Patterns of Synchrony From Animal Gaits to Binocular Rivalry

Gateways to Emergent Behavior in Science and Society ICAM/SFI Workshop September 24, 2013

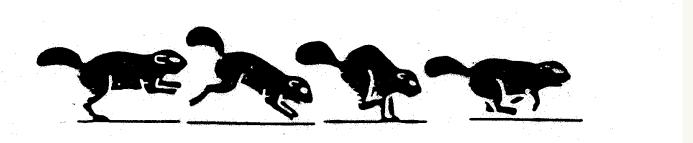
Marty Golubitsky

Mathematical Biosciences Institute
and

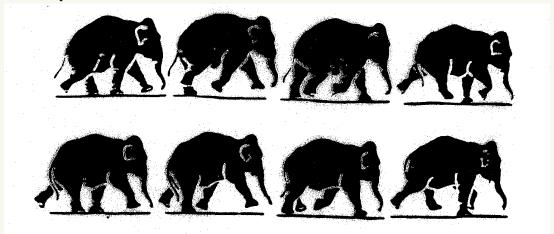
Department of Mathematics
Ohio State University

Quadruped Gaits

Bound of the Siberian Souslik

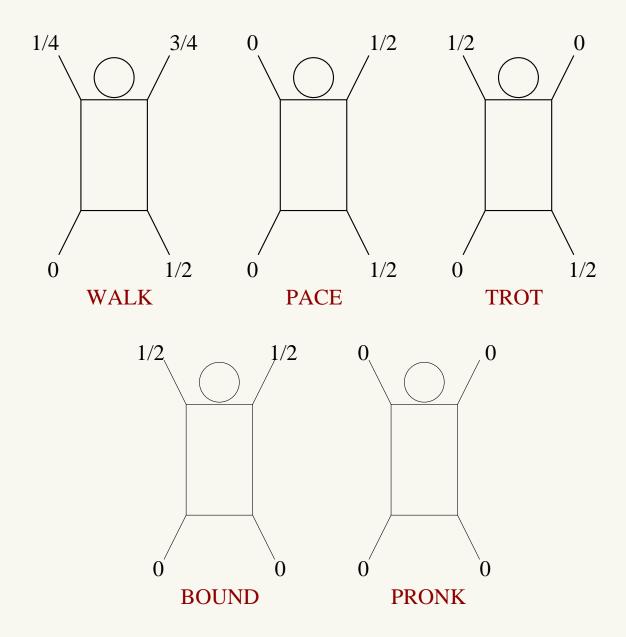


Amble of the Elephant



Trot of the Horse

Rigid Gait Phases for Quadruped Gaits



Gait Symmetries / Central Pattern Generators

Gait	Spatio-temporal symmetries			
Trot	(Left/Right, $\frac{1}{2}$)	and	(Front/Back, $\frac{1}{2}$)	
Pace	(Left/Right, $\frac{1}{2}$)	and	(Front/Back, 0)	
Walk	(Figure Eight, $\frac{1}{4}$)			

- Network of neurons (CPG) that produces gait rhythms
- Hodgkin Huxley (1952)
 Neuron modeled by system of differential equations
- Design simplest network that produces independent

walk, trot, pace

Collins and Stewart (1993); G., Stewart, Buono, and Collins (1999)

Rigid Phase-Shift Synchrony

- Symmetry: sends solutions to solutions ${f Z}_2$ symmetry
- Nodes oscillate in phase: $x_2(t) = x_1(t)$ Nodes half-period out of phase: $x_2(t) = x_1(t + \frac{1}{2}T)$

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- Let x(t) be a hyperbolic T-periodic solution

$$H=\{\gamma\in\Gamma:\gamma\{x(t)\}=\{x(t)\}\}$$
 spatiotemporal symmetries $\gamma\in H\Longrightarrow\theta\in[0,1)$ such that $\gamma x(t)=x(t+\theta T)$

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- H is rigid to equivariant perturbations
- Example: $H = \mathbf{Z}_2(1\ 2); \quad \theta = 0 \quad \text{or} \quad \theta = \frac{1}{2}$

Two Identical Cells: Solutions via Hopf Bifurcation

$$\begin{array}{cccc}
 & \dot{x}_1 & = & f(x_1, x_2, \lambda) \\
 & \dot{x}_2 & = & f(x_2, x_1, \lambda) \\
 & 0 & = & f(0, 0, \lambda)
\end{array}$$

$$x_1, x_2 \in \mathbf{R}^k$$

$$\bullet \quad J(\lambda) = \left[\begin{array}{cc} \alpha(\lambda) & \beta(\lambda) \\ \beta(\lambda) & \alpha(\lambda) \end{array} \right]; \quad \left[\begin{array}{c} x \\ x \end{array} \right], \left[\begin{array}{c} x \\ -x \end{array} \right] \text{ invariant subsp's}$$

 $\alpha =$ linear internal dynamics; $\beta =$ linear coupling

• eigenvalues of J are eigenvalues of $\alpha + \beta$ and $\alpha - \beta$

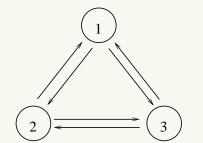
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 - $\alpha =$ linear internal dynamics; $\beta =$ linear coupling
- eigenvalues of J are eigenvalues of $\alpha + \beta$ and $\alpha \beta$
- $\alpha + \beta$ critical: synchronous periodic solutions
- $\alpha \beta$ critical: half-period out of phase periodic solutions

Three-Cell Bidirectional Ring: $\Gamma = \mathbf{D}_3$



$$\dot{x}_1 = f(x_1, x_2, x_3)
\dot{x}_2 = f(x_2, x_3, x_1) \quad f(x_2, x_1, x_3) = f(x_2, x_3, x_1)
\dot{x}_3 = f(x_3, x_1, x_2)$$

• Discrete rotating waves: $H = \mathbb{Z}_3(1\ 2\ 3), \ \theta = \frac{1}{3}$

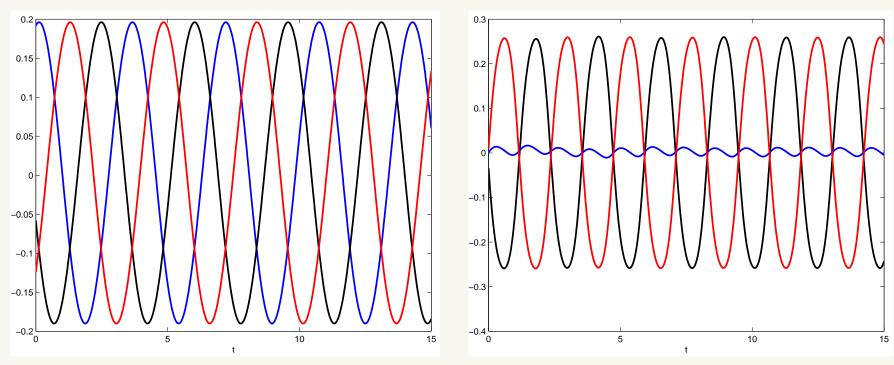
$$x_3(t) = x_2(t + \frac{1}{3}T) = x_1(t + \frac{2}{3}T)$$

• Out-of-phase periodic solutions: $H = \mathbf{Z}_2(1\ 3), \ \theta = \frac{1}{2}$

$$x_3(t) = x_1 \left(t + \frac{1}{2}T \right)$$
 and $x_2(t) = x_2 \left(t + \frac{1}{2}T \right)$

G. and Stewart (1986); van Gils and Valkering (1986)

Time Series and Phase Shifts



- Discrete Rotating Wave: $x_3(t) = x_2(t + \frac{1}{3}T) = x_1(t + \frac{2}{3}T)$
- Out-of-phase: $x_3(t) = x_1(t + \frac{1}{2}T)$ and $x_2(t) = x_2(t + \frac{1}{2}T)$

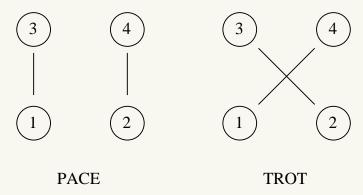
G. and Stewart (1986); van Gils and Valkering (1986)

Four Cells Do Not Suffice

Network produces walk. There is a four-cycle symmetry

$$(1\ 3\ 2\ 4)$$

Four-cycle permutes pace to trot

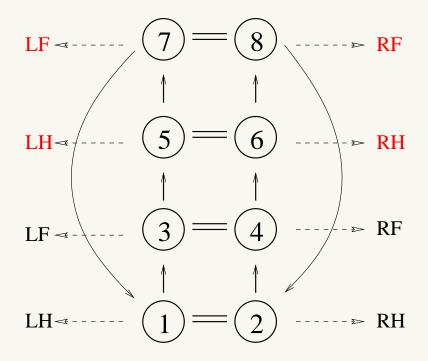


 CPG cannot be modeled by four-cell network where each cell gives rhythmic pulsing to one leg

G., Stewart, Buono, and Collins (1999)

Central Pattern Generators (CPG)

- Use gait symmetries to construct network and rhythms
 - 1) walk \Longrightarrow four-cycle ω in symmetry group
 - 2) pace or trot \Longrightarrow transposition κ in symmetry group
- Simplest network has $\mathbf{Z}_4(\omega) \times \mathbf{Z}_2(\kappa)$ symmetry



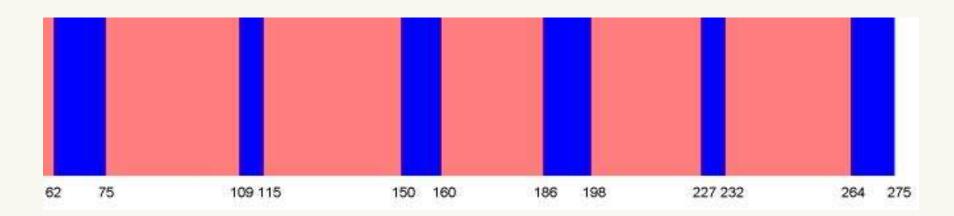
G., Stewart, Buono, and Collins (1999); Buono and G. (2001)

Primary Gaits or Hopf Bifurcation from Stand: $H = \mathbf{Z}_4(\omega) \times \mathbf{Z}_2(\kappa)$

Kernel of H	Phase Diagram	Gait
$\mathbf{Z}_4(\omega) \times \mathbf{Z}_2(\kappa)$	0 0 0 0	pronk
${f Z}_4(\omega)$	$\begin{array}{cc} 0 & \frac{1}{2} \\ 0 & \frac{1}{2} \end{array}$	pace
${f Z}_4(\kappa\omega)$	$ \begin{array}{ccc} $	trot
$\mathbf{Z}_2(\kappa) \times \mathbf{Z}_2(\omega^2)$	$\begin{array}{cc} \frac{1}{2} & \frac{1}{2} \\ 0 & 0 \end{array}$	bound
${f Z}_2(\kappa\omega^2)$	$\begin{array}{ccc} \frac{1}{4} & \frac{3}{4} \\ 0 & \frac{1}{2} \end{array}$	walk
${f Z}_2(\kappa)$	$\begin{array}{cc} 0 & 0 \\ \frac{1}{4} & \frac{1}{4} \end{array}$	jump

G., Stewart, Buono, and Collins (2000)

The Jump

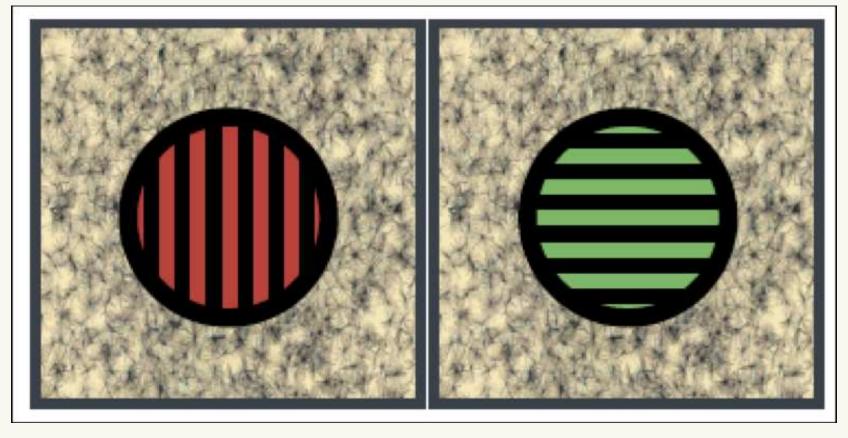


- Average Right Rear to Right Front = 31.2 frames
- Average Right Front to Right Rear = 11.4 frames
- \bullet $\frac{31.2}{11.4} = 2.74$

G., Stewart, Buono, and Collins (2000)

Binocular Rivalry: Different Images Presented to Two Eyes

How does the brain deal with CONTRADICTORY information



Often modeled by two units



Simplest Rivalry Equations for Two Units

• Units a and b consist of an activity variable $*^E$ (firing rate) and a fatigue variable $*^H$ (reduces activity on long time scale)

$$\begin{array}{rcl}
\varepsilon \dot{a}^{E} & = & -a^{E} + \mathcal{G} \left(I - \beta b^{E} - g a^{H} \right) \\
\dot{a}^{H} & = & a^{E} - a^{H} \\
\varepsilon \dot{b}^{E} & = & -b^{E} + \mathcal{G} \left(I - \beta a^{E} - g b^{H} \right) \\
\dot{b}^{H} & = & b^{E} - b^{H}
\end{array}$$

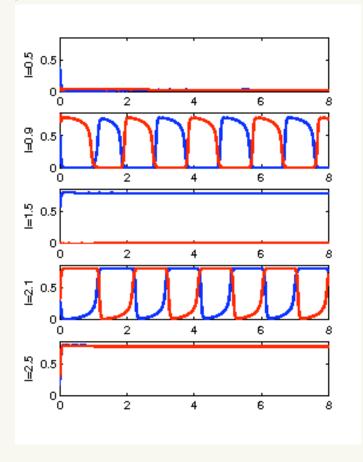
 $\begin{array}{l} \bullet \quad \mathcal{G} = \text{gain function} \\ \beta = \text{reciprocal inhibition} \\ I = \text{external signal strength} \\ g = \text{strength of reduction of } *^E \text{ by } *^H \\ \varepsilon \ll 1 \text{ is ratio of time scales on which } *^E \text{ and } *^H \text{ evolve} \\ \end{array}$

Laing and Chow (2002), Curtu, Shpiro, Rubin, and Rinzel (2008); Wilson (2009); Curtu (2010); Diekman, G., McMillen, and Wang (2012)

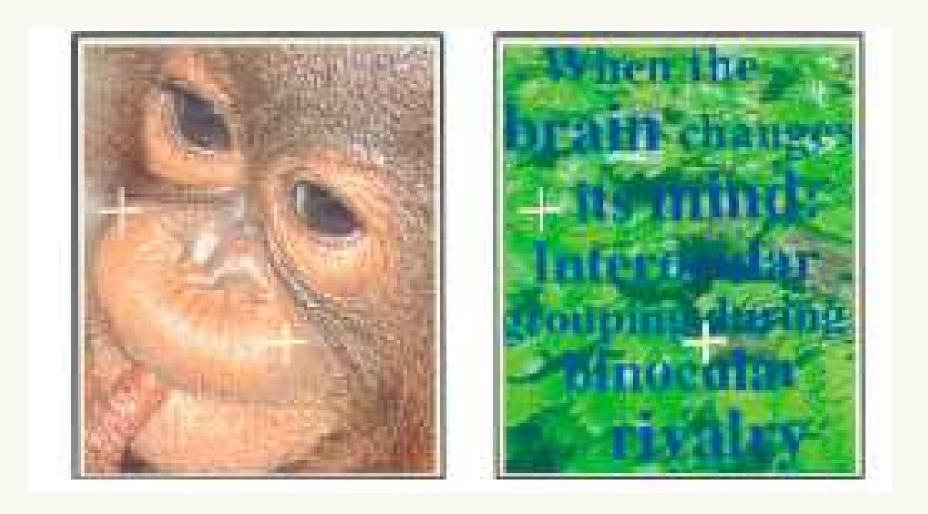
Two Unit Model Solution Types

Three types of states:

- Fusion = equilibria in which units have equal values
 Rigid fused states forced by symmetry
- Winner-Take-All = equilibria with different activity levels
- Rivalry = two units oscillate in periods of dominance
 Rigid rivalry forced by symmetry

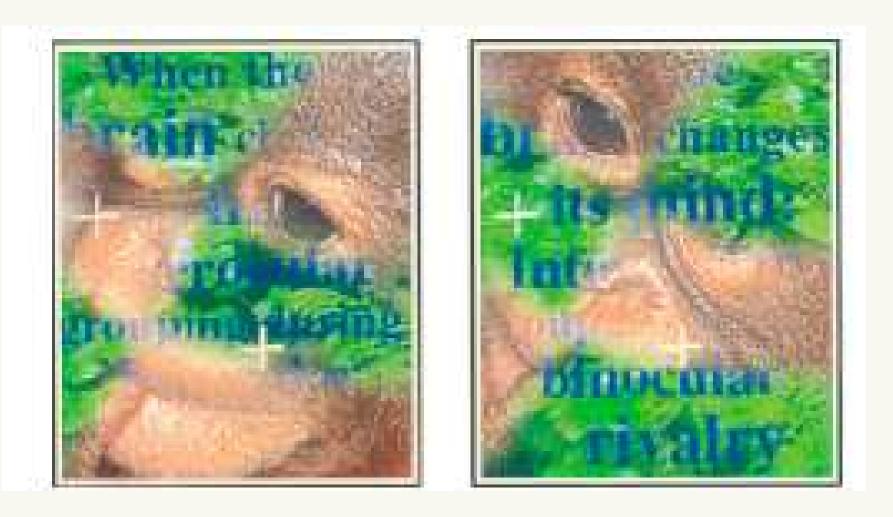


Kovács First Experiment: Conventional Monkey and Text



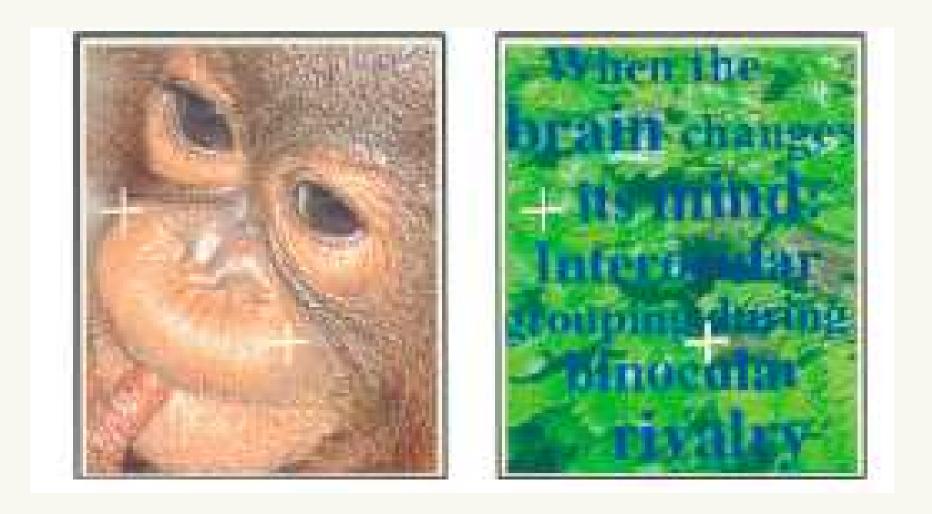
Kovács, Papathomas, Yang, and Fehér (1996)

Kovács Second Experiment: Scrambled Monkey and Text



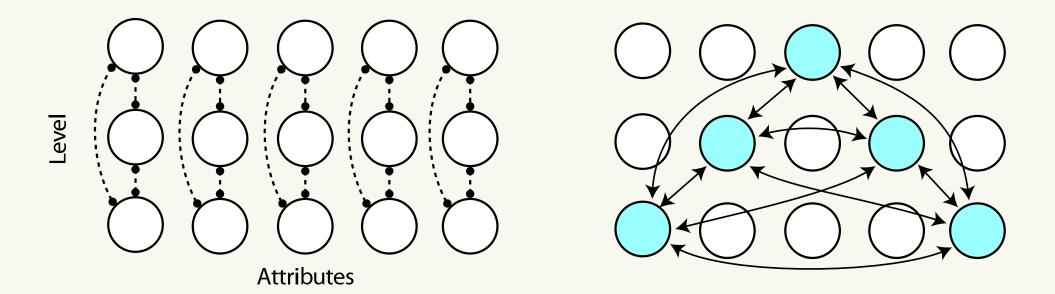
Kovács, Papathomas, Yang, and Fehér (1996)

Scrambled Monkey and Text: Interocular Groupings



Kovács, Papathomas, Yang, and Fehér (1996)

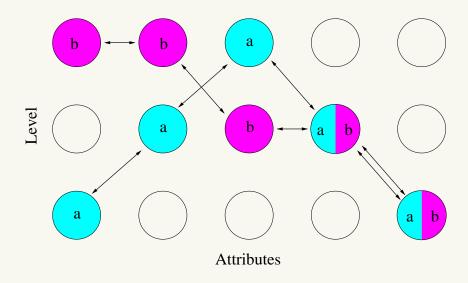
Wilson's Model for Generalized Rivalry



- Columns represent attributes; rows represent level of attribute
- (L) Inhibition between nodes in column (dashed lines)
- (R) Excitation between nodes in **learned** pattern (solid lines)

Wilson (2008, 2009)

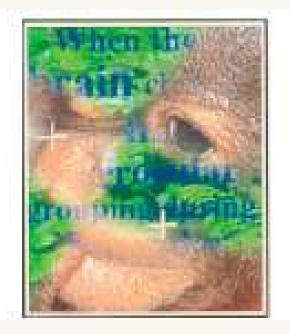
Two Learned Patterns *a* **and** *b*

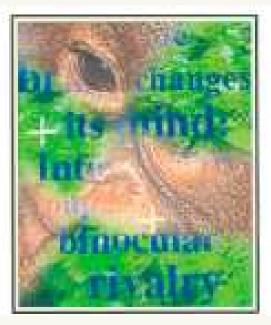


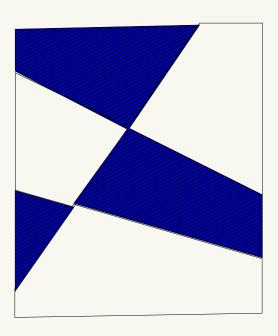
- Learned pattern: one node from each attribute column Excitation between nodes in learned pattern
- Derived Patterns: Patterns that are not learned

Diekman, G., McMillen, and Wang (2012, 2013)

Rivalry in Monkey and Text

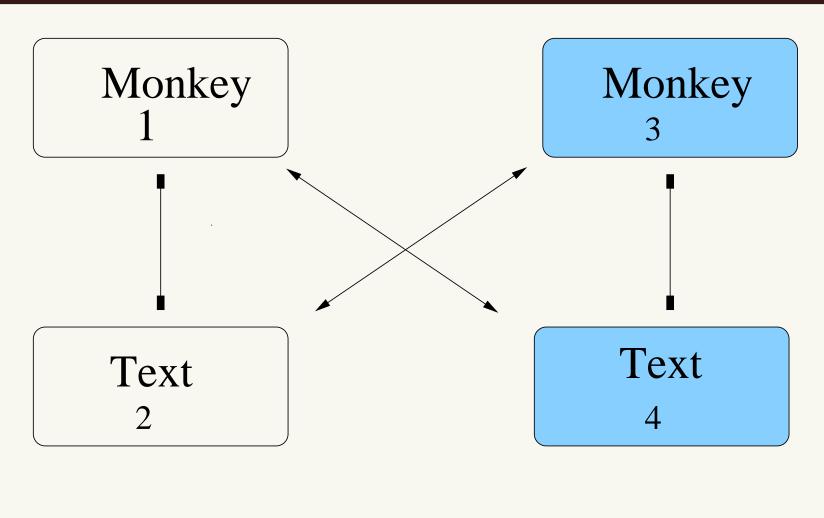






Kovács, Papathomas, Yang, and Fehér (1996)

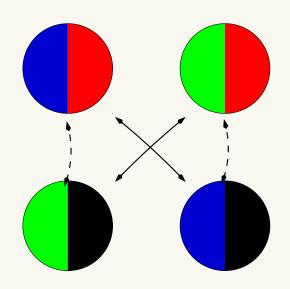
Wilson Network for Second Kovacs Experiment



WHITE AREA

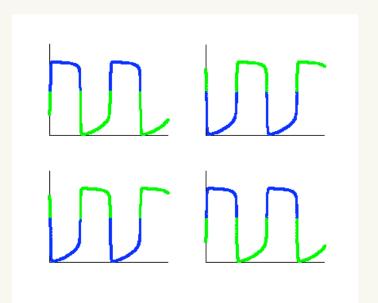
BLUE AREA

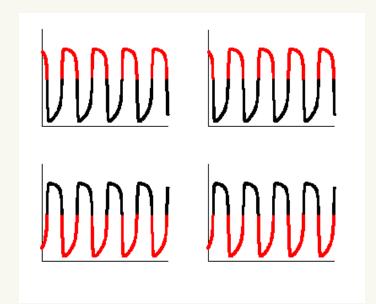
2 Attributes; 2 Levels; 2 Learned Patterns; 2 Derived Patterns



- $x_1^E > x_2^E$ and $x_3^E > x_4^E$ (whole monkey; derived, RED)
- $x_1^E > x_2^E$ and $x_4^E > x_3^E$ (mixed image; learned, BLUE)
- $x_2^E > x_1^E$ and $x_3^E > x_4^E$ (mixed image; learned, GREEN)
- $x_2^E>x_1^E$ and $x_4^E>x_3^E$ (all text; derived, BLACK)

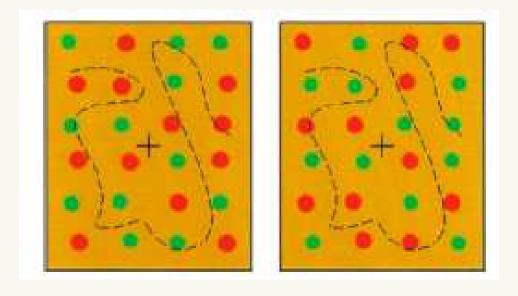
Patterns in Wilson Model of Kovacs Second Experiment



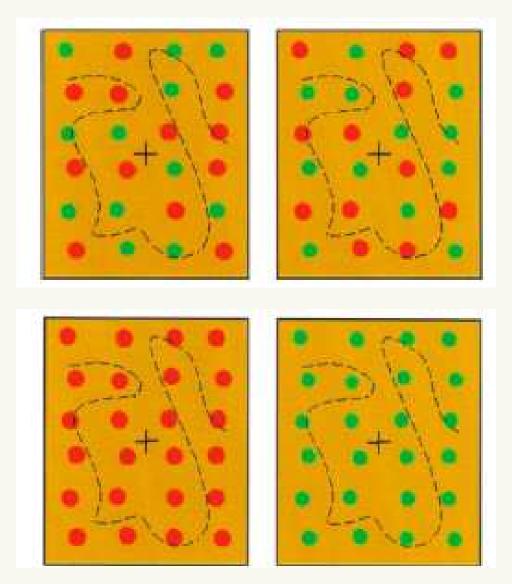


- 2 learned patterns: BLUE and GREEN
 - 2 **derived** patterns: RED and BLACK
- Rivalry: BLUE-GREEN (learned); BLACK-RED (derived)

Third Kovacs Experiment: Scrambled Disc Patterns

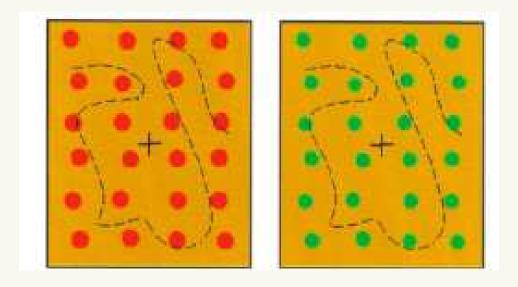


Third Kovacs Experiment: Scrambled Disc Patterns

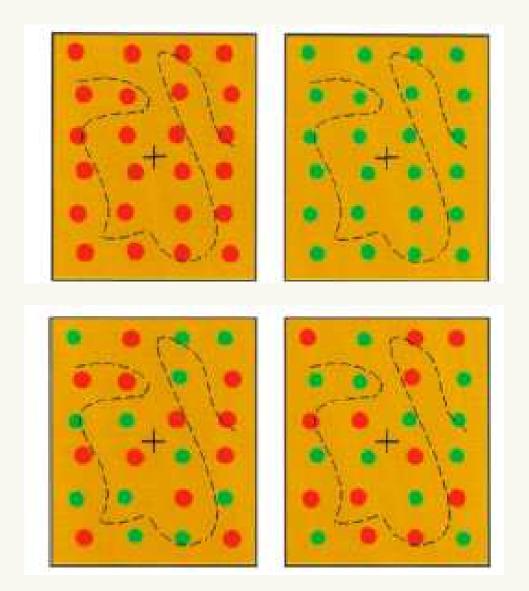


Kovács, Papathomas, Yang, and Fehér (1996)

Fourth Kovacs Experiment: Conventional Disc Patterns



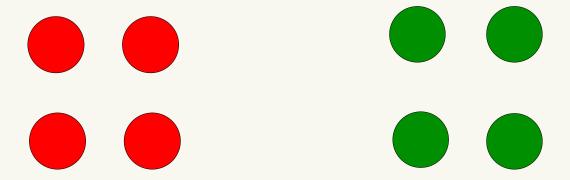
Fourth Kovacs Experiment: Conventional Disc Patterns



Kovács, Papathomas, Yang, and Fehér (1996)

Tong's Simplified Rivalry Between Disc Patterns

Rivalry between two learned patterns



Tong, Meng, and Blake (2006)

Tong's Simplified Rivalry Between Disc Patterns

Rivalry between two learned patterns



Tong, Meng, and Blake (2006)

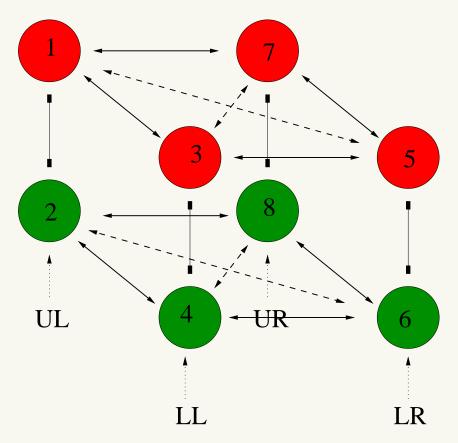
Rivalry between two derived patterns should also be observed



Rivalry Network for Conventional Tong Experiment

Two learned patterns: RED and GREEN

Symmetry group $\Gamma = \mathbf{D}_4 \times \mathbf{Z}_2(\rho)$

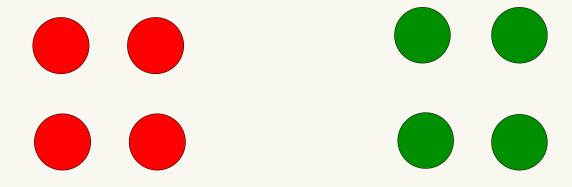


Diekman, G., and Wang (2013)

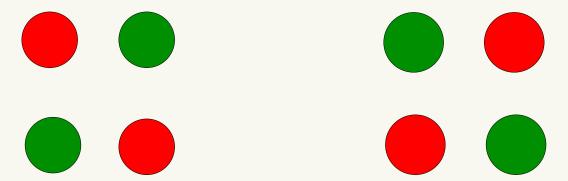
Patterns in Conventional Tong Experiment (1)

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Rivalry between learned patterns

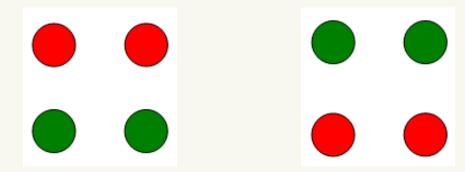


Rivalry between derived patterns

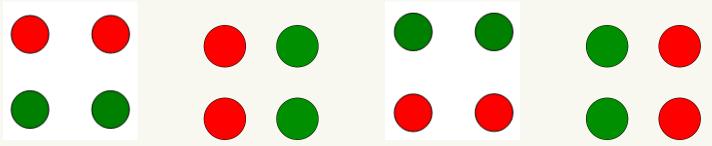


Patterns in Conventional Tong Experiment (2)

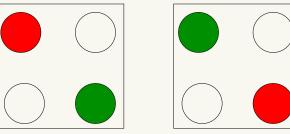
Adjacent colors



Adjacent colors rotate by 90° in quarter period



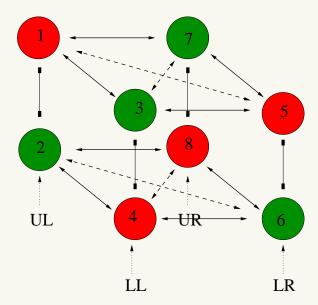
Two dots on diagonal alternate between; other two dots are fused



Wilson Network for Scrambled Tong Experiment



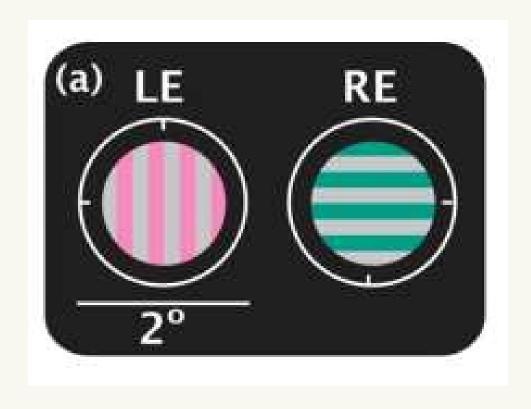
Learned Images in simplification of scrambled *colored dot* experiment



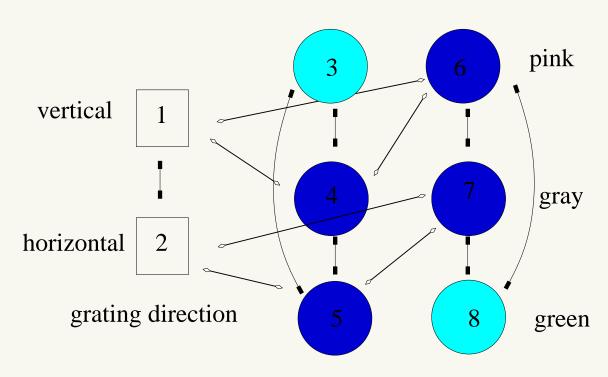
Network corresponding to simplified scrambled experiment

• Symmetry group is $\mathbf{D}_4 \times \mathbf{Z}_2(\rho)$ Learned & derived interchanged with conventional experiment

Shevell, St. Clair, and Hong (2008)



Wilson Network for Shevell, St. Clair, and Hong



left / top color right / bottom color

Symmetries

$$\rho = (1\ 2)(4\ 5)(6\ 7)
\tau = (3\ 8)(4\ 6)(5\ 7)
\rho\tau = (1\ 2)(3\ 8)(4\ 7)(5\ 6)
\rho = -1 = \rho\tau; \tau = +1$$

Fusion States

Gray-Pink
$$\leftrightarrow$$
 Green-Gray if $x_3^E < x_4^E = x_5^E$ Pink-Green if $x_3^E > x_4^E = x_5^E$

Maximally fused states = $\{x_1 = x_2; x_3 = x_8; x_4 = x_5 = x_6 = x_7\}$

Shevell, St. Clair & Hong 2008 Observed Percepts



Percepts observers reported during experiments

Synchrony and Synchrony Subspaces

RIGID PHASE-SHIFT SYNCHRONY

• $X(t) = (x_1(t), \dots, x_n(t))$ exhibits partial synchrony if

$$x_c(t) = x_d(t)$$

for two nodes c and d.

polydiagonal = subspace

 $\Delta = \{x : x_c = x_d \text{ for some subset of pairs of cells } c, d\}$

• **synchrony subspace** = polydiagonal flow-invariant ∀ admissibles

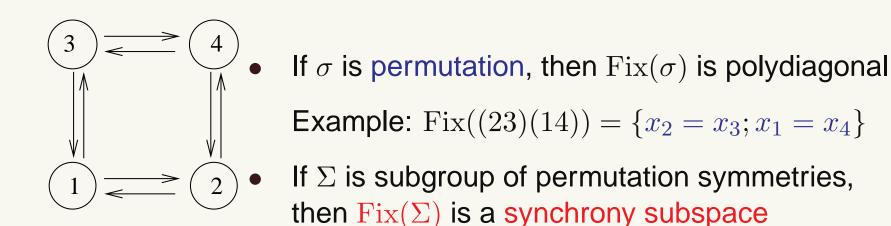
All solutions in a synchrony subspace exhibit partial synchrony

Synchrony Subspaces from Network Symmetry

- σ is a symmetry of $\dot{X}=F(X)$ if it maps solutions to solutions Equivalent to $F(\sigma x)=\sigma F(x)$
- $\operatorname{Fix}(\Sigma) = \{x \in \mathbf{R}^n : \sigma x = x \mid \forall \sigma \in \Sigma \}$ is flow-invariant Proof: $F(x) = F(\sigma x) = \sigma F(x)$

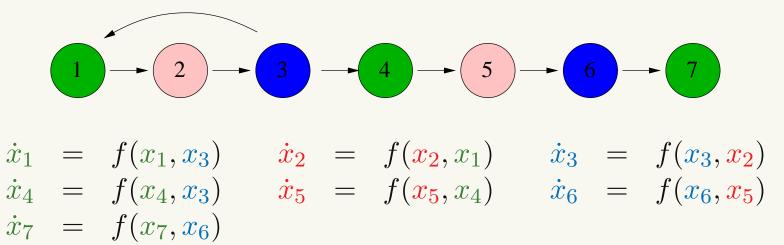
Synchrony Subspaces from Network Symmetry

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Balanced Colorings

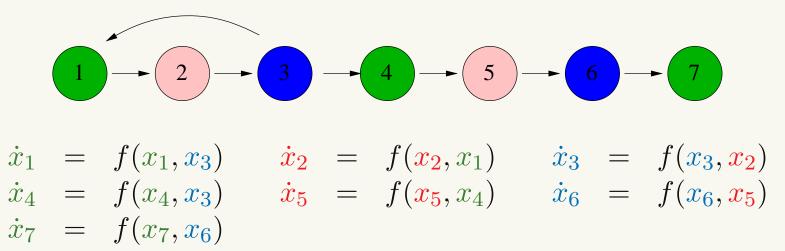
Synchrony subspaces do not have to be fixed-point subspaces



• $\Delta = \{x : x_1 = x_4 = x_7; \ x_2 = x_5; \ x_3 = x_6\}$ is flow-invariant

Balanced Colorings

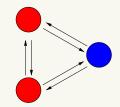
Synchrony subspaces do not have to be fixed-point subspaces



- $\Delta = \{x : x_1 = x_4 = x_7; \ x_2 = x_5; \ x_3 = x_6\}$ is flow-invariant
- ullet Color cells the same color if cell coord's in polydiagonal Δ are equal
- Coloring is balanced if all cells with same color receive equal number of inputs from cells of a given color
- Theorem 1: synchrony subspace ← balanced

Stewart, G., and Pivato (2003); G., Stewart, and Török (2005)

Quotient Networks with Self-Coupling & Multiarrows



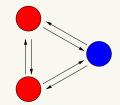
Synchrony subspace Fix(1 2) in bidirectional ring

$$\dot{x}_1 = f(x_1, x_2, x_3)$$
 $\dot{x}_2 = f(x_2, x_3, x_1)$ where $f(x, y, z) = f(x, z, y)$
 $\dot{x}_3 = f(x_3, x_1, x_2)$

Quotient network:



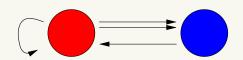
Quotient Networks with Self-Coupling & Multiarrows



Synchrony subspace Fix(1 2) in bidirectional ring

$$\dot{x}_1 = f(x_1, x_2, x_3)$$
 $\dot{x}_2 = f(x_2, x_3, x_1)$ where $f(x, y, z) = f(x, z, y)$
 $\dot{x}_3 = f(x_3, x_1, x_2)$

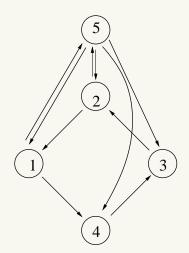
Quotient network:

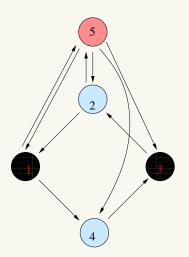


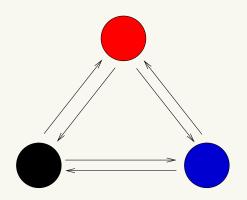
- Cell network with synchrony subspace leads to quotient network
- Theorem 2: Admissible ODE restricts to quotient admissible ODE
 Quotient admissible ODE lifts to admissible ODE

G., Stewart, and Török (2005)

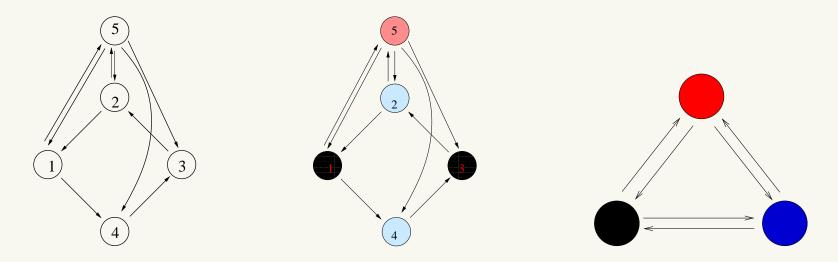
Asymmetric Network; Symmetric Quotient



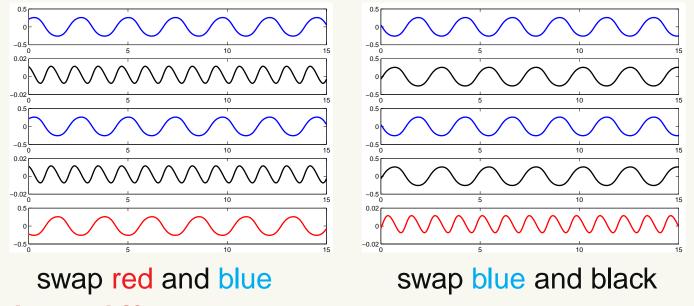




Asymmetric Network; Symmetric Quotient



• Quotient is bidirectional 3-cell ring with S₃ symmetry



Rigid phase shift; no symmetry

Rigid Phase-Shift ⇔ **Pattern of Phase-Shift Synchrony**

Theorem 3: Assume network is **transitive**.

Nonzero **rigid** phase-shift synchrony if and only if phase-shift forced by **symmetry on quotient network**

• Let Z(t) be the periodic solution. Quotient network corresponds to the synchrony subspace

$$\Delta_Z = \{x_i = x_j \quad \text{if} \quad z_i(t) = z_j(t) \ \forall t\}$$

Stewart and Parker (2008, 2009); G., Romano and Wang (2010, 2011)

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Casey Diekman	NJIT	Rivalry

Wilson Networks

Coupled cell networks have nodes and arrows & equivalence classes of nodes and arrows. Wilson network = coupled cell network such that

- (a) Nodes partition into **attribute columns**: $C = A_1 \cup \cdots \cup A_m$ where all nodes in an attribute column are cell equivalent. *Example*: Attribute might be color of a dot
- (b) Pattern = choice of one node in each attribute column Learned patterns given; other patterns derived Binocular rivalry has two learned patterns
- (c) Two types of arrows: *inhibitory* and *excitatory*
 - Nodes in same attribute column connected by inhibitory arrow
 - Nodes in same learned pattern connected by excitatory arrow
 - Inhibitory arrows and excitatory arrows are not arrow equivalent

Wilson considers networks where nodes are cell equivalent, inhibitory arrows are arrow equivalent, and excitatory arrows are arrow equivalent

Rivalry Networks

A rivalry network is a Wilson network such that

- (a) Attributes partition into **attribute types**Nodes in attribute equivalent columns are cell equivalent
 Inhibitory arrows are equivalent iff have equivalent attributes
- (b) **Feature** is property of pairs of nodes in different attribute columns Excitatory arrows equivalent iff connect nodes with same features

Examples:

- Dot experiments have distance feature: assigns to pair of nodes distance between dots
 Arrows can be equivalent only if connect equidistant nodes
- Dot experiments have color feature
 Excitatory arrows connecting nodes of same color and arrows connecting nodes of different colors are not equivalent