

# Introduction

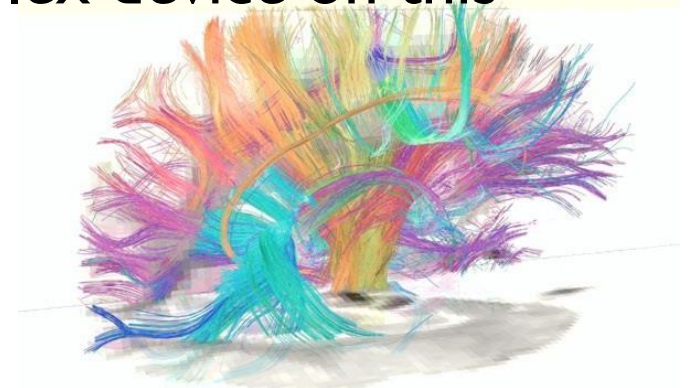
Computational Approaches to Neuroscience (NSCI 850)

Gunnar Blohm

# Some facts

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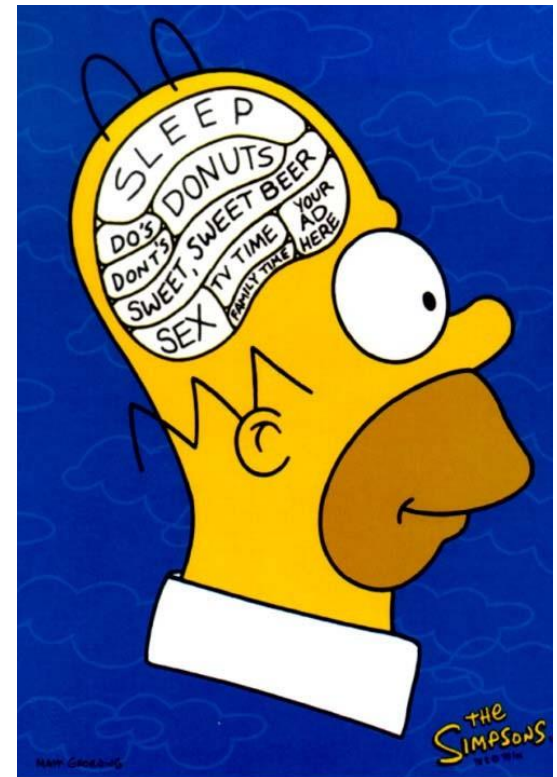
- ▶ The human brain = the most complex device on this planet
  - ▶  $10^{10}$  neurons
  - ▶  $10^{14}$  synapses
  - ▶ 100,000 miles of dendrites
  - ▶ # different potential pathways: | followed by 7 million miles of typewritten 0s
  - ▶ Our brain is capable of having more ideas than there is atoms in the universe
- ▶ Question:
  - ▶ What are the computational principles of this device that produce our cognitive and behavioural abilities?



# Outline

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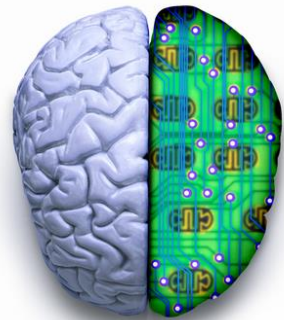
- ▶ What's computational neuroscience?
- ▶ Goals
- ▶ Why modeling of the brain?
  
- ▶ Course content
- ▶ Deliverables & evaluation
- ▶ Organization



# What is computational neuroscience?

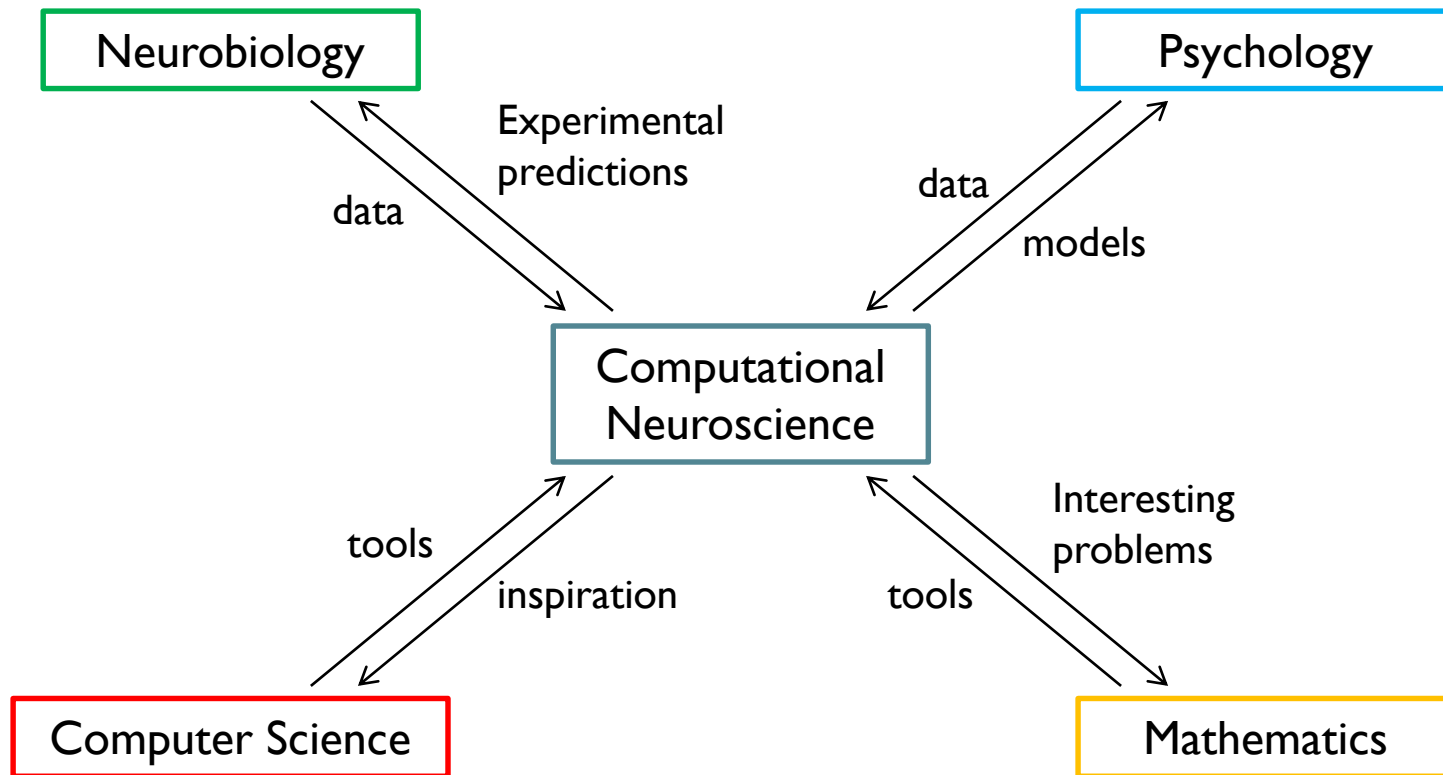
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- ▶ It is the use of mathematical, engineering and computer science tools to attempt answering neuroscience questions.
- ▶ Builds mathematical models describing computations in the brain that give rise to mental abilities
- ▶ A way to put “word models” into mathematical language and to analyze the result
  - ▶ Identify hidden assumptions
  - ▶ Explain observations
  - ▶ Make predictions
- ▶ Interdisciplinary: neuroscience, cognitive science, psychology, electrical engineering, computer science, physics, mathematics, life science, health science, medicine, etc...



# Computational vs. Neuroscience

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# Goals

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- ▶ Gain complete understanding of some experimental phenomenon
- ▶ Identify hypotheses, assumptions, unknowns
- ▶ Make quantitative predictions
- ▶ Build a theoretical brain as a model of the real brain (stroke lesions etc)
- ▶ Inspire new technologies
- ▶ Models of neurological diseases to help treatment, rehabilitation, quality of life
- ▶ Guidance in designing useful experiments (i.e. animal research)

# Why model brain function?

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- ▶ Models help answering three potential types of questions about the brain (Dayan & Abbott, 2001)
  - ▶ Descriptive = What?
    - ▶ Compact summary of large amounts of data
  - ▶ Mechanistic = How?
    - ▶ Show how neural circuits perform complex function
  - ▶ Interpretive = Why?
    - ▶ Computations in the brain are usually performed in an optimal or nearly optimal way
    - ▶ Understanding optimal algorithms and their implementation to explain why the brain is designed the way it is

# Why model brain function?

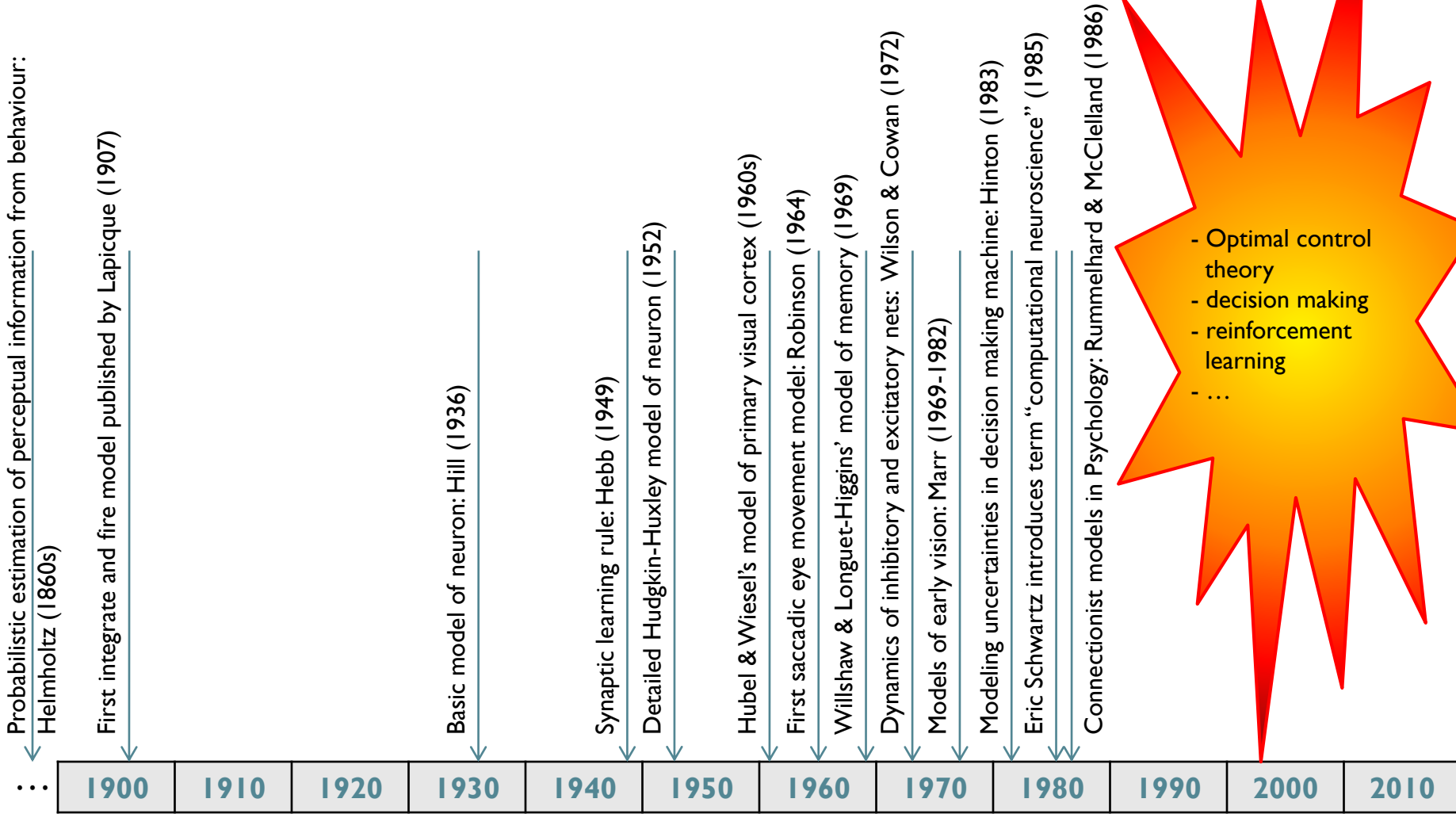
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- ▶ To inspire new technologies
  - ▶ Nature has always inspired technological advances
    - ▶ E.g. Leonardo Da Vinci
  - ▶ Artificial intelligence
    - ▶ Neural networks inspired machine learning (1984)
    - ▶ Learning algorithms are often not biologically plausible
    - ▶ Many open questions remain (e.g. language, autonomy, creativity)
  - ▶ Transferable skills from computational neuroscience
    - ▶ Mathematical formulation and simulation of complex dynamical systems
    - ▶ Understanding and analysis of dynamical systems
    - ▶ Signal processing



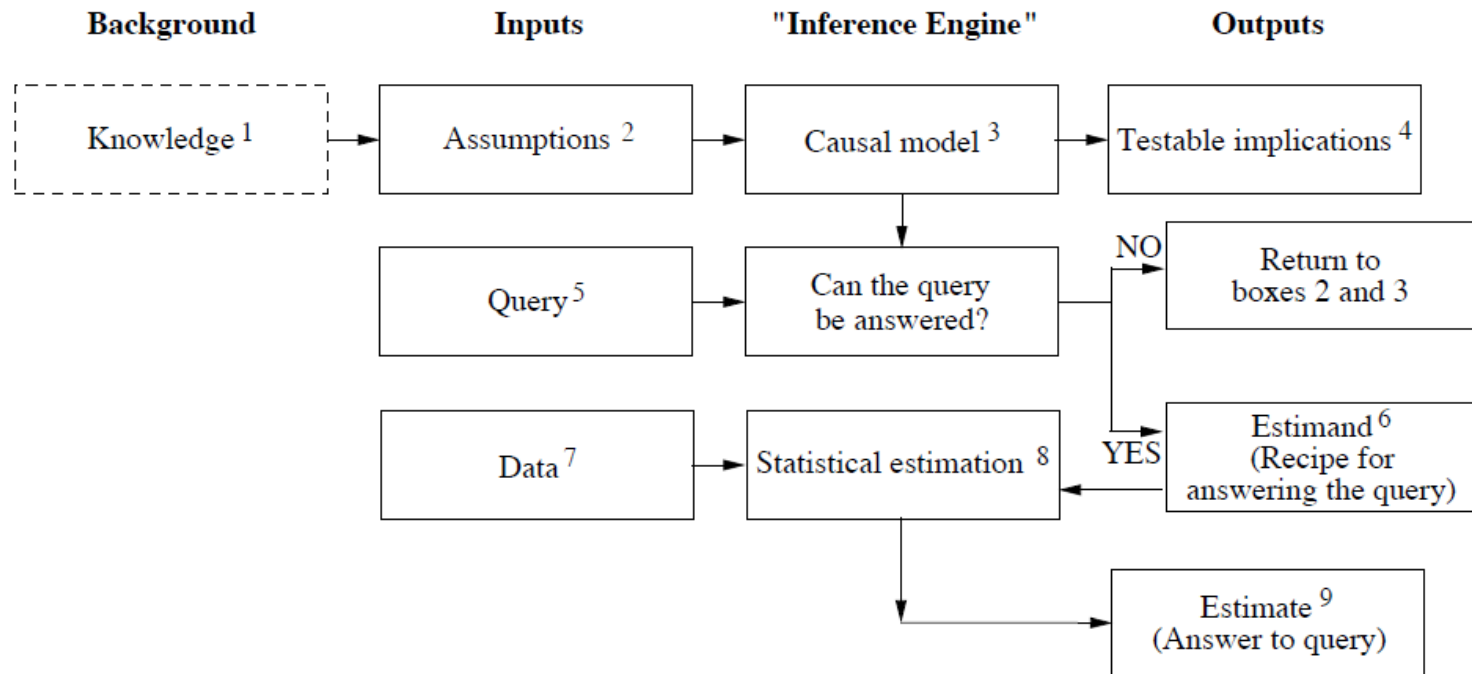


# History of the field



# Models vs. AI

- ▶ Data alone cannot lead to generalization
- ▶ Models are needed for that!



Pearl & Mackenzie, "The book of why", 2018

# Course organization

# Course info

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- ▶ <http://www.compneurosci.com/NSCI850.html>
- ▶ Access to documents on web site:
  - ▶ Username: NSCI
  - ▶ Password: NNet



# Computational anatomy of the brain

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## ▶ Learning approach

### ▶ Low-level approaches

- ▶ Spiking low-level approach: class 2-3
- ▶ Intermediate-scale rate-based networks: class 4-5
- ▶ High-levels systems approach: class 6

### ▶ Transcendental approaches

- ▶ The Bayesian brain: class 7
- ▶ Optimal control theory: class 8
- ▶ Reinforcement learning: class 9

### ▶ Specific models: class 10-11

# Organization

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- ▶ **Specific models**
  - ▶ Goal: hands-on modelling experience
  - ▶ Approach:
    - ▶ Chose topic of interest
    - ▶ Develop your own model:
      - Model background (physiology, experimental finding etc)
      - Model equations
      - Model implementation
      - Model simulations
      - Model discussion
    - ▶ OR reproduce existing model:
      - Model background
      - Model implementation & simulations (reproduce results)
      - **New** model results / predictions / analyses / etc
      - Critical model discussion & suggestions for improvements + implementation of at least one improvement
  - ▶ Product: 30-45 min lecture on model (including questions)

# Organization

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- ▶ Specific models
  - ▶ Make a choice before Mar 4
- ▶ Suggestions:
  - ▶ Visual system (early vision, feature/color selection, stereo vision...)
  - ▶ Eye movements (saccades, pursuit, eye-head coordination, VOR, deficits, vergence...)
  - ▶ Decision making
  - ▶ Disorders (Alzheimer, Schizophrenia, Parkinson, epilepsy...)
  - ▶ Attention
  - ▶ Motor control (muscles, feedback...)
  - ▶ Multi-sensory integration
  - ▶ Central pattern generators (swimming, walking etc)
  - ▶ Reference frame transformations
  - ▶ Learning & memory
  - ▶ ...

# Further readings

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- ▶ Trappenberg: Fundamentals of Computational Neuroscience. Oxford University Press 2010
- ▶ Dayan & Abbott: Theoretical Neuroscience. MIT Press 2001
- ▶ <http://matrixcookbook.com/>
- ▶ Wilson: Spikes, Decisions and Actions. Oxford University Press 1999
- ▶ Izhikevich: Dynamical systems in Neuroscience. MIT Press 2006
- ▶ Wallisch et al.: Matlab for Neuroscientists. Academic Press 2008
- ▶ Gerstner & Kistler: Spiking neuron models. Cambridge University Press 2002
- ▶ Gupta et al.: Static and dynamic neural networks. Wiley-Interscience 2003
- ▶ Shadmehr & Wise: The computational neurobiology of reaching and pointing. MIT Press 2005

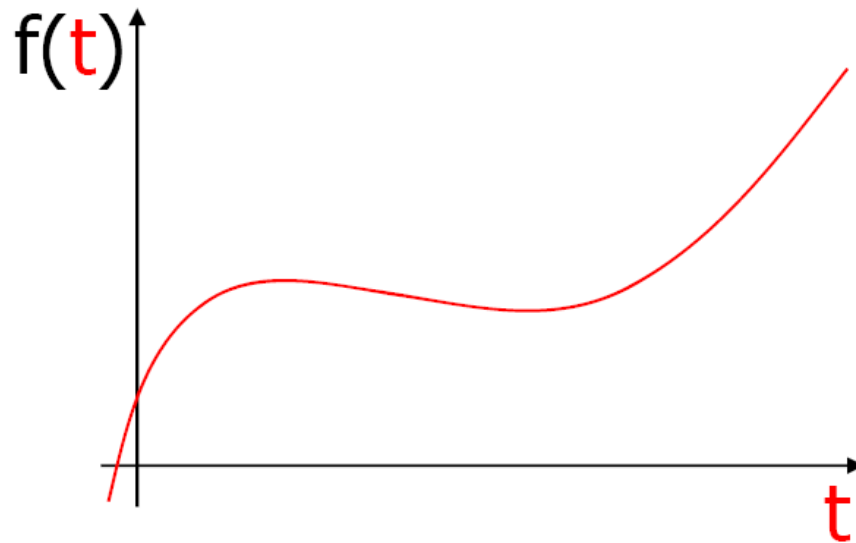


# Differential calculus and ODE primer

# Derivatives

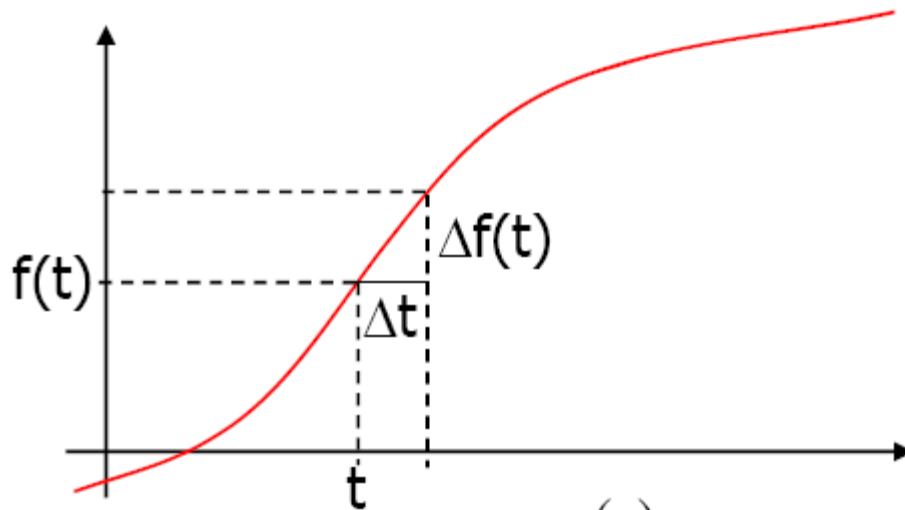
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- ▶ Consider a function  $f$  (e.g. neuronal activity) that changes over time
- ▶  $f = f(t)$



# Derivatives

- ▶ Definition of derivative



$$\frac{df(t)}{dt} = \lim_{\Delta t \rightarrow 0} \frac{\Delta f(t)}{\Delta t}$$

- ▶ Other notations  $\frac{df(t)}{dt} = f'(t) = \dot{f}$

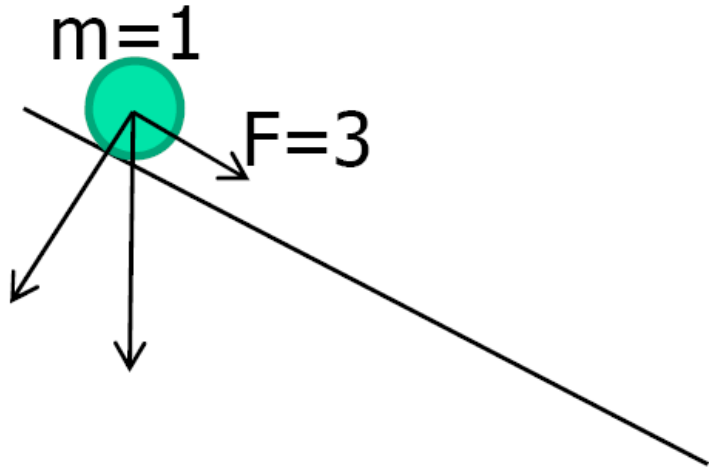
- ▶ Sign of derivative

- ▶ Positive: function increases
- ▶ Negative: function decreases

# Example

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- ▶ How does the velocity of the ball change over time?



- ▶ Newton's law:  $F = m \cdot a$
- ▶ From the definition of acceleration:  
 $a = dv/dt = 3$

# Example cont'd

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- ▶  $dv/dt = 3$  is a differential equation!
- ▶ Differential equations describe how the derivative of a function behaves
- ▶ It says that function  $v(t)$  is a function whose derivative is equal to 3 at all times
- ▶ What is the function that satisfies this equation?
- ▶ Is there just one such function or are there many of them?
- ▶ There are an infinity of solutions:  $v(t) = 3*t + c$
- ▶ Importance of initial conditions

# Another common example

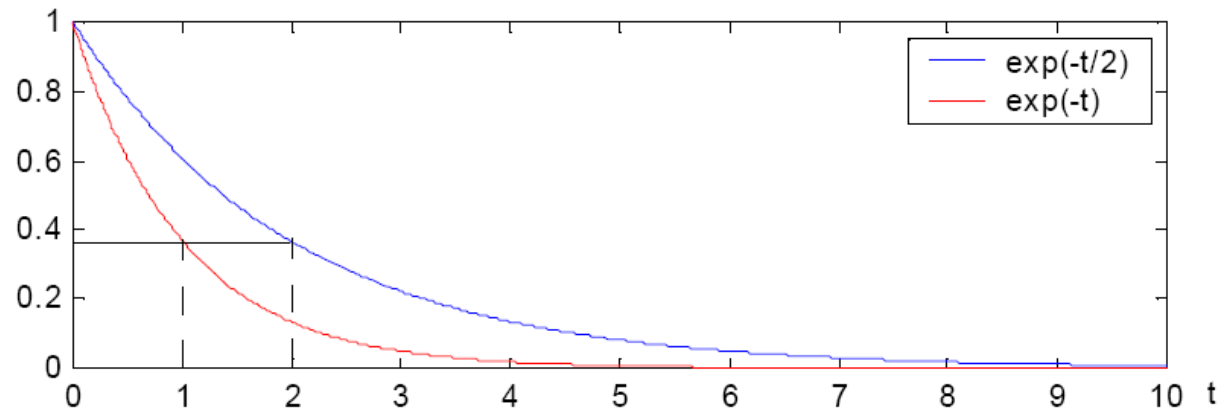
- ▶ Let us consider the following ordinary differential equation (ODE)

$$\tau \frac{df(t)}{dt} = -f(t)$$

- ▶ What is the solution? Is it unique?

- ▶ Solution:  $f(t) = ce^{-t/\tau}$

- ▶ Parameter tau?





The end

***Course material:*** [www.compneurosci.com/NSCI850.html](http://www.compneurosci.com/NSCI850.html)