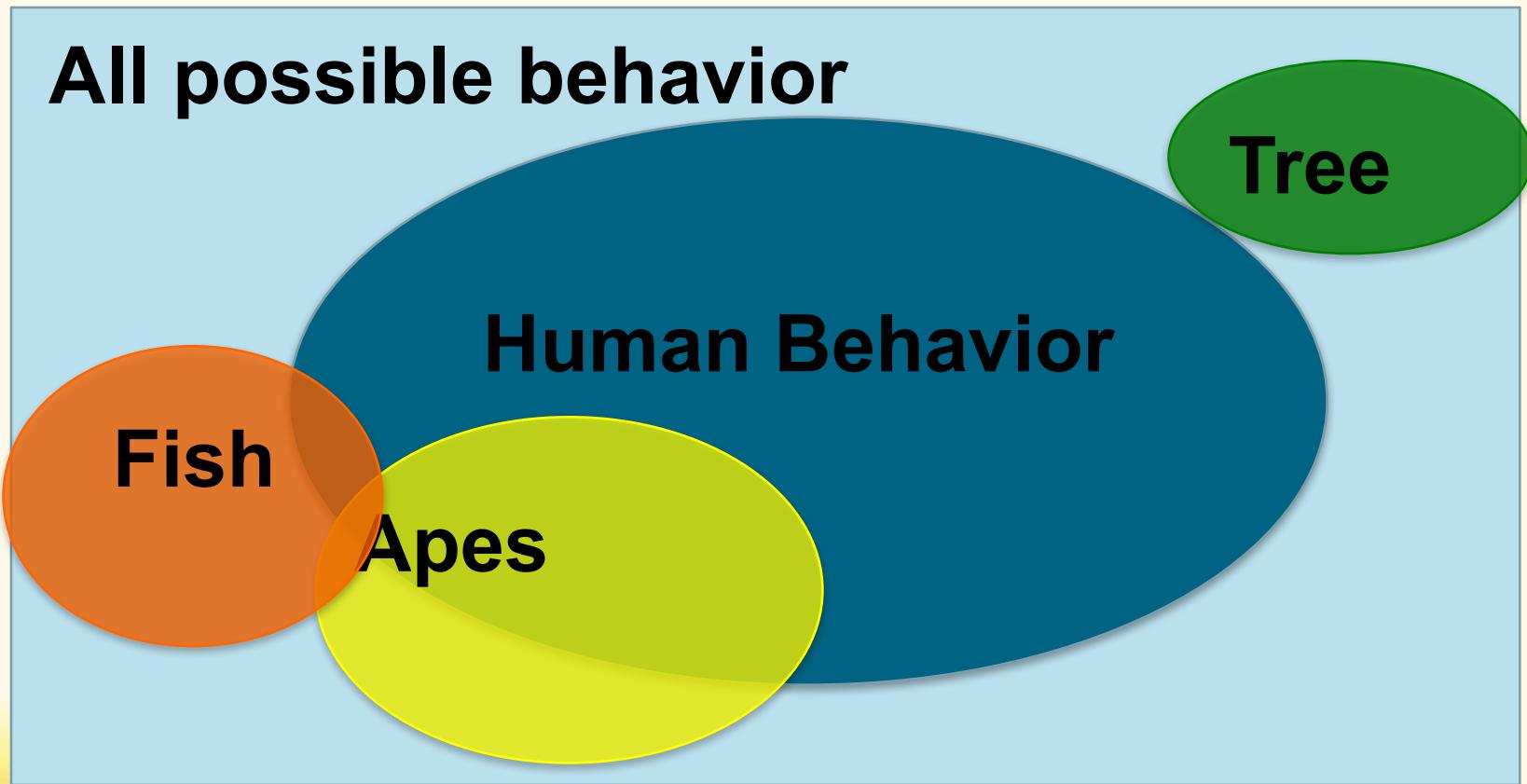
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## **illustration with Venn Diagrams**

**All possible behavior**

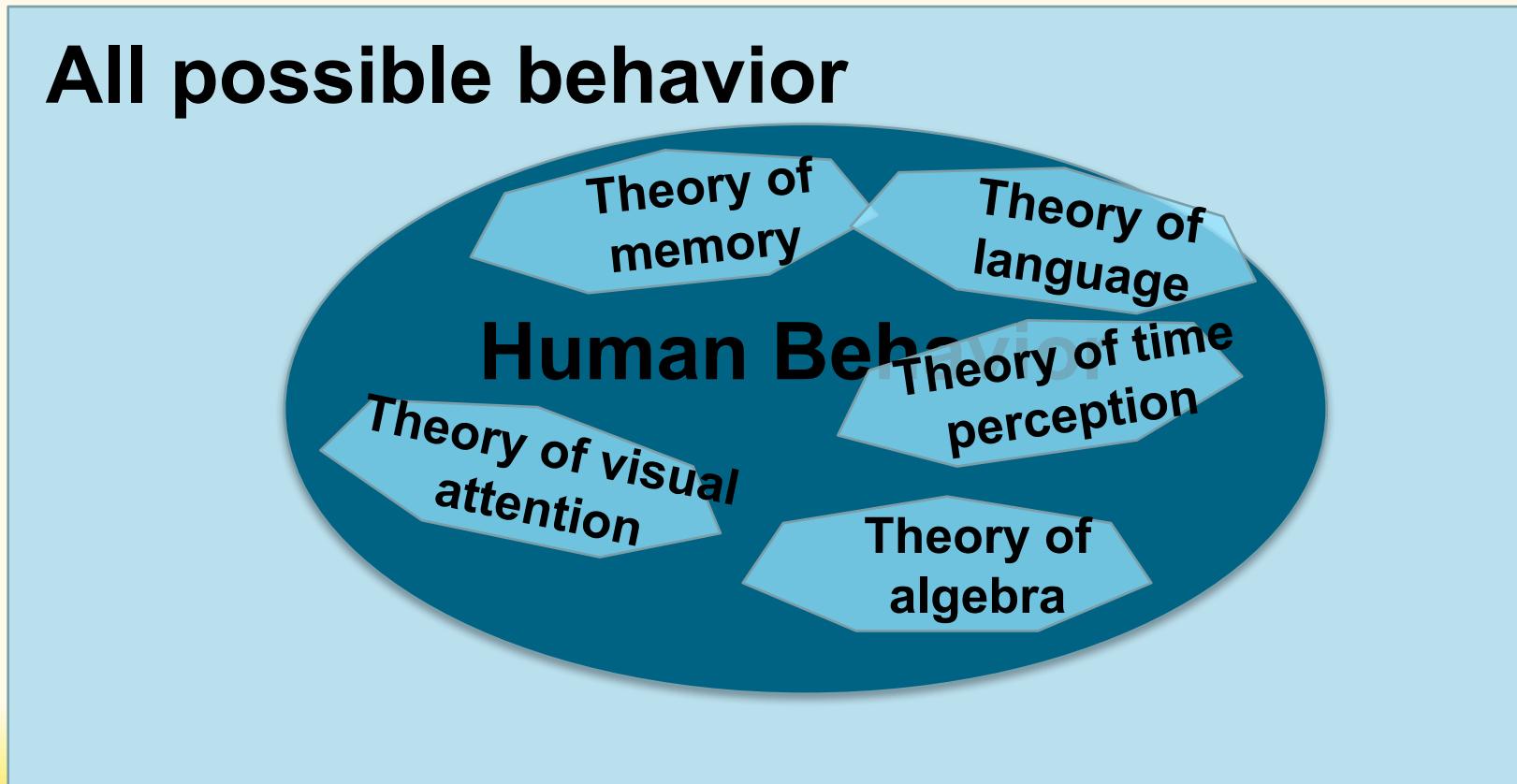
# Unified Theory of Cognition

## illustration with Venn Diagrams



# Unified Theory of Cognition

## illustration with Venn Diagrams



# Unified Theory of Cognition

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## illustration with Venn Diagrams

All possible behavior

A Venn diagram consisting of two overlapping circles. The larger circle is light blue and labeled "All possible behavior". The smaller circle is also light blue and labeled "Cognitive Architecture: 1 Theory of Human Cognition". The two circles overlap significantly, representing how a single unified theory can encompass all possible cognitive behaviors.

Cognitive Architecture:  
1 Theory of Human Cognition

# Unified Theory of Cognition

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- **Newell's goal:**
  - Develop 1 theory that can describe (almost) all facets of human behavior
  - Describe interactions between theories of different domains/situations/components
- ***Cognitive Architecture*** (Bell & Newell, 1971)
  - Within the architecture you develop models of specific tasks/settings

# Example of Cognitive Model in Architecture

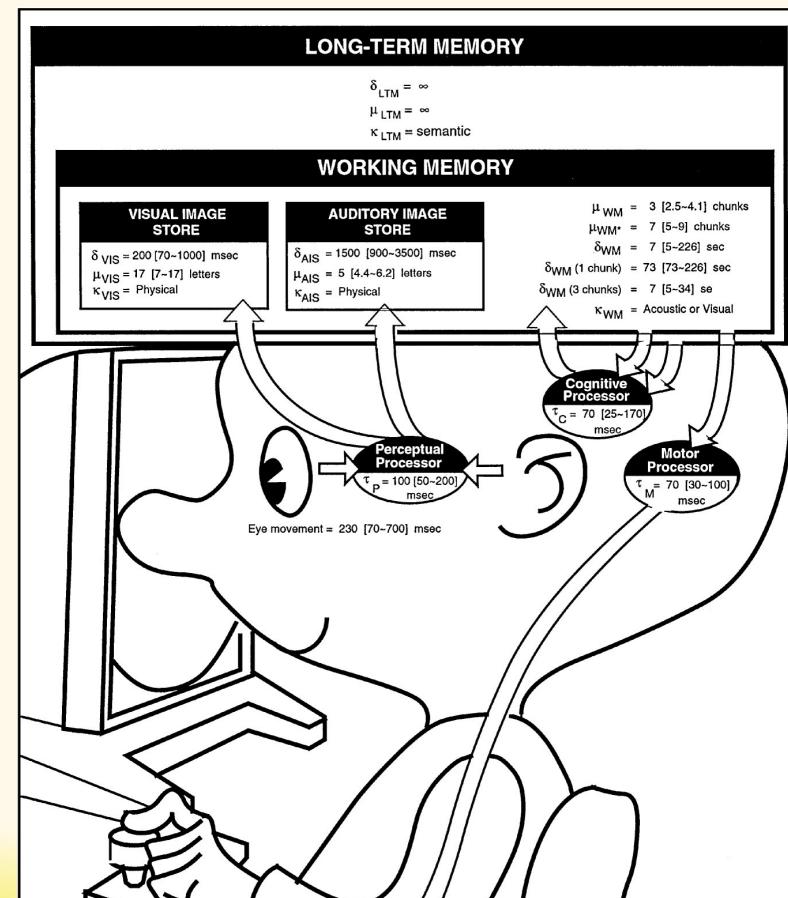
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# Cognitive Architecture & Cognitive Model

- **Cognitive Architecture (John Anderson, 2007):**  
*“a specification of the structure of the brain at the level of abstraction that explains how it achieves the function of the mind”*

- **Explains:**
  - What brain can & can't do (function)
  - How it does that (structure)
  - What *general* parameters and equations govern behavior



# Cognitive Architecture & Cognitive Model

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- **Cognitive Model**
  - A model of a *specific* task or process
  - Developed *within* or *inspired by* an architecture

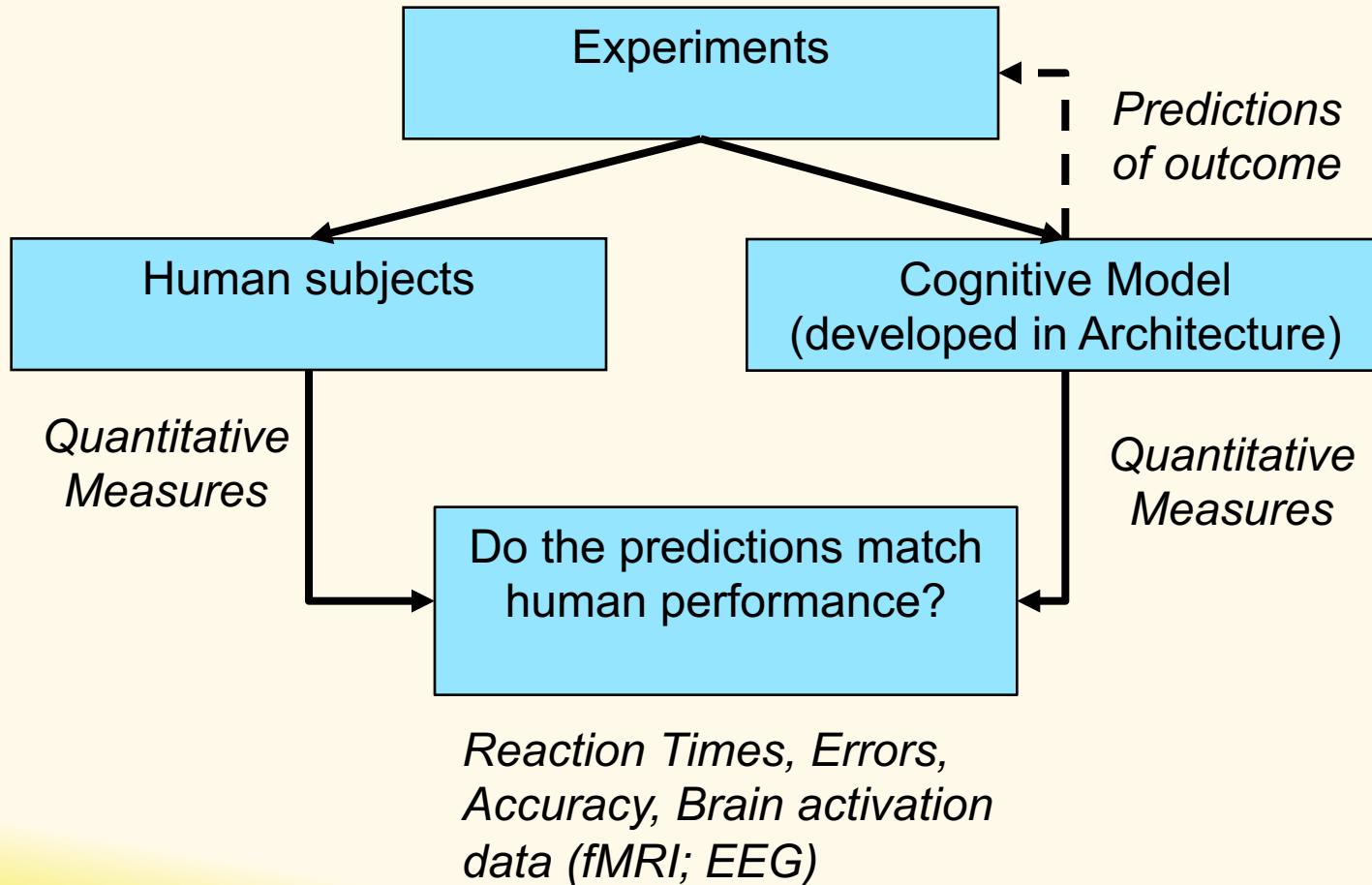
# Cognitive Architecture & Cognitive Model

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- **Architecture questions:**
  - What is our general memory capacity?
  - How do we forget information over time in general?
  - How much visual information can we process per time unit?
  - How do we control our hands in general?
- **Model questions:**
  - How do we calculate  $101 \times 7 - 3$  (given architecture)
  - How do we control a car (given architecture)
  - How do we divide time between driving and making a phone call (given architecture)

# Testing a Model

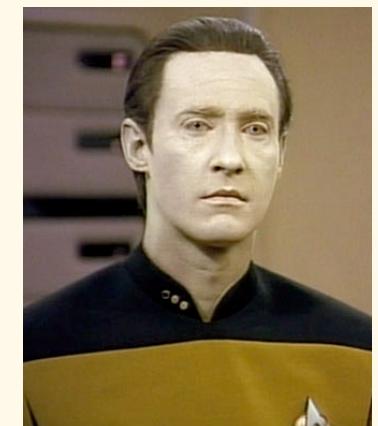
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# Cognitive Architectures & Turing Machine

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- Alan Turing (1936): Turing Machines can compute all ‘computable’ functions
  - Computer simulations can (in principle) do more than humans
- Implication:
  - *Constrain* the cognitive architecture
  - Do tasks in a “human way”



# Cognitive architectures vs Turing machine

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© Festo

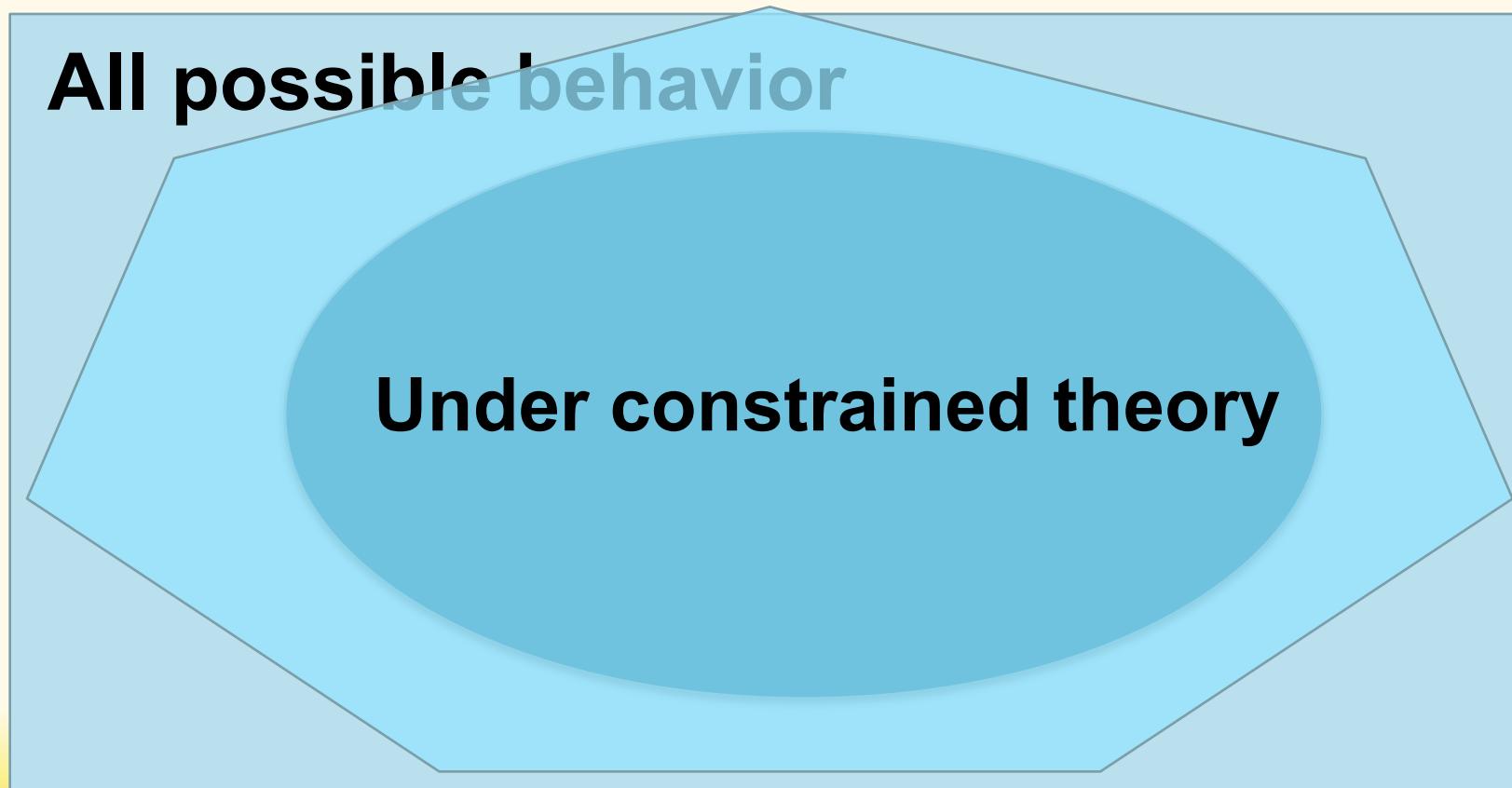
**Cognitive Science style AI:  
Understand human by replicating**

**Engineering :  
Develop (more) intelligent systems**



# Venn diagrams – revisited

Under constrained model (Turing machine / Mr Data)



# Venn diagrams – revisited

## Over constrained model

All possible behavior

Human Behavior

Over constrained  
theory

# **Unified Theory of Cognition**

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## **Ideal architecture**

**All possible behavior**

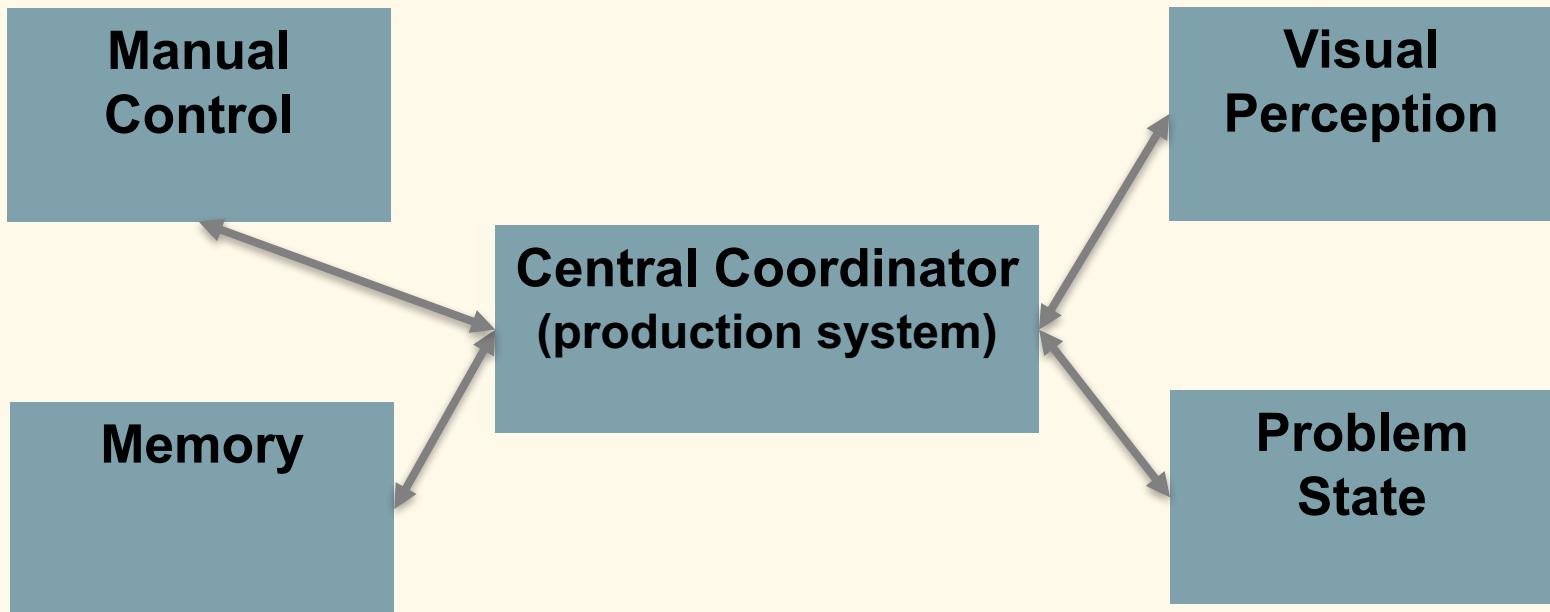


**“Ideal”  
Cognitive Architecture**

# Example: Multitasking in ACT-R

(Anderson, 2007; Salvucci & Taatgen, 2011)

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# **One model in cognitive architecture**

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- 1. Identify which functions are achieved by the brain/cognition on each time step**
- 2. Specify each process using principles of the architecture**
  - Which module executes this process & how?**

**(exact details left out for now)**

# Multitasking in Cognitive Architecture

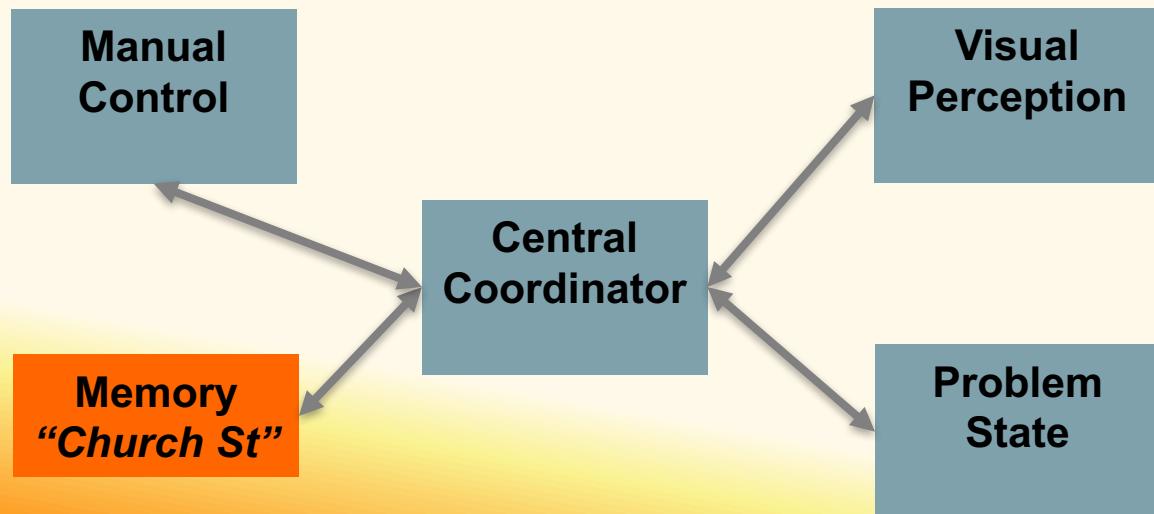
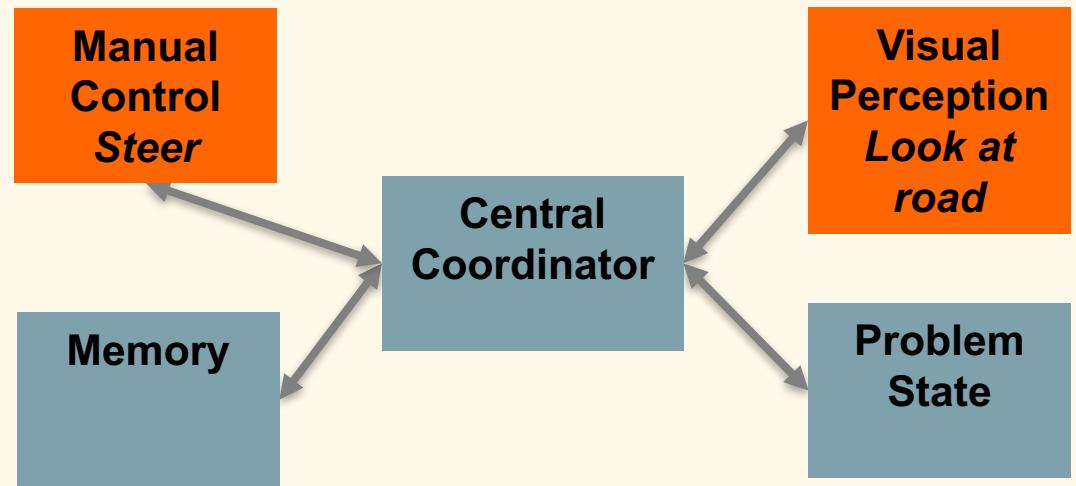
(Salvucci & Taatgen, 2008, 2011)

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- **1 task = 1 cognitive model**
- **Multitasking =  
executing multiple models concurrently**

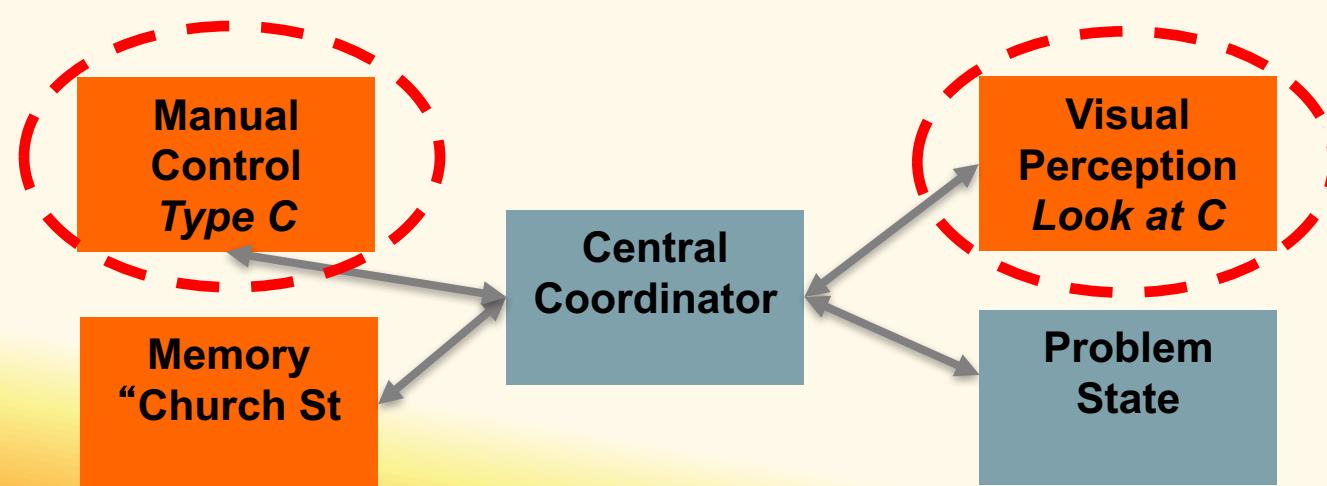
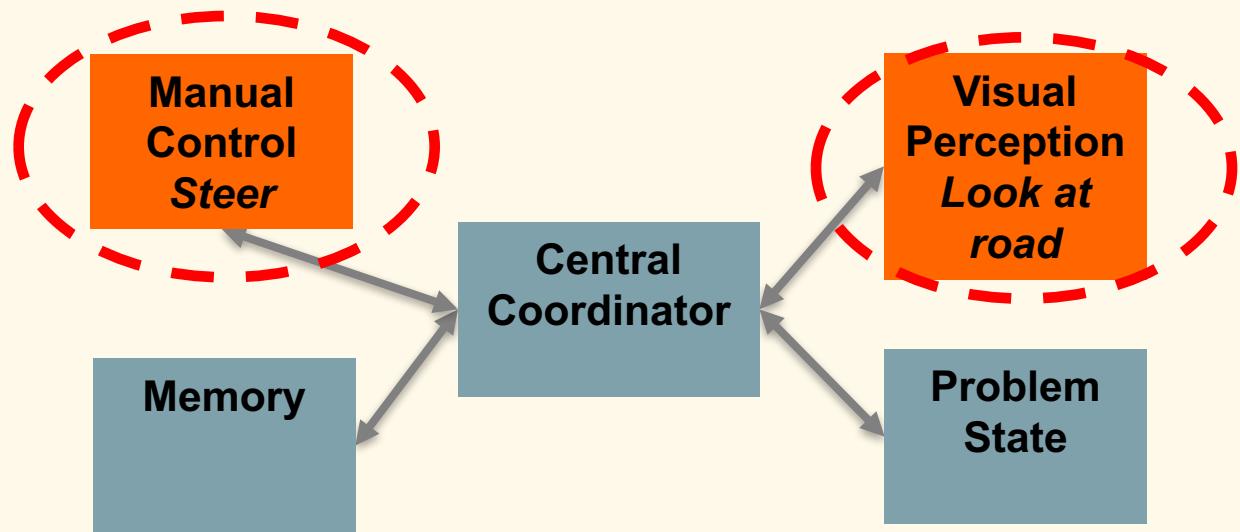
# Multitasking: Typing while driving

“Actively Driving and memory retrieval can be done at same time”



# Multitasking: Typing while driving

“Actively Driving and typing can not be done at same time”



# Models are not static

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- **Which module is active changes over time**
  - See detailed example in appendix of slides

# Example: Distract-R



See: <http://cog.cs.drexel.edu/distract-r/>

# Advantages cognitive architectures

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## 1. Formal specification: One framework to describe behavior in multiple settings

- Compare: Newton's laws in physics

## 2. Working code is detailed

- Difficult to be vague/unspecified about aspects
- Question about the model? Look at the code

## 3. Model reuse & predictions

- Test in novel settings (e.g., driving on different roads)
- Multitasking: combine models for various multitasking situations

## 4. Understand by building & active testing

# Disadvantages of architectures

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1. Requires many details (all modules)
2. Some assumptions are hidden
  - E.g. Default parameter settings
3. Not always clear which component of model is critical/essential for behavior
4. Sometimes models are *too specific or not specific enough* for the process that you are interested in
  - e.g., ACT-R misses some details on human vision
5. Falsification is difficult

# Today's topics

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- **Why use models?**
  - In science
  - In practice / industry
- **What is a cognitive architecture?**
  - What are the (dis-)advantages?
  - What is contrast with cognitive model?
  - Example: multitasking in ACT-R
- **What makes a good model?**
  - Level of abstraction
- **If you want to know more...**

# Articles to read (required for exam)

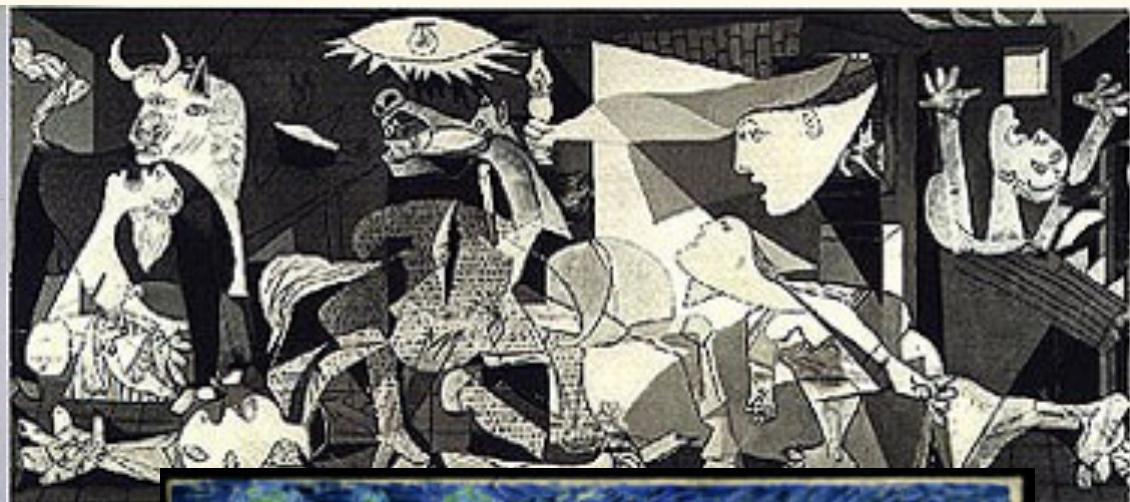
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1. Anderson, J. R. (2002). Spanning seven orders of magnitude: a challenge for cognitive modeling. *Cognitive Science*, 26(1), 85–112.
1. Cooper, R. P., & Peebles, D. (2015). Beyond Single-Level Accounts: The Role of Cognitive Architectures in Cognitive Scientific Explanation. *Topics in Cognitive Science*, 7, 243–258.

Try to get general gist  
(study questions provided at end of slide deck)

# Abstraction continuum

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# Abstraction continuum

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- Two perspectives:
  1. Newell (1990) Unified theories of cognition  
(discussed in Anderson, 2002)
  2. Marr (1982) Vision. Chapter 1  
(discussed in Cooper & Peebles, 2015)

# Abstraction level

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- **Newell: time bands**
- **Marr: levels**
- **You need to know the levels and their characteristics**  
**(See appendix of slide deck, articles)**
  - Be able to determine what is best level of abstraction for given problem
- **Models are not always 100% one band or level. These are continua.**
  - Each has relatively more focus on one band / level

# Implications of abstraction

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- Abstraction is *always* necessary:
  - Ignore (some parts of) ‘lower’ or ‘higher’ level
  - This makes research powerful:
    - No need to specify irrelevant ‘details’
    - Focus on *relevant* issues
- Issue: what degree of freedom do you take in (under-/over-) specifying a model?
  - Can you defend each of its assumptions?
  - What aspects do you need to measure in your data to develop & validate your model?

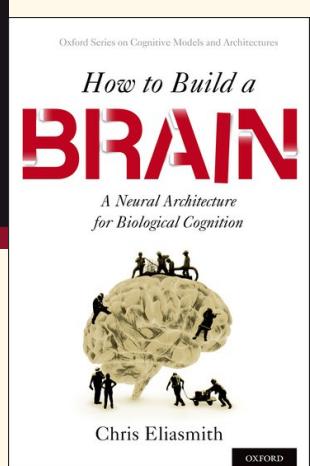
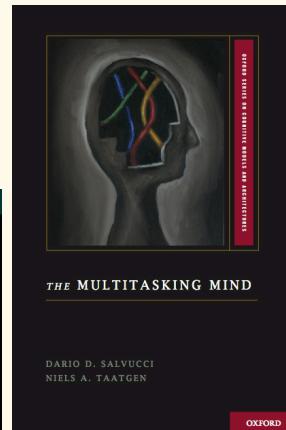
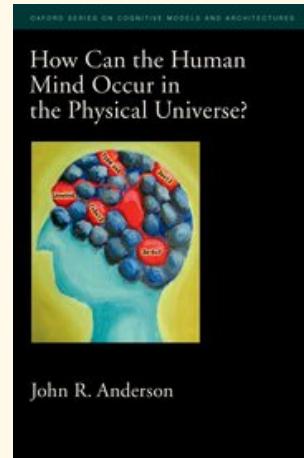
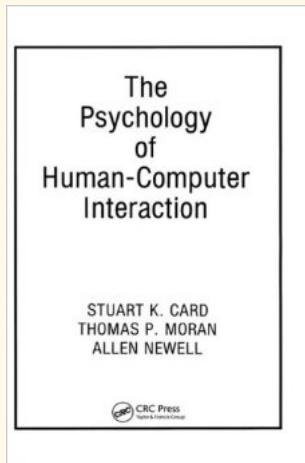
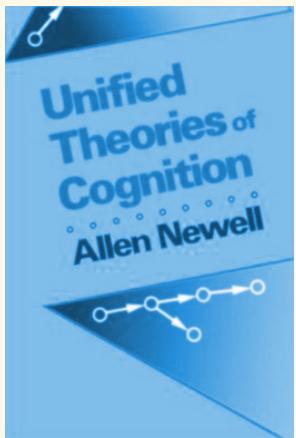
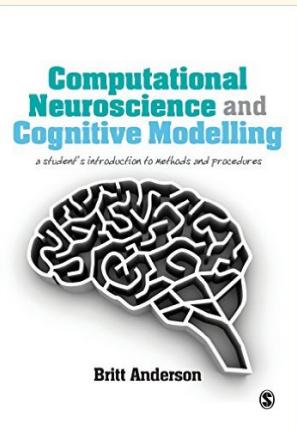
# If you want to know more

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- **Cognitive Modeling (INFOMCM): block 2**
  - Machine learning
  - Processing models
  - Bayesian models

Implementation	Ben Harvey
Algorithmic / Comp	Chris Janssen
Computational	Rick Nouwen
	& Frans Adriaans
- Programming: implement & test models in R
- Literature, lectures, presentations
- Suitable for students without Psychology degree! (who are interested)  
(in fact: aimed at!)

# If you want to know more: textbooks



Accessible, general intro  
on variety of models

Classic for  
unified theories

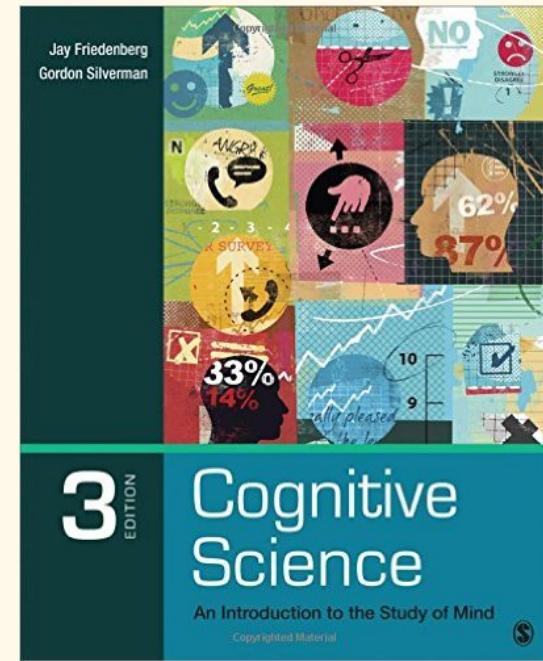
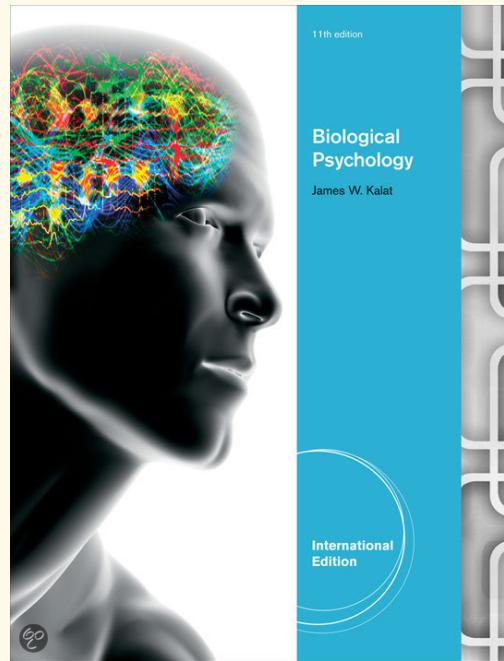
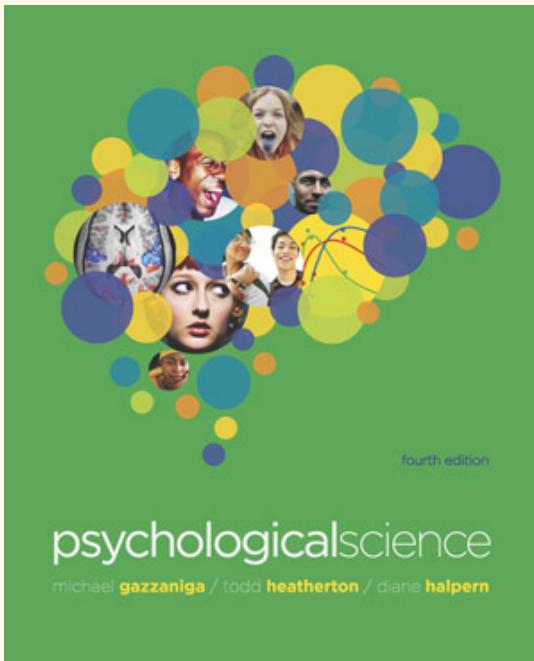
Human-Computer  
Interaction  
(older, but brilliant!)

Cognitive Architectures  
(esp Chapter 1)

Multitasking (esp C1:  
abstraction & application)

Neuroscience  
& Engineering

# For more general psych background (not modeling)



- M.S. Gazzaniga, T.F. Heatherton and D.F. Halpern (2011) Psychological Science (4th edition). ISBN 9780393913361. (general overview)
- Kalat (2013). Biological Psychology. 11th edition. ISBN: 9781111839529. (some knowledge of biology is useful)
- Friedenberg, J. & Silverman, G. (2015). Cognitive Science: an introduction to the study of mind (3rd edition). Thousand Oaks, CA.: Sage (more cognitive science / formal → most relevant to AI)

# Next week

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- Different location (check course schedule)

# Today's topics

---

- **Why use models?**
  - In science
  - In practice / industry
- **What is a cognitive architecture?**
  - What are the (dis-)advantages?
  - What is contrast with cognitive model?
  - Example: multitasking in ACT-R
- **What makes a good model?**
  - Level of abstraction
- **If you want to know more...**

# Questions?

---

**Chris Janssen**

**c.p.janssen@uu.nl  
www.cpjanssen.nl**

# **Study questions lecture (not bullet proof)**

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- **What are benefits & disadvantages of using models?**
- **What are the benefits of models for industry/practice?**
- **What is a unified theory of cognition?**
- **What is difference between cognitive model and cognitive architecture?**
- **What is relationship of cognitive models and architectures with strong/weak AI, Turing machine, over & under constrained theory**
- **What are different levels of abstraction, their characteristics, benefits, disadvantages (Newell and Marr perspective)**
- **Be able to apply these principles to case studies**

# **Study questions Anderson (not bullet proof)**

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- **Describe each of Newell's 4 bands. How do they relate to time units and systems?**
- **What is the decomposition thesis?**
- **What is the relevance thesis?**
- **What is the modeling thesis?**
- **Why is the “unit-task” level useful for each thesis?**

# **Study questions Cooper & Peebles (not bullet proof)**

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- **Describe each of Marr's 3 levels**
- **What was Marr's criticism of cognitive systems that were not rooted in CL?**
- **What are limitations of CL?**
- **What are limitations of IL?**
- **What is the value of tying the CL or IL to the ARL?**
- **What is the value of specifying a model in a cognitive architecture (or with interacting subfunctions)?**

# **Additional slides to help understand articles**

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# Abstraction continuum (Newell, 1990)

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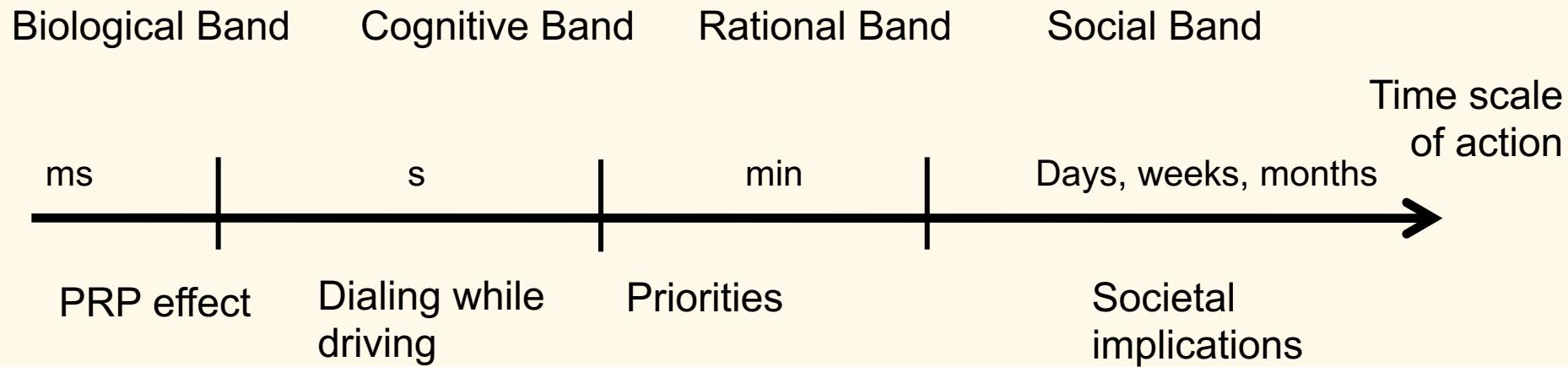
- In what process are you interested?
- What are the steps that are involved in this process over time?
- “Through which lens are you looking”?



TheDigitalPicture.com Reviews

# Abstraction continuum; time scale of action

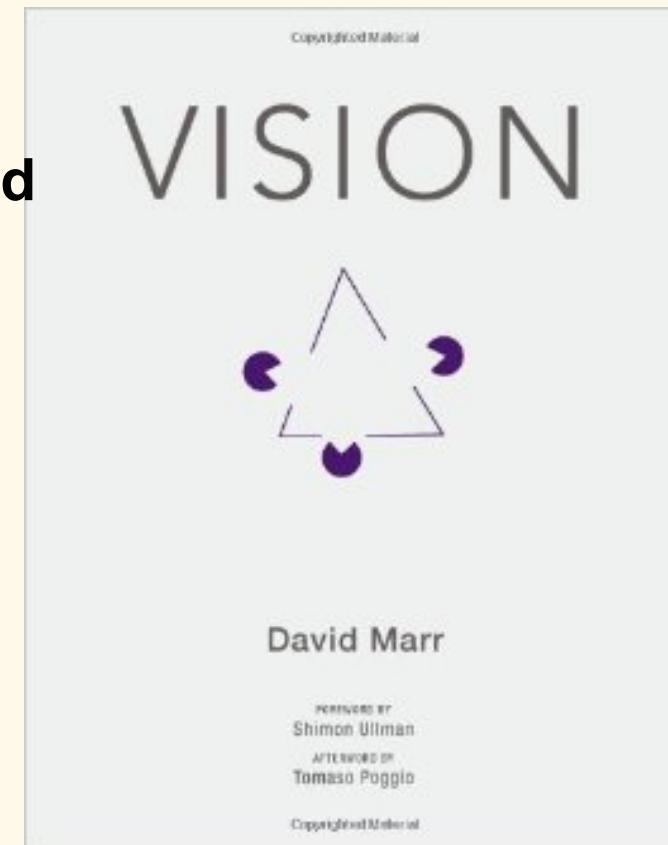
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# Abstraction continuum

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- Alternative view:  
David Marr (1982) Vision Chapter 1
  - 3 levels:
    - Computational theory
    - Algorithmic theory
    - Implementation theory
  - Each level is important
  - Each level explains *different* aspects of behavior
- More detailed



# Implications of time scale (1)

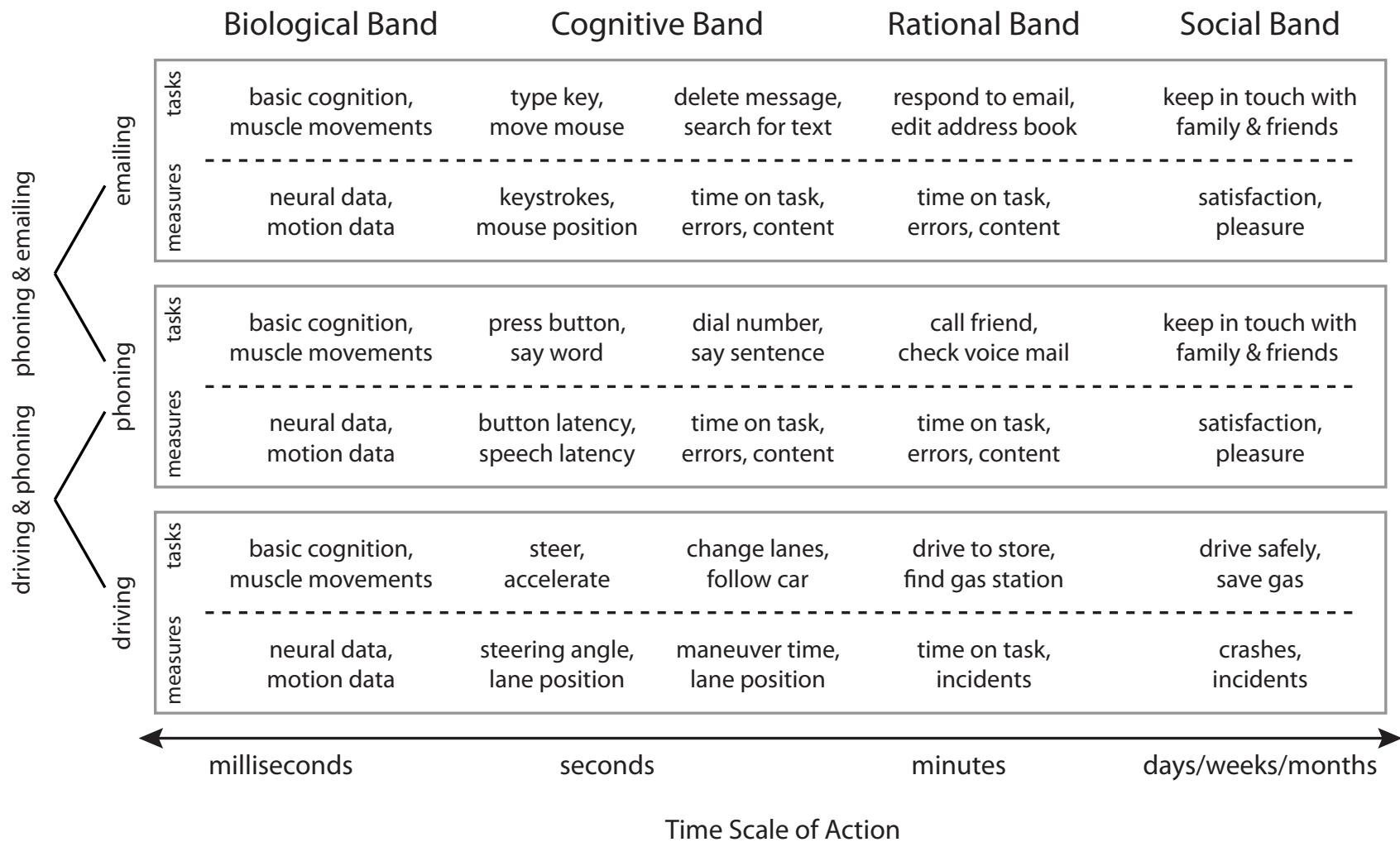
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- **Choice influences:**
  - What you *can* and *should* measure
  - What you should specify in your model
- **Cognitive band: behavioral data**
  - Reaction times, accuracy, choices, eye-gaze  
(sometimes: fMRI, EEG)
- **Biological band: data for very short time intervals**
  - Eye-gaze, EEG, fMRI, ...

# Implications of time scale(2)

---

- **Abstraction is *always* necessary:**
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  - This makes research powerful:
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    - Allows to focus on *relevant* issues
- **Issue: what degree of freedom do you take in (under-/over-) specifying a model?**
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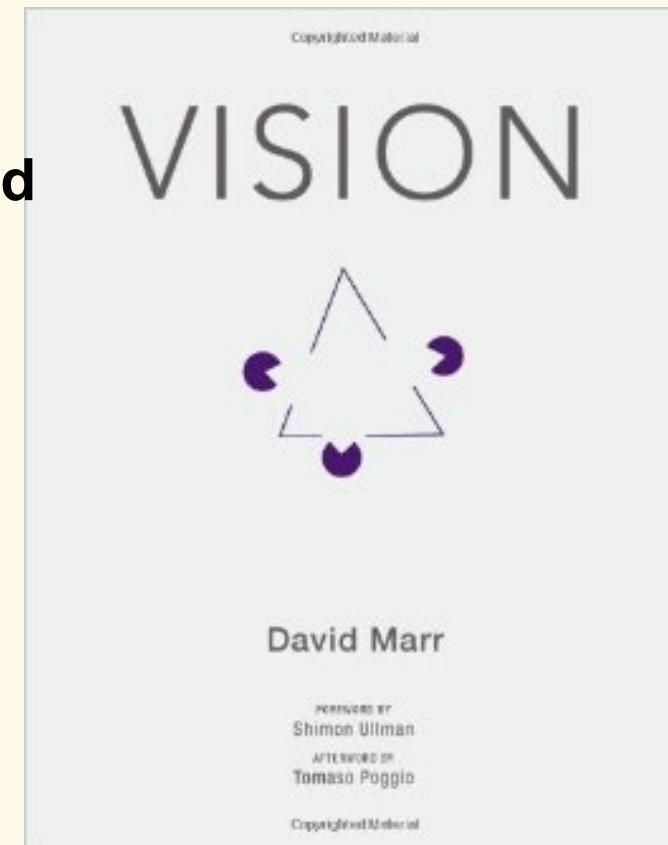
**Figure 1.3: The Abstraction Continuum (derived from Newell, 1990).**

(From Salvucci & Taatgen (2011) The multitasking mind

# Abstraction continuum

---

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# Marr: Computational level

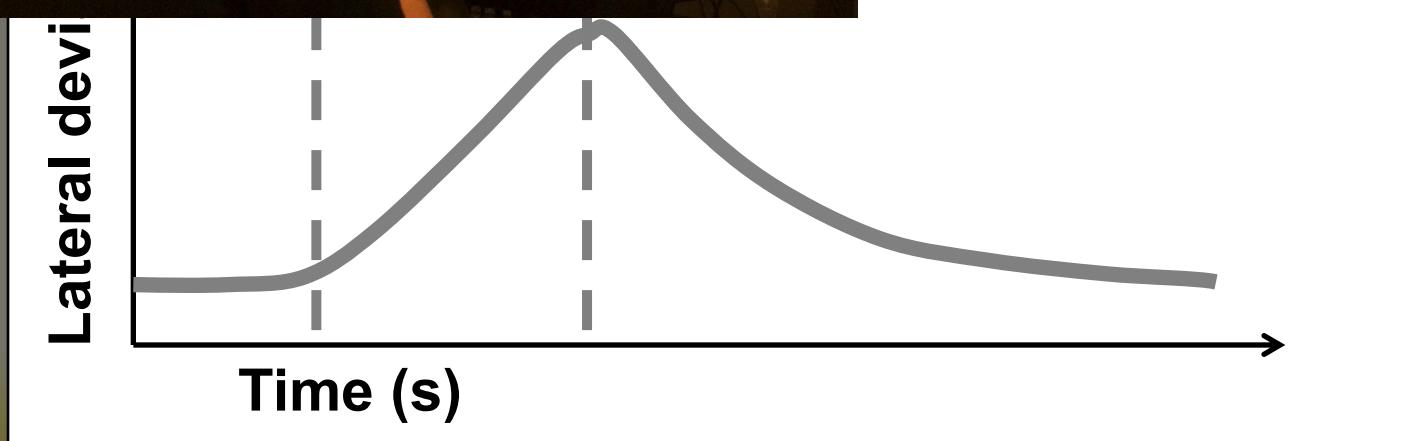
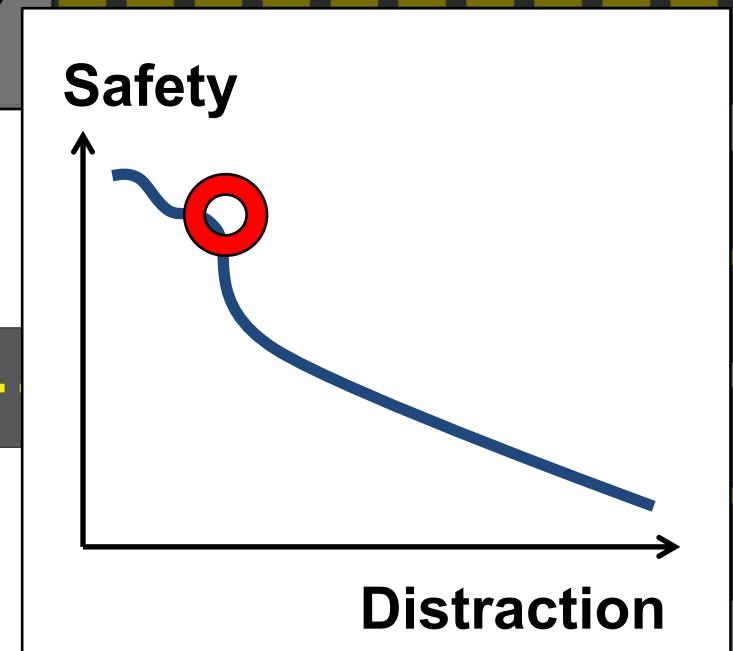
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- **Goal**  
**“WHY do we do what we do”?**
- **What they do not answer:**
  - What is done? What strategies are used? (algorithmic)
  - How does brain/neurons achieve this? (implementation)
- **Common characteristics:**
  - Small set of equations explains behavior
  - Explanation involves characteristics of statistics of the environment (e.g., adaptive argument)

# Example: Computational level (with algorithmic flavor)

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**See also video clip of example  
(computational model with algorithmic flavor)**

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<https://youtu.be/rQhj0vjVZFU>

# Marr: Algorithmic level

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- **Goal :**
  - How is a computational theory implemented?
  - What is the “input”, what is “output”
  - What type of algorithm/strategy solves this problem?
- **What they do not answer:**
  - Why is implementation like this? What is purpose (computational)
  - What is physical implementation? How brain achieves this (implementation)
- **Common characteristics:**
  - Detailed algorithms
  - That describe aspects of the process

# Example: Algorithmic level

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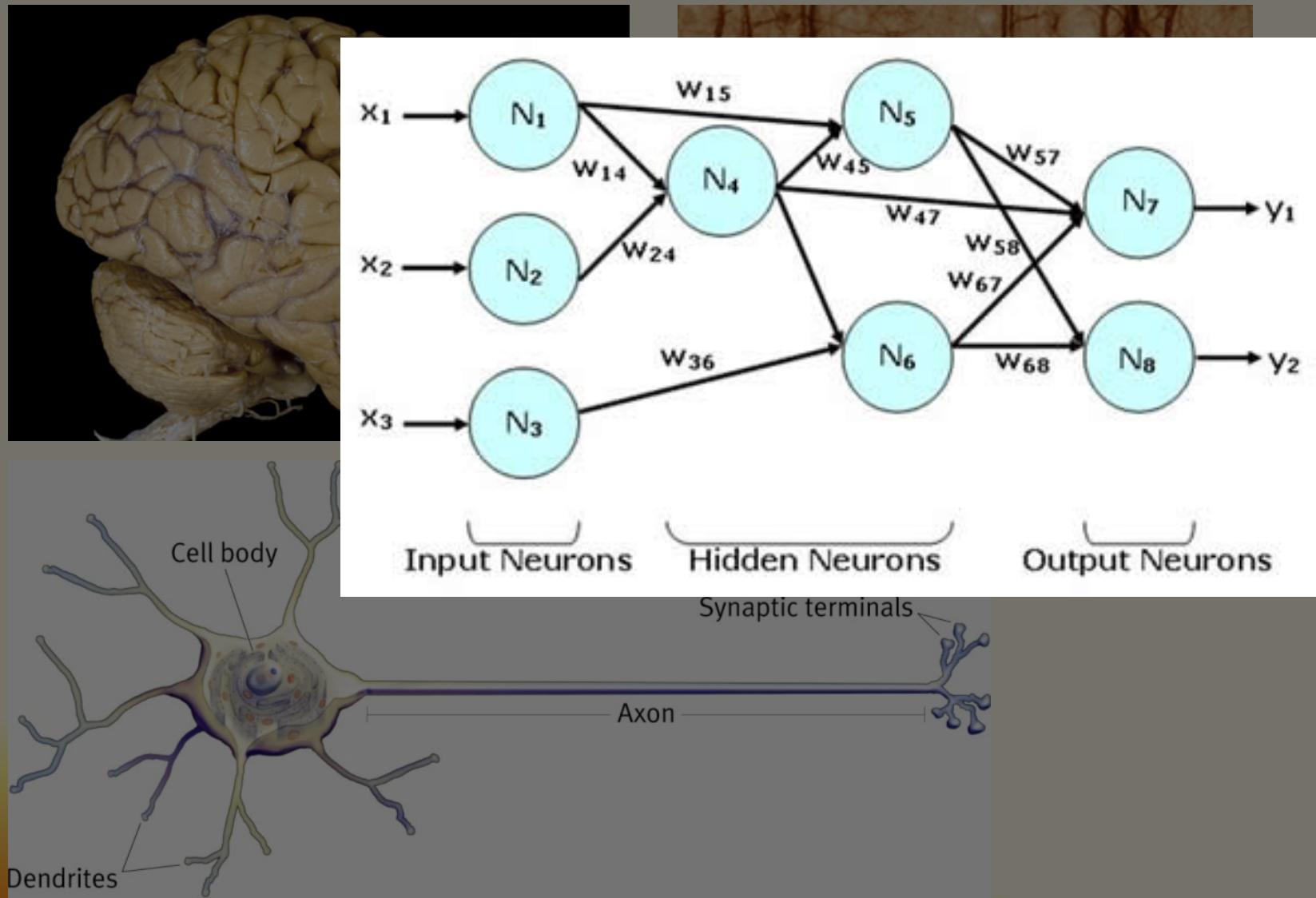


# Marr: Implementation level

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- **Goal:**
  - What is the physical implementation?
- **What they do not answer:**
  - Why is the implementation like this? What is the purpose (computational)
  - What is done? What strategies? (algorithmic)
- **Common characteristics:**
  - Describe WHERE in the brain the process occurs
  - WHAT areas are connected to each other and HOW
  - Many many minor details, small time scale

# Example: Implementation level



# Example implementation level

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- **(Artificial) Neural networks:**
  - E.g. Learning of past tense  
(e.g., Rummelhart & McClelland, 1986)
- **Good neuroscience models**
- **Architecture oriented implementation models...**
  - **Chris Eliasmith (U. Waterloo)**
    - <http://arts.uwaterloo.ca/~celiasmi/>
  - **Randall O'Reilly (U. Colorado)**
    - <http://psych.colorado.edu/~oreilly/>

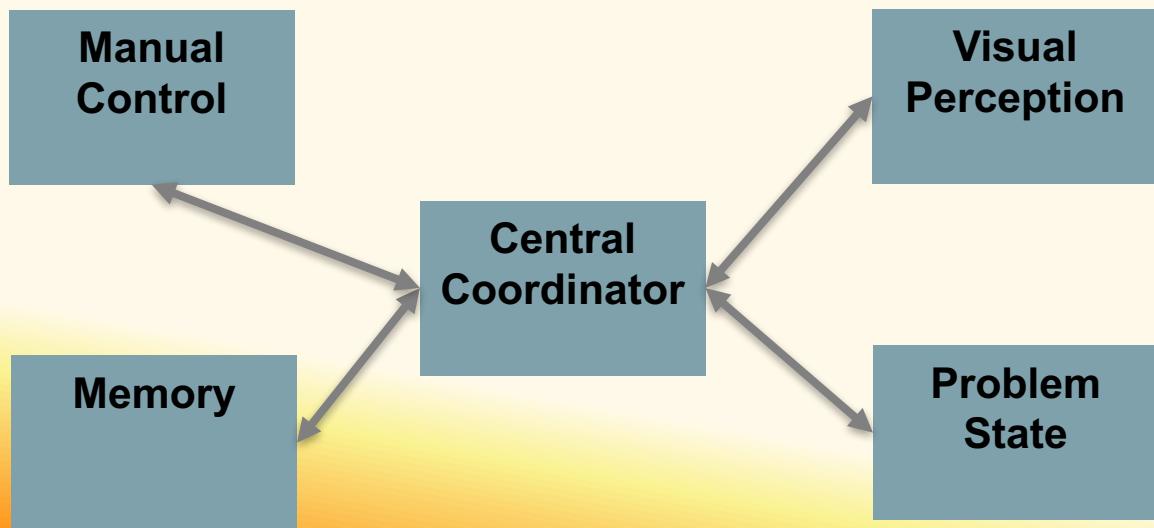
# Detailed slide process over time

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# Example 1: Type in address

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- Which functions are achieved & by which module?



# Type in address

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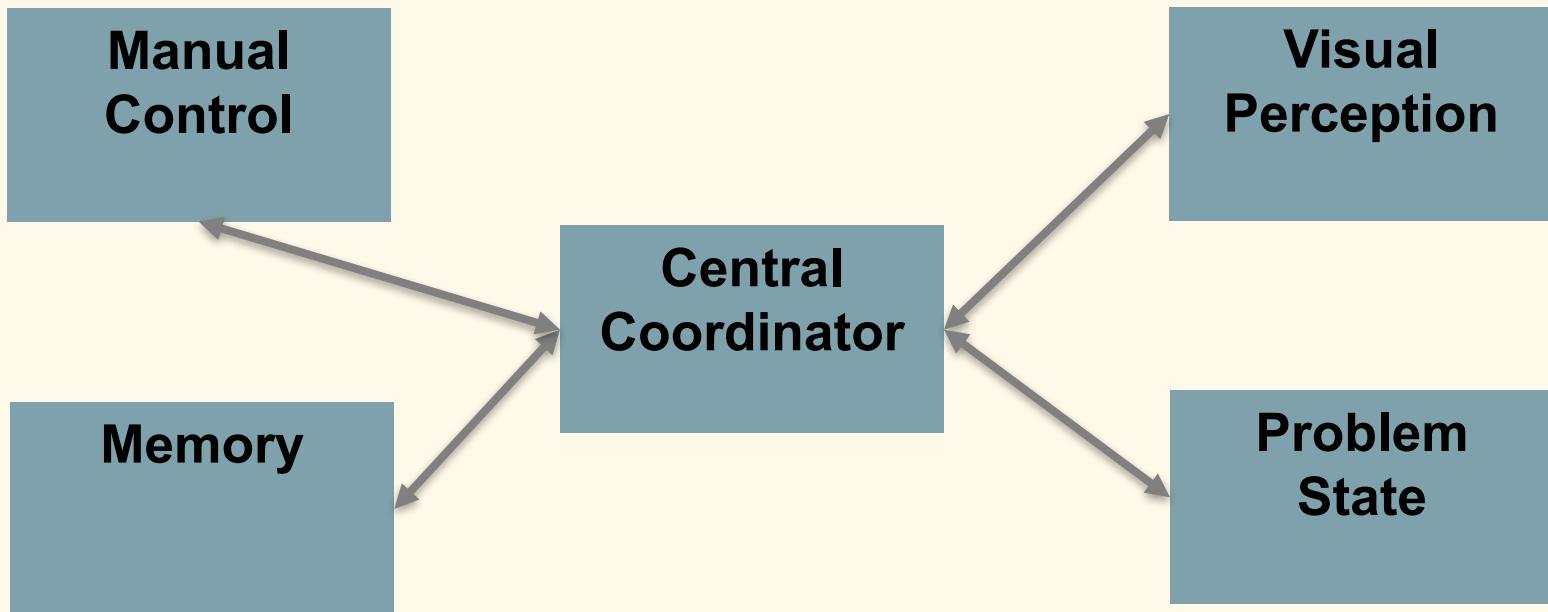
**Executed *functions*:**

- Look at the phone for address (vision)
- Look at sat-nav (vision)
- Store address in memory (memory)
- Type address (motor)
- Remember where you are (problem state)
- Execute actions in the right order (central coordinator)



# Type in address

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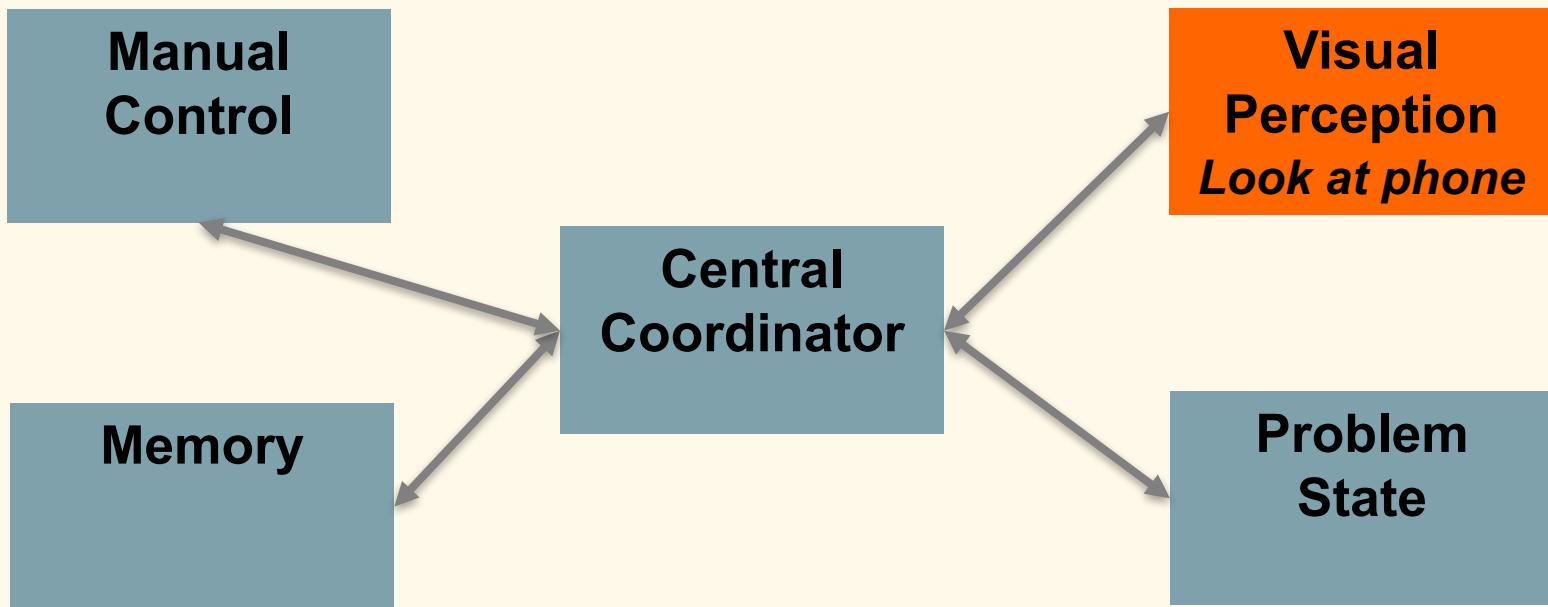
# Type in address

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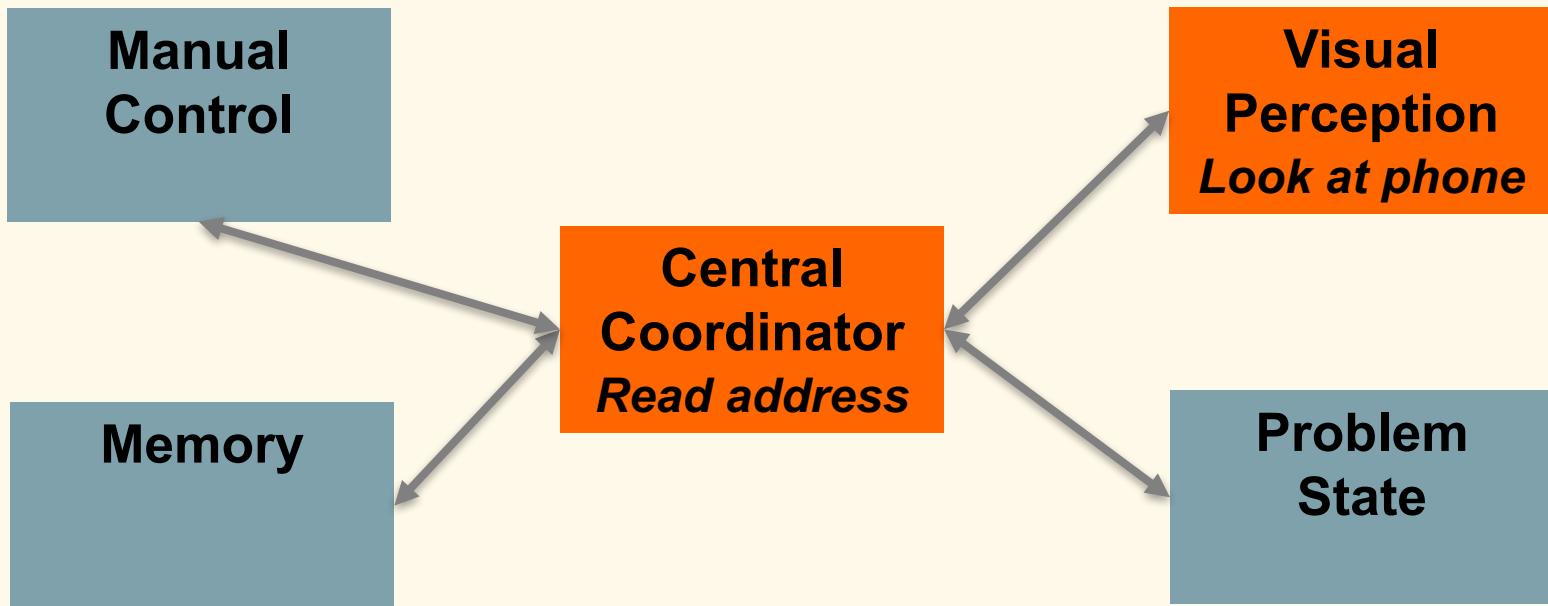
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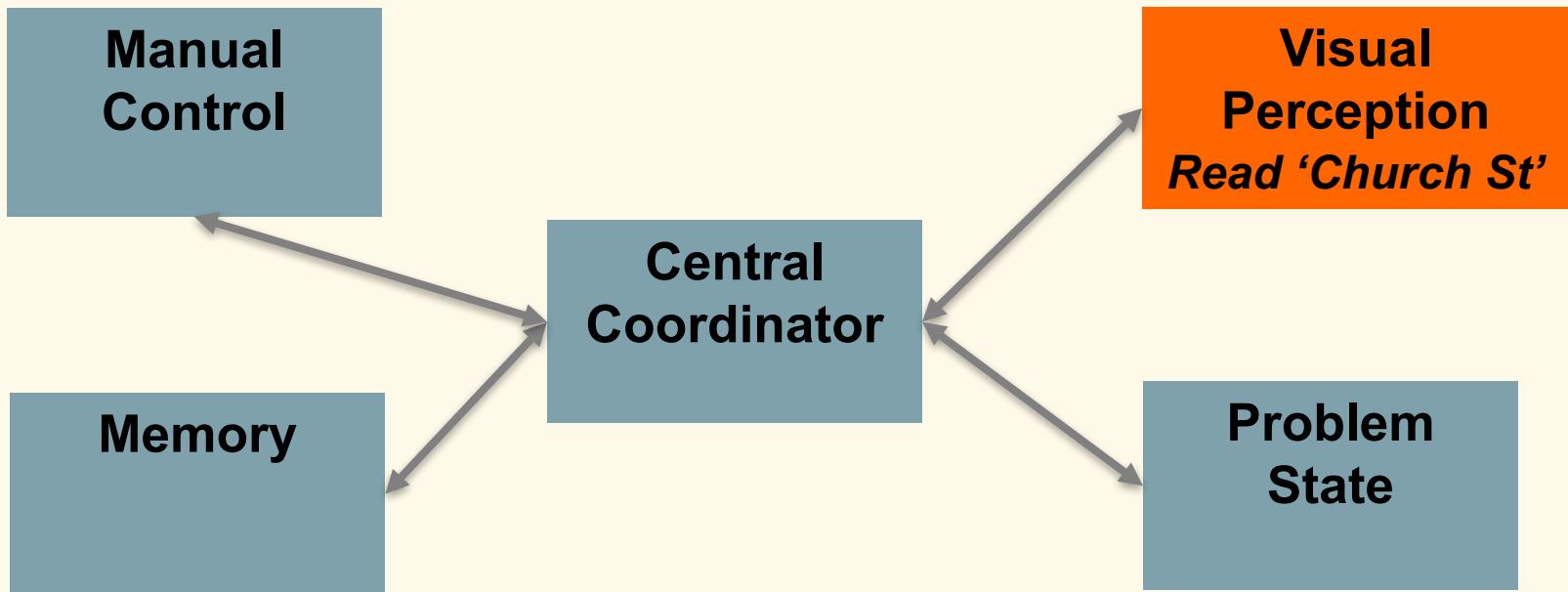
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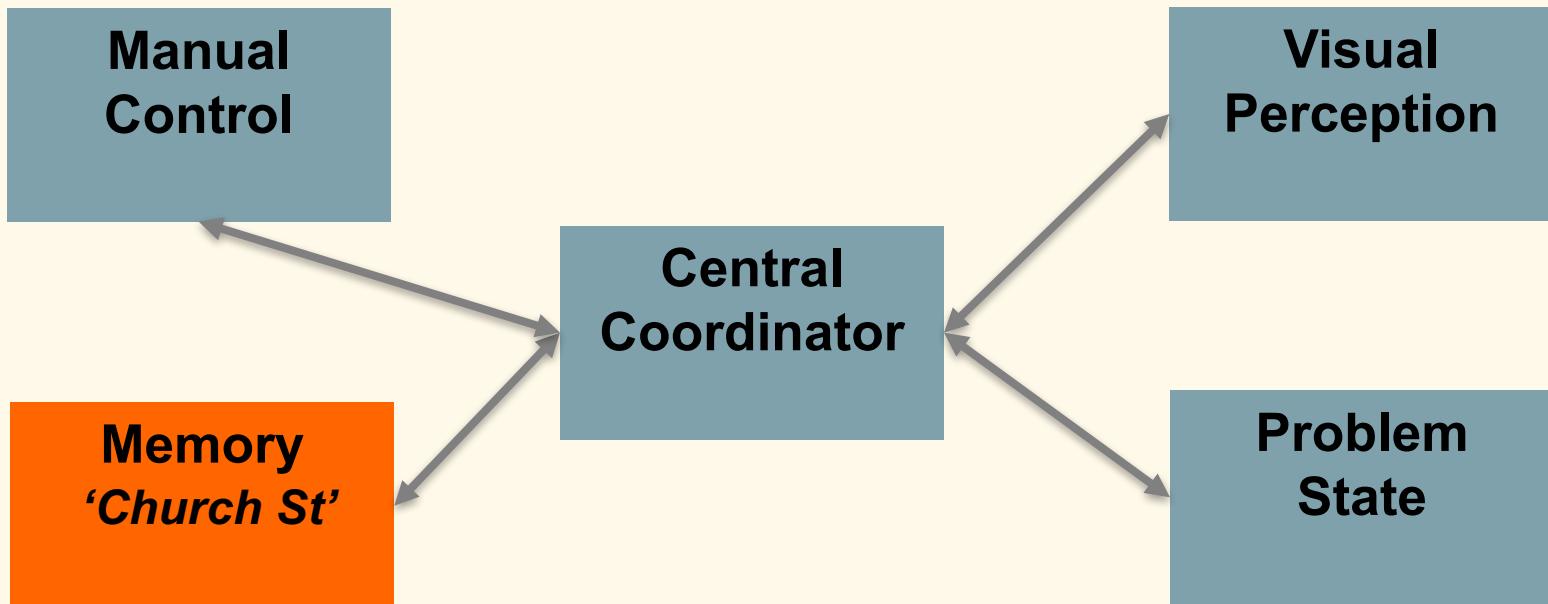
# Type in address

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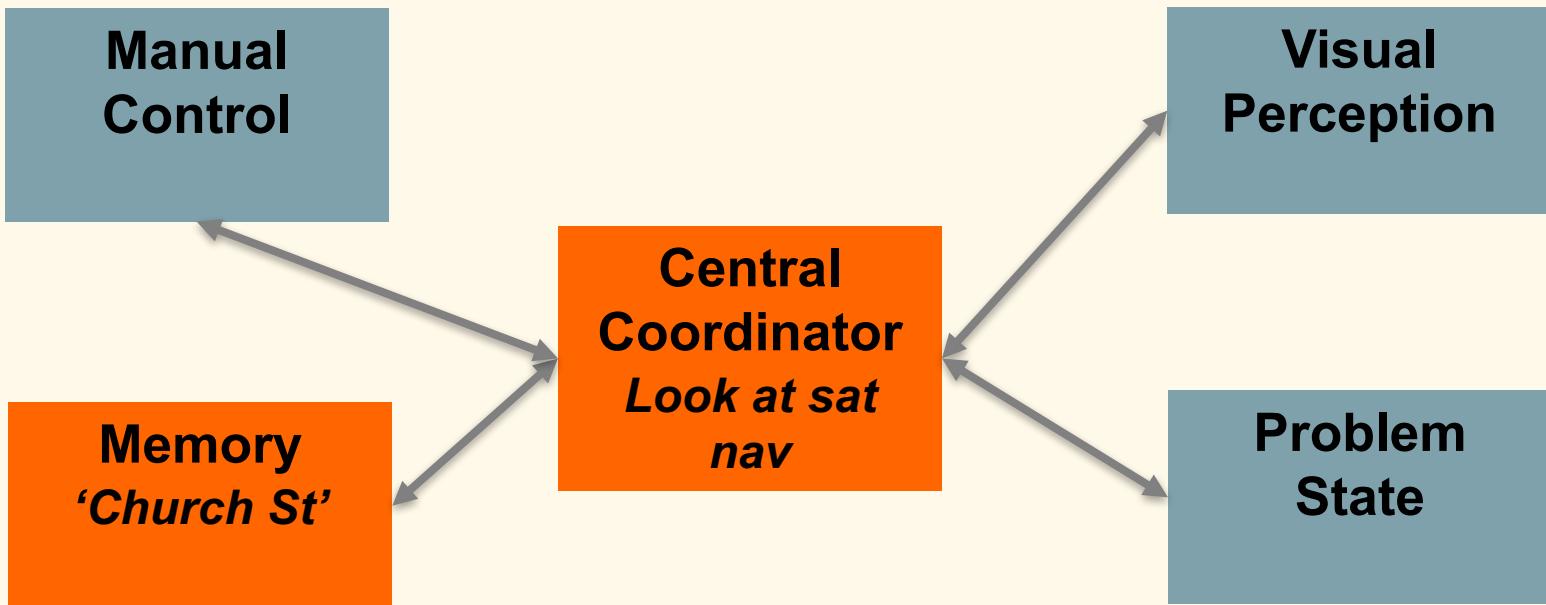
# Type in address

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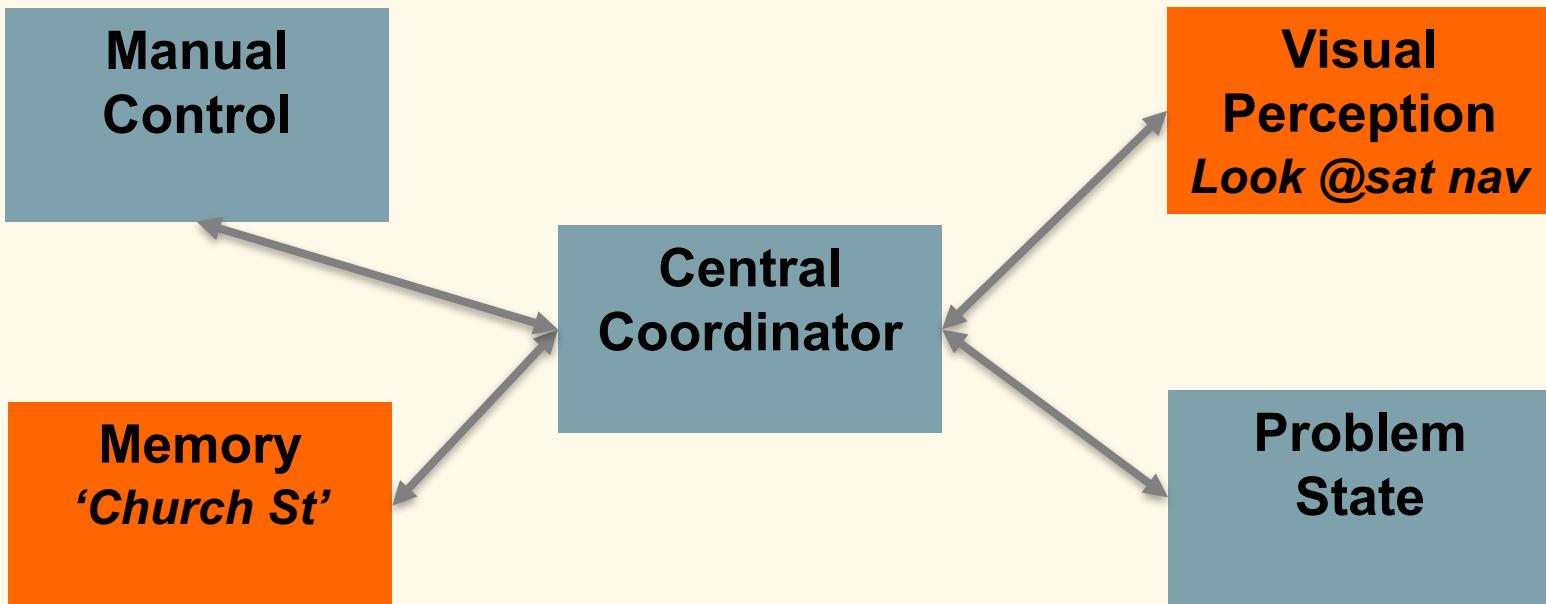
# Type in address

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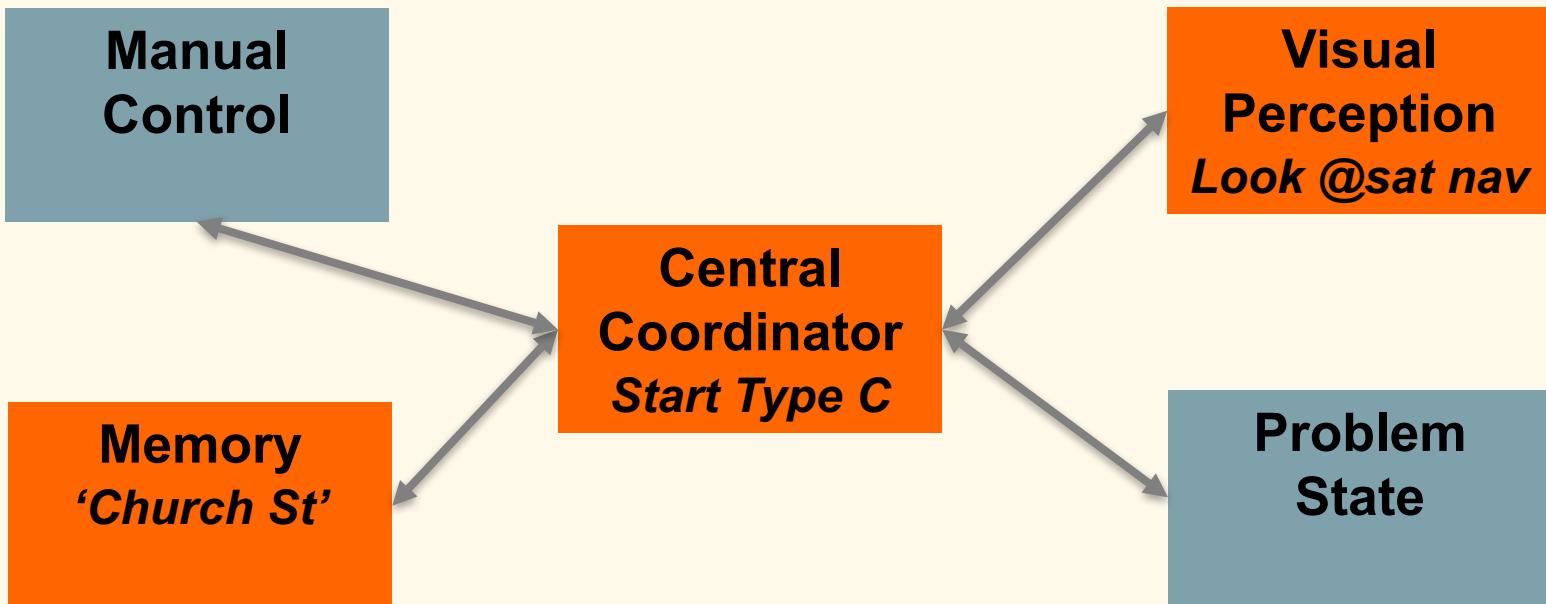
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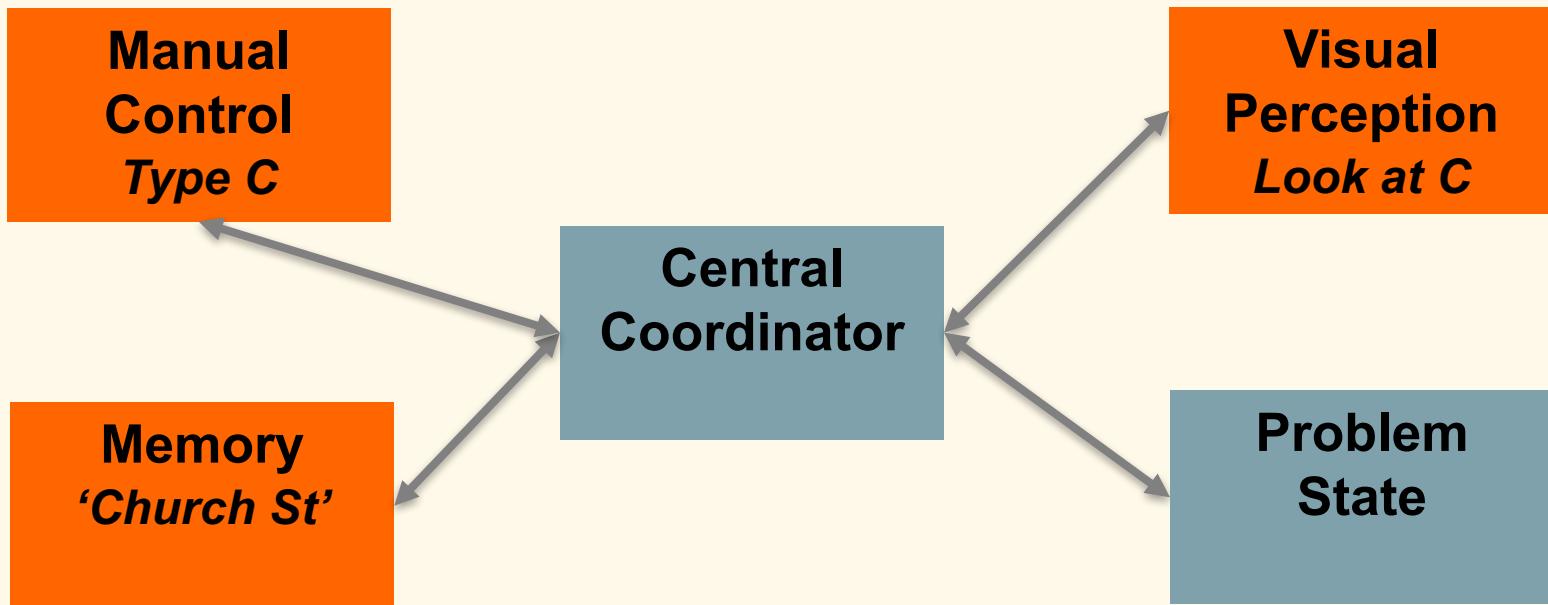
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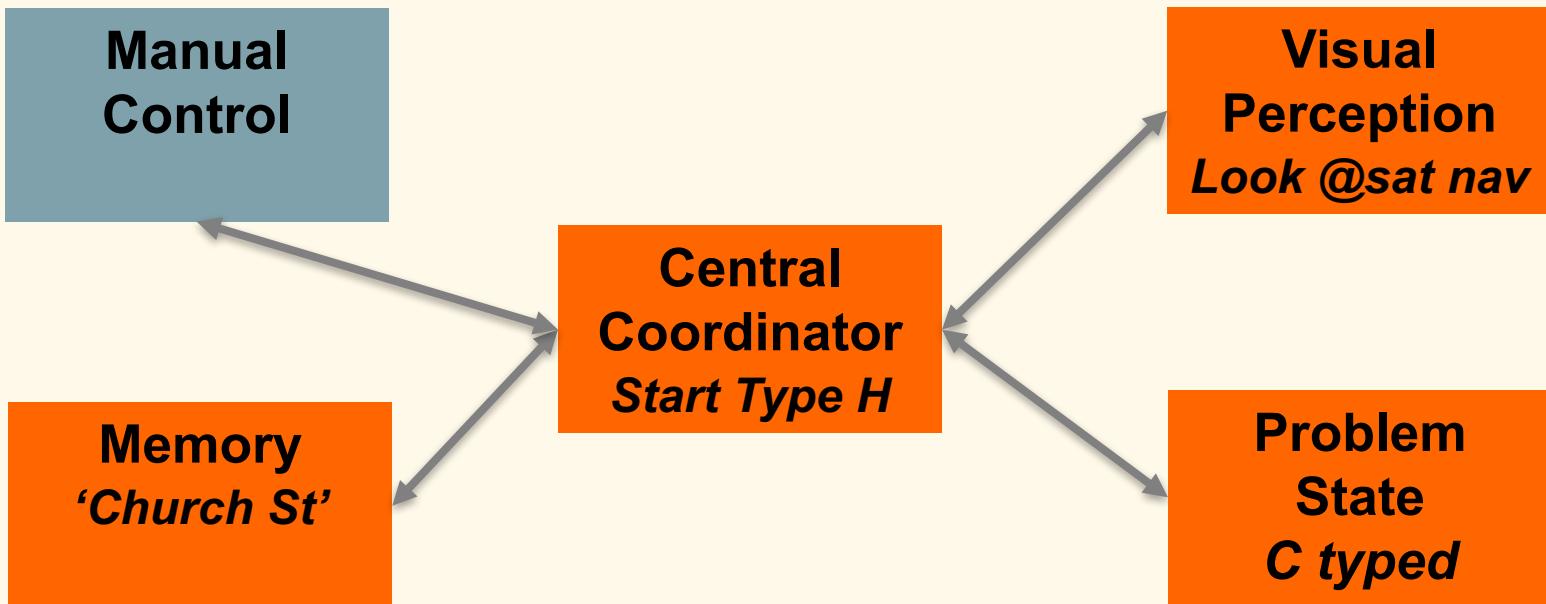
# Type in address

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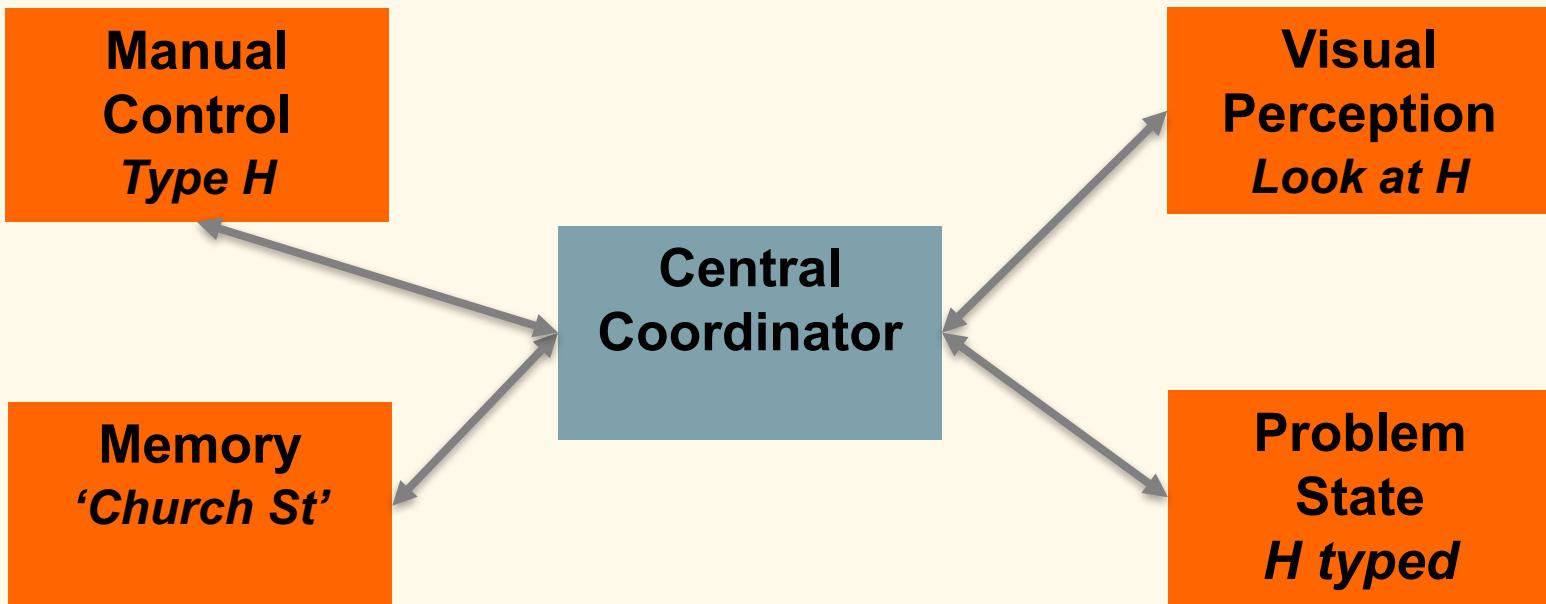
# Type in address

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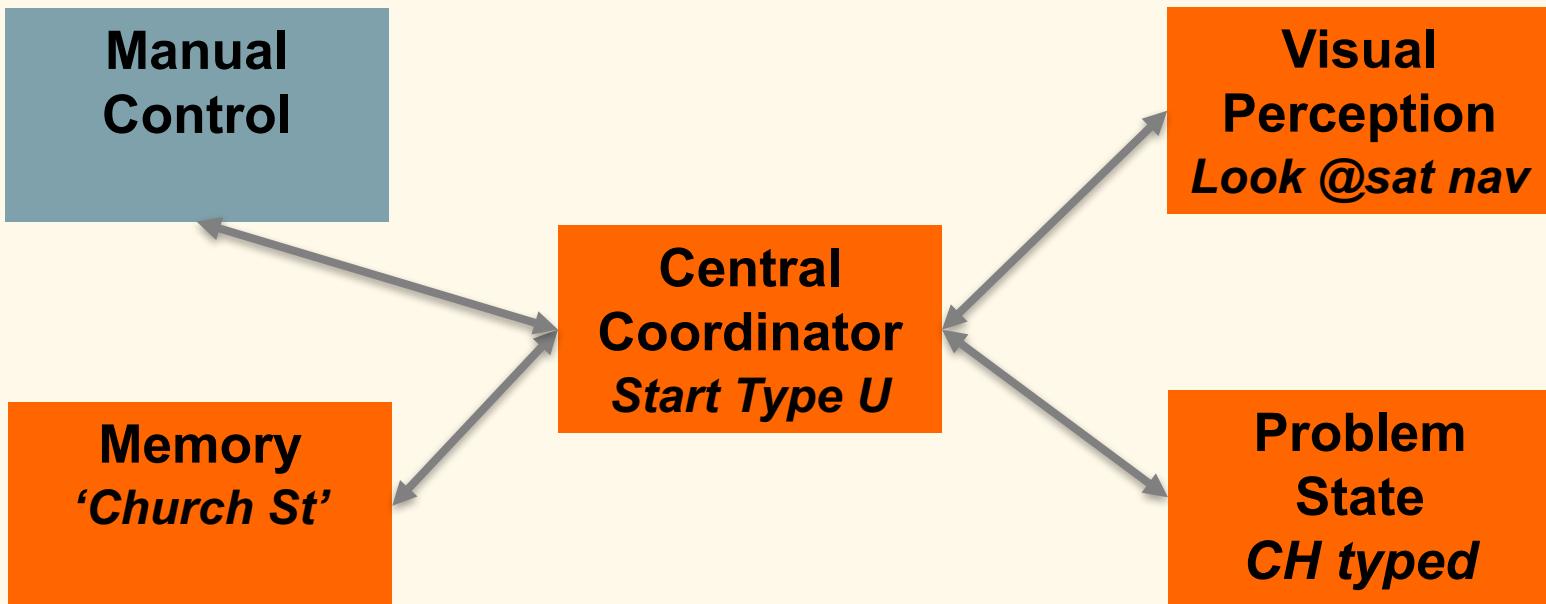
# Type in address

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# Type in address

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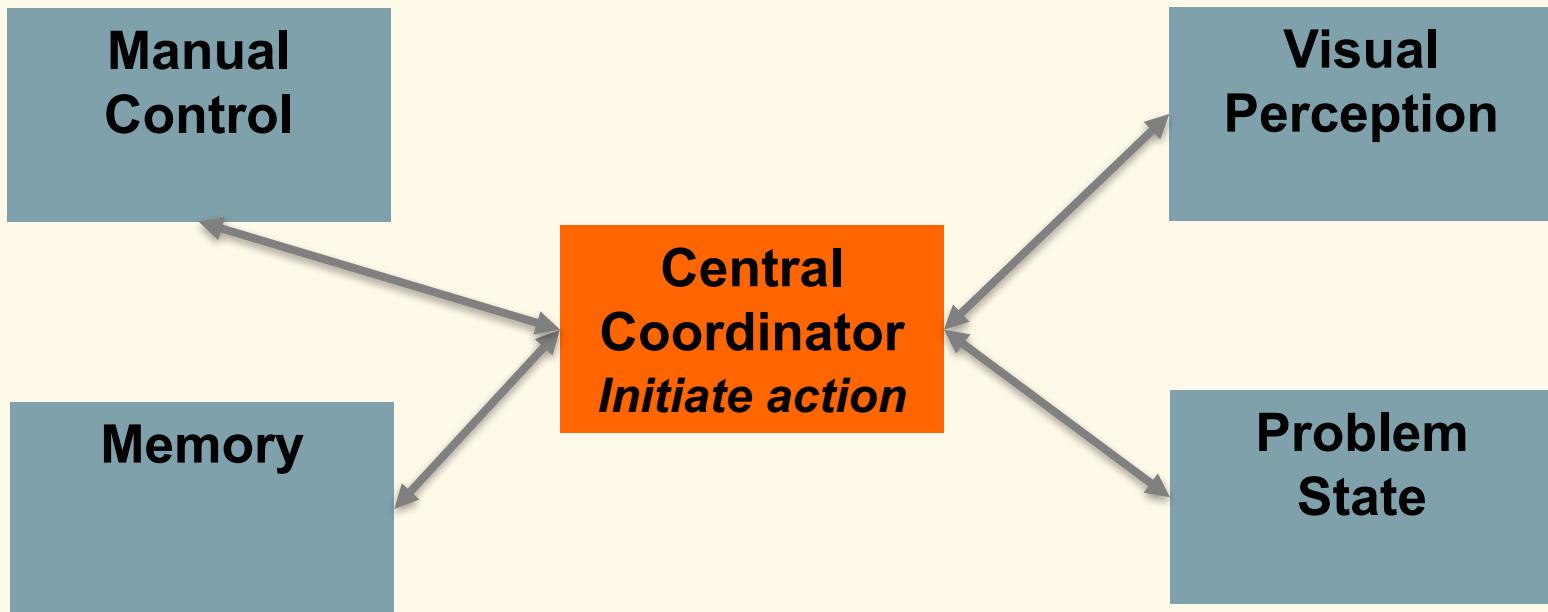
# Example 2: Drive simulated car

- **Executed *functions*:**
  - Look at road (vision)
  - Identify center of road (vision)
  - Determine distance to center (“Black box”)
  - Use steering wheel to improve lateral position (manual)



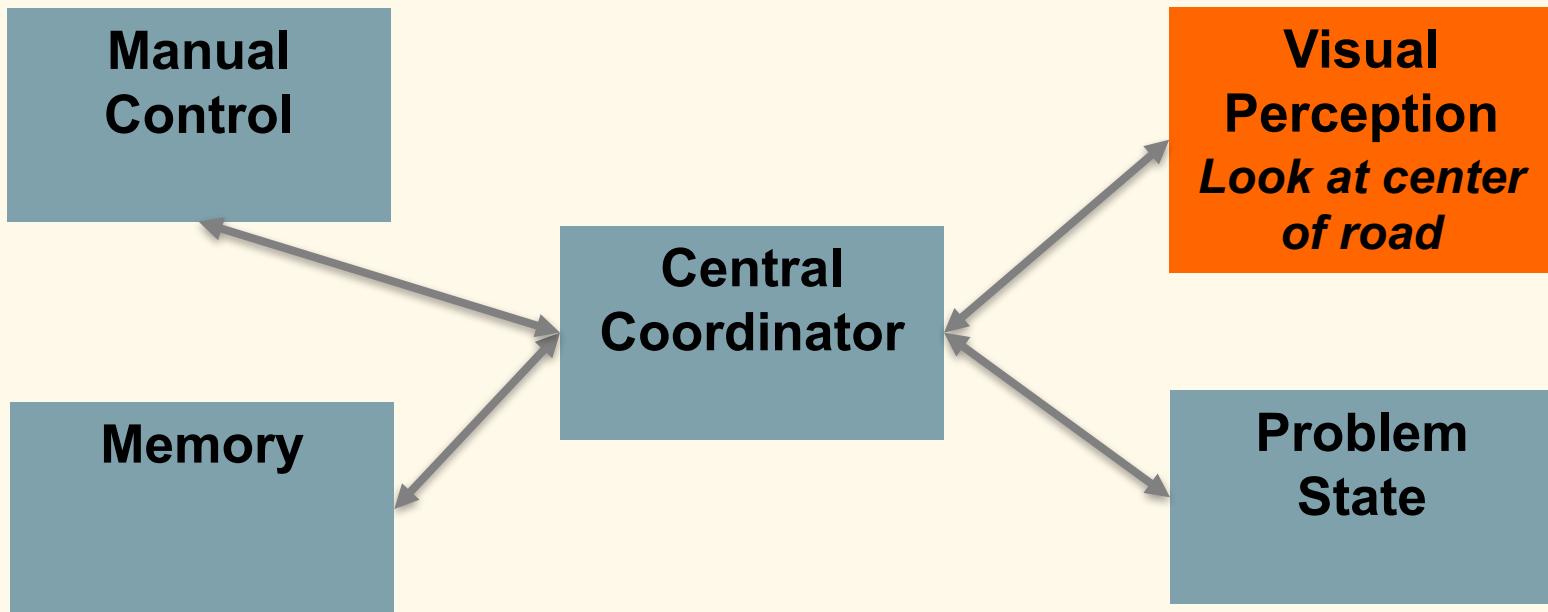
# Simulated driving

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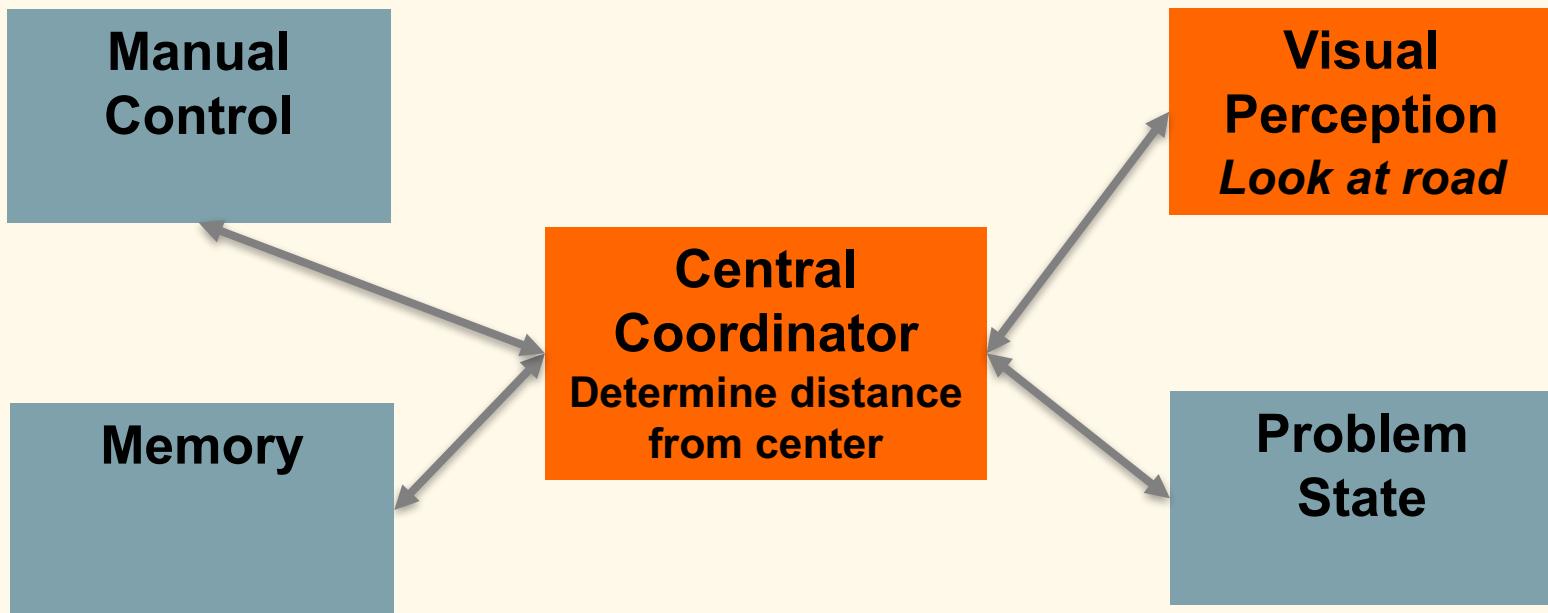
# Simulated driving

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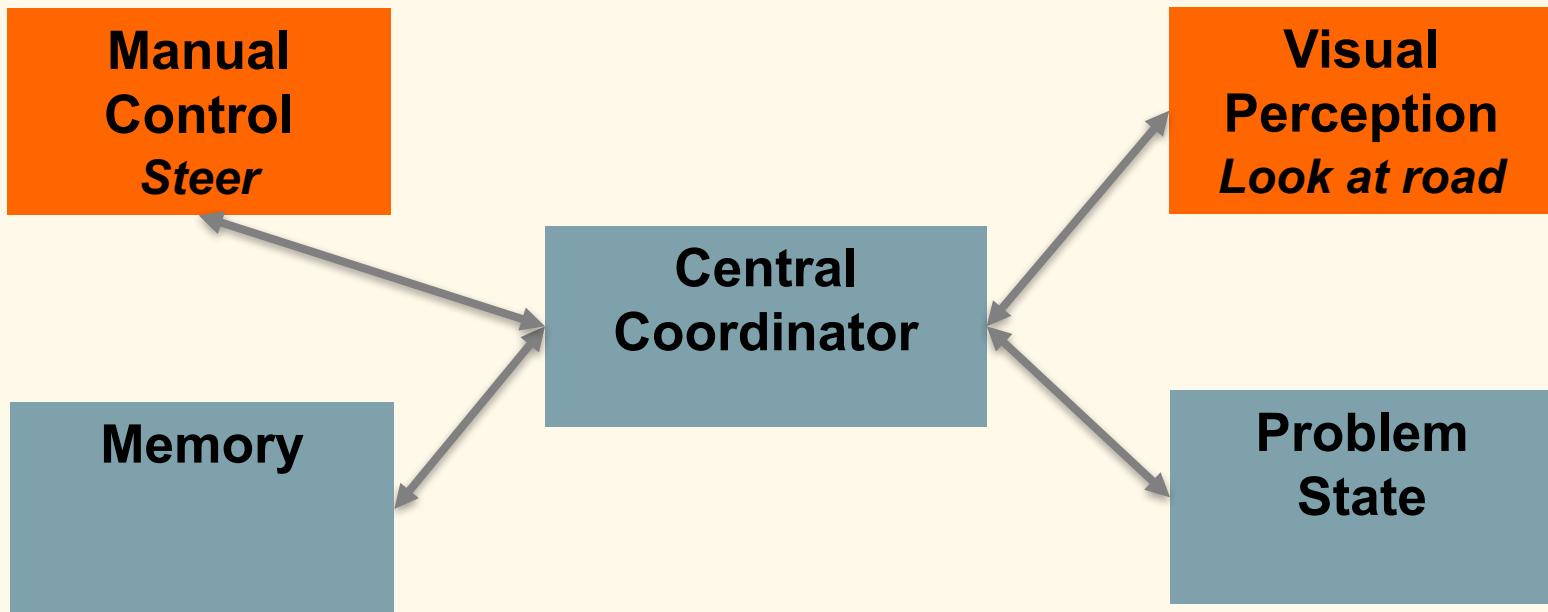
# Simulated driving

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# Simulated driving

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# Multitasking

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- 2 or more tasks “concurrently”  
(or in rapid succession)

