

## Explanations in Neuroscience 2

### Explaining Without Mechanisms?

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## Deductive Nomological Explanations Again

- Based in part on examples from physics, proponents of the D-N model viewed laws as the critical feature of an explanation
- From Galileo's law of free fall, explain why an object fell 64 feet in 2 second
$$d = 1/2 a t^2$$
$$t = 2$$
$$a = 32$$
$$\therefore d = 64$$
- $d = 1/2 a t^2$  is a law that explains why the object falls the distance it does
- The function of laws in D-N explanations can be generalized to equations that describe a domain of phenomena

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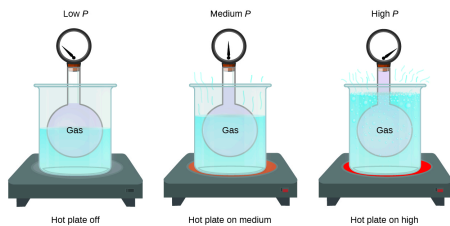
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## Laws and Dynamics

- Some laws/equations characterize simple and easily intelligible relations between variables
- In a gas, temperature = volume x pressure



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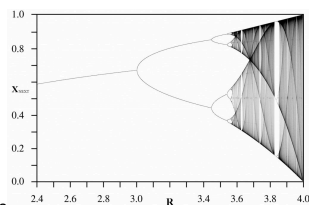
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## Some Equations Generate Surprising Behavior

- Some laws are deceptive:  $x_{t+1} = Ax_t(1-x_t)$ 
  - for values of A less than 3, successive iterations will ultimately approach a constant value
  - A little above A=3.4, the system will stabilize to an oscillation between two values
  - Around A=3.5, it stabilizes in an oscillation between 4 values
  - In some intervals, the system will never stabilize but become chaotic



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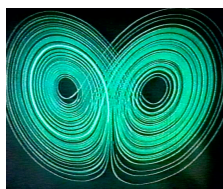
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## Complex Behavior from Simple Equations

- Some systems, even relatively simple ones, exhibit very complex trajectories through state space

$$\begin{aligned}\frac{dx}{dt} &= \sigma(y - x), \\ \frac{dy}{dt} &= x(\rho - z) - y, \\ \frac{dz}{dt} &= xy - \beta z.\end{aligned}$$



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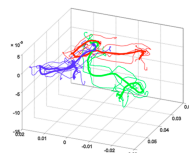
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## Dynamical Explanation

- A set of differential equations specifies how variables characterized in the equations will change over time
- One can use such a set of equations to model a system
  - And represent the behavior of the system as a trajectory through a state space which has a dimension for every variable
  - Time appears not as a variable but as a succession of points



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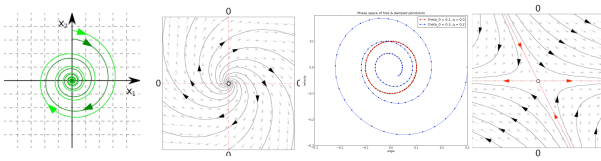
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# Attractors

- Many dynamical systems exhibit an attractor structure
  - Starting the system from different values of variables one can trace the resulting trajectories
    - Sometimes all will converge (point attractor)
    - Other times they will diverge (point repeller)
    - Sometimes they will converge to a circle (cyclic attractor)
    - There may be multiple basins of attraction separated by a separatrix




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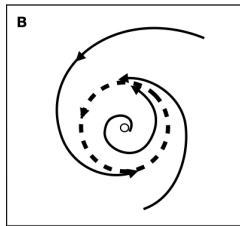
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## Circadian Rhythms and Cyclic Attractors

- Hypothesis: rhythm results from a protein inhibiting the transcription of its own gene
  - As more of the protein is synthesized, the more the transcription is inhibited until it stops
  - As the protein degrades, letting transcription begin again




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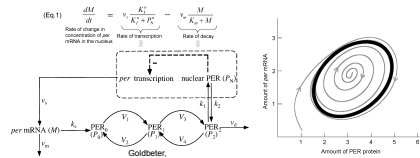
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## Recomposing Using Computational Models

- Should one trust one's intuitions?
  - Will a feedback mechanism generate sustained oscillations?
- Goldbeter (1995) created a computational model that showed that with biological plausible parameters, it could generate sustained oscillations




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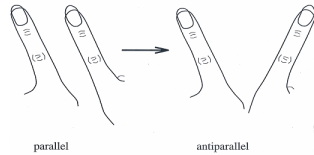
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## Explaining in Terms of Attractors without Mechanisms

- Bilateral animals move their limbs in coordinated ways, but speeding up can alter the relation (e.g., from walking to running)
- A simple illustration due to Scott Kelso and described by Chemero
  - Move your index fingers in parallel
  - Start slowly and gradually speed up
  - At some point you will no longer be able to maintain the parallel movement and will lapse into antiparallel movement




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## Clicker Question

What, according to Chemero, explains why parallel coordination cannot be sustained at higher frequencies?

- The behavior of the central pattern generator in the brain
- The fact that an attractor disappeared at higher frequencies
- The person's failure to try hard enough to maintain parallel coordination
- None of the above

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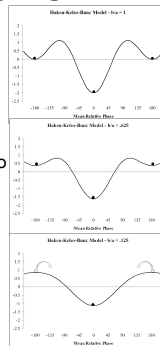
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## Dynamical Explanations without Mechanisms

- What explains the change?
  - The phase between two limbs is described by the Haken-Kelso-Bunz equation
    - $V(\phi) = -a \cos\phi - b \cos 2\phi$ ,
  - where  $V$  is change in relative phase,  $\phi$  represents the relative phase and the ratio of the parameters  $b/a$  is inversely related to the rate
    - When  $b/a = a$ , there are two relatively deep attractors but
    - As  $b/a$  declines, a point is reached at which there is only one attractor




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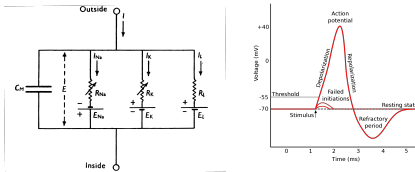
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## Does an Equation Explain?

- Recall the Hodgkin-Huxley equation describing the action potential

$$I = C_m \frac{dV_m}{dt} + \bar{g}_K n^4 (V_m - V_K) + \bar{g}_{Na} m^3 h (V_m - V_{Na}) + \bar{g}_l (V_m - V_l),$$

- The equation provides a good description of how voltage changes over time
  - From it you can derive the graph of the action potential




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## Does an Equation Explain?

- The Hodgkin-Huxley equation describes the action potential
  - But does it explain it?
- Craver: No. The equation represents curve fitting (by Huxley's own account)
  - what do  $n$ ,  $m$ , and  $h$  represent?
  - Why are they raised to particular powers?
    - Only once these were connected to ion channels and gates on them was the action potential explained
- Levy: Yes. What Hodgkin and Huxley did was identify how the different currents together generated the action potential
  - The channels and gates belong to a yet lower level

$$I = C_m \frac{dV_m}{dt} + \bar{g}_K n^4 (V_m - V_K) + \bar{g}_{Na} m^3 h (V_m - V_{Na}) + \bar{g}_l (V_m - V_l),$$

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## Is A Mechanism Required for Explanation?

- Craver and Levy agree that an equation can serve to explain
  - But only if its terms correspond to components of a mechanism
    - They disagree about which components actually do this work in the case of the Hodgkin-Huxley equation
- Chemero disagrees
  - Citing the Haken-Kelso-Bunz equation suffices
    - From it we can show when one attractor disappears and the corresponding form of coordination is no longer possible

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## Discussion Question

Does a dynamical equation that describes a phenomenon sufficiently accurately to predict it under varying conditions suffice to explain that phenomenon?

- A. Yes. With the equation one understands why the phenomenon occurs as it does.
- B. Yes, if the equation reveals complexity in the phenomenon itself that accounts for features of the phenomenon.
- C. Yes, if the terms of the equation can be related to the components of the mechanism that generates the phenomenon.
- D. No. The equation just describes the phenomenon. It doesn't explain it.

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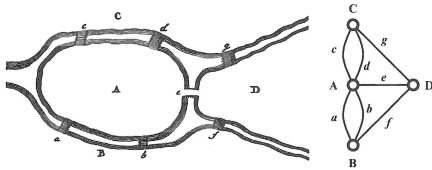
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## Network Explanations

- In the 17th century Leonard Euler posed a problem:
  - Could one find a route to cross all seven bridges of Königsberg each just once?



- Challenge: prove that it is not possible
  - For each node other than the first and last, there must be an even number of bridges. Why?

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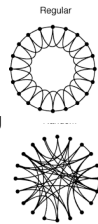
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## Understanding Massive Mechanisms

- Strategy: Appeal to properties of particular kinds of network to explain features of systems that instantiate them
- Most work in graph theory in the 20th century focused on regular lattices and random networks
  - Regular lattices exhibit high clustering but long characteristic path length
- The main alternative that was considered was random networks
  - Random networks exhibit short characteristic path length, but low clustering



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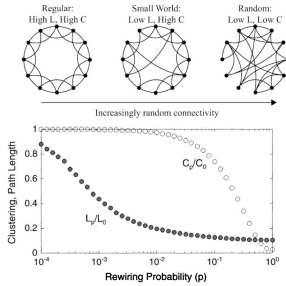
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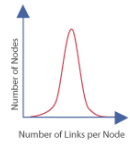
# Small-World Networks

- Watts and Strogatz (1998) showed that in between lattices and random networks lie networks with both short characteristic path length and high clustering
  - These are referred to as *small worlds*
- They argued that small worlds were efficient for information processing
  - Allows for group of nodes to specialize while still being connected to the whole



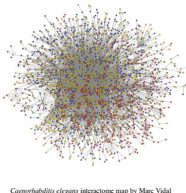
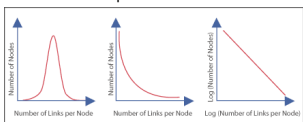
# Node Degree

- The number of connections from a node is known as its degree
- Most 20th century graph theory analyses assumed that node degree is distributed normally—i.e., Gaussian



# Small Worlds with Scale Free Distribution

- Barabási discovered that in many real world networks the distribution exhibits a power law relation—most nodes have few connections but a few nodes have many connections
  - They referred to these as *scale-free networks*
- Most nodes have few connections
  - Can be eliminated with minimal effect
- A few nodes have many connections
  - Eliminating them can have catastrophic effects



## Discussion Question

Both LAX and Carlsbad airports are shut down for one week. Why will shutting LAX affect airplane travel elsewhere in the world, but not Carlsbad?

- A. LA is a bigger city than Carlsbad
- B. LAX has more runways than Carlsbad
- C. The number of airports to which you can fly from LAX is greater than that for Carlsbad
- D. People are more likely to transfer between planes at LAX than at Carlsbad

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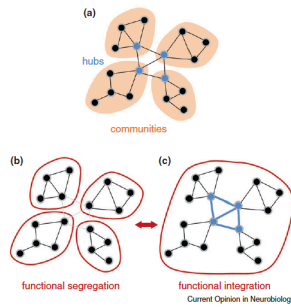
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## Interconnected Communities of Specialists

- Hubs
  - Nodes with an unusually large number of connections
    - within a local cluster (community)
    - with nodes in other clusters (communities)
- Hubs can create a network of specialists that still communicate



Sports, 2013

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## Applying the Network Approach to the Brain

- Define networks in terms of brain structure
  - Clustering neurons that are interconnected by axons and dendrites
- Define networks in terms of brain activity
  - Clustering neurons whose activity is correlated
- In these networks identify hubs and communities

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## Discussion Question

How do dynamical explanations and network explanations relate to mechanistic explanations

- A. They are competitors. May the best account win
- B. Both dynamical and network analyses provide ways to understand the organization of mechanisms
- C. Dynamical and network analyses are useful supplements to mechanistic accounts
- D. Other

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