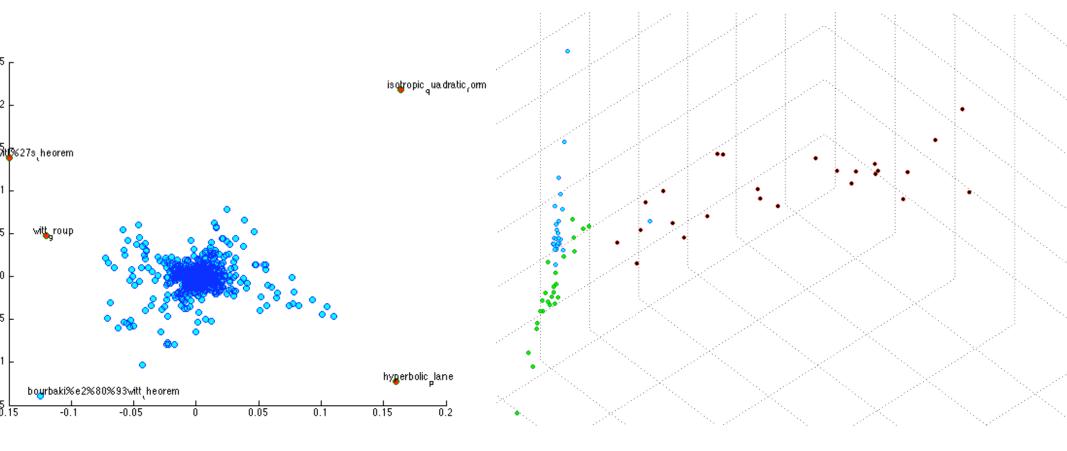
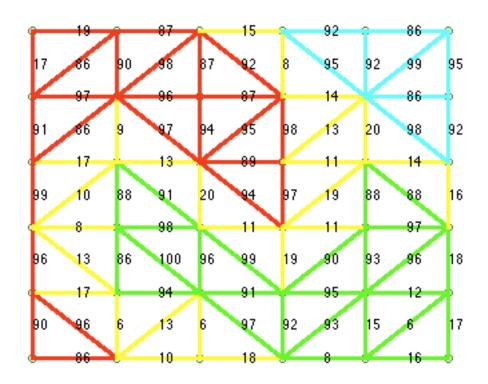
Fraud Detection, Quantum Mechanics, and Complex Systems Lecture 3, CSSS10



Greg Leibon
Memento, Inc
Dartmouth College

The Vagabond Problem: Finding a partition $\{R_i\}_{i=1}^K$ of our states into regions that minimizes

$$\sum_{i} P(X_{t+dt} \in \bar{R}_i \mid X_t \in R_i)$$



N=200; K=10; [states]=PlotTheState(N,K);

$$\chi_R(i) = \left\{ egin{array}{ll} rac{1}{\sqrt{\omega(R)}} & \mbox{if } i \in R \\ 0 & \mbox{if } i \in ar{R} \end{array}
ight.$$
 The solution (I think!)

if R_1 and R_2 are disjoint $\langle \chi_{R_1}, \chi_{R_2} \rangle_{\omega} = \delta_1^2$

Recall: $\langle f, g \rangle_{\omega} = f^{tr}[\omega]g$

How we really solve this?

Vagabond Theorem:

$$P(X_{t+dt} \in \bar{R} \mid X_t \in R) \approx ||\chi_R||_{Dir}^2 dt$$

proof:

$$P(X_{t+dt} \in \overline{R} \mid X_t \in R) = \frac{P(X_{t+dt} \in \overline{R}, X_t \in R)}{P(X_t \in R)} \qquad \text{(def. conditional prob.)}$$

$$= \frac{\sum_{i \in \overline{R}, j \in R} P(X_{t+dt} = i, X_t = j)}{\omega(R)} \qquad \text{(def.)}$$

$$= \frac{\sum_{i \in \overline{R}, j \in R} P(X_{t+dt} = i \mid X_t = j) P(X_t = i)}{\omega(R)} \qquad \text{(def. conditional prob.)}$$

$$= \frac{\sum_{i \in \overline{R}, j \in R} P(X_{t+dt} = i \mid X_t = j) P(X_t = i)}{\omega(R)} \qquad \text{(instantaneous transition rate)}$$

$$= \frac{\sum_{i \in \overline{R}, j \in S} P_{ij} \pi_i}{\omega(R)} dt \qquad \qquad (\omega_i = \tau_j \pi_i)$$

$$= \sum_{i \in \overline{R}, j \in R} \pi_i P_{ij} (\chi_R(i) - \chi_R(j))^2 dt \qquad \qquad \text{def.}$$

$$= (1/2) \sum_{i,j} \pi_i P_{ij} (\chi_R(i) - \chi_R(j))^2 dt \qquad \qquad \text{(summand 0 if } i, j \in S \text{ or } i, j \in \overline{R})$$

$$= \langle \chi_R, \Delta_\pi \chi_R \rangle_\pi dt \qquad \qquad \text{Green's Identity}$$

$$= ||\chi_R||_{Dir}^2 dt \qquad \qquad \text{conformal invariance}$$

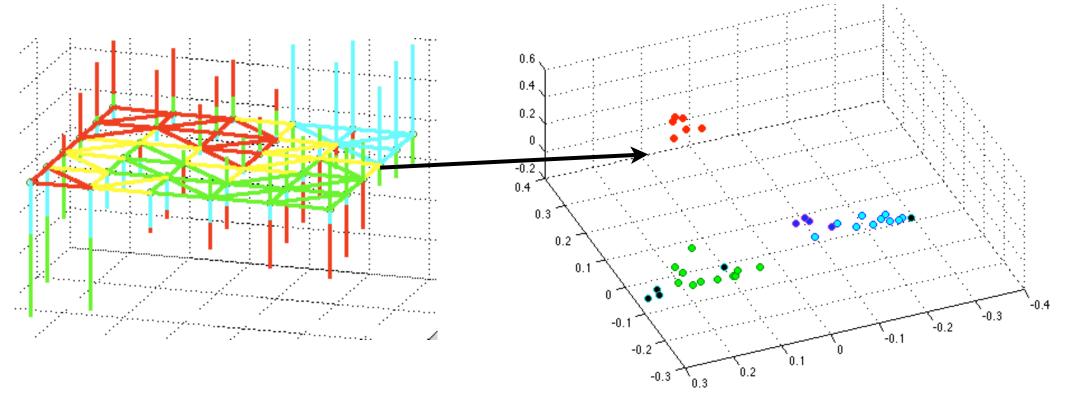
Q.E.D

$$\sum_{i} P(X_{t+dt} \in \bar{R}_i \mid X_t \in R_i) = \sum_{i} ||\chi_{R_i}||_{Dir}^2.$$

Now we can relax the Vagabond Problem and search for a ω -orthonormal set $\{f_i\}_{i=1}^L$ such that

$$\sum_{i}^{L}||f_{i}||_{Dir}^{2}$$

is minimized.



Now we cluster in Euclidean space...

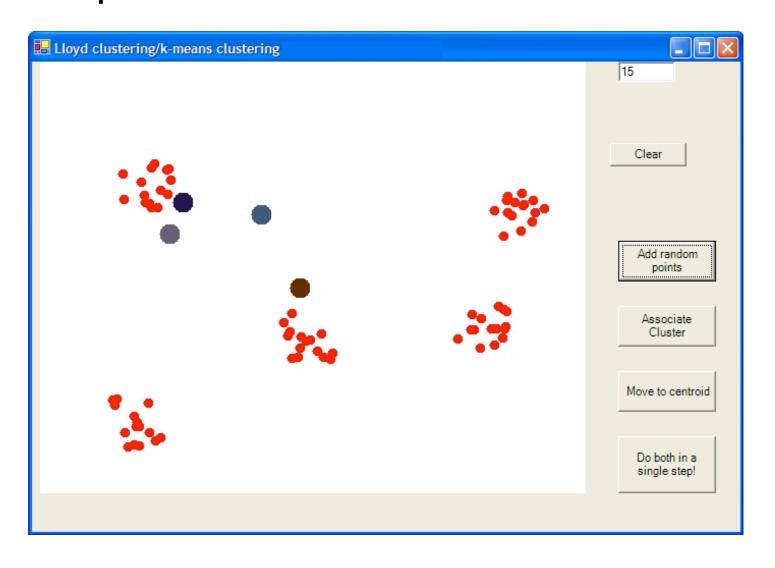
K-means

- Simplest clustering algorithm is k-means
- To run requires fixing K=#(Clusters)
- Requires an Euclidean type embedding
- We are attempting to minimizing a loss function:

$$L = \sum_{k=1}^{K} \sum_{x_i \in C_k} (x_i - \mu_k)^2$$

K-means algorithm:

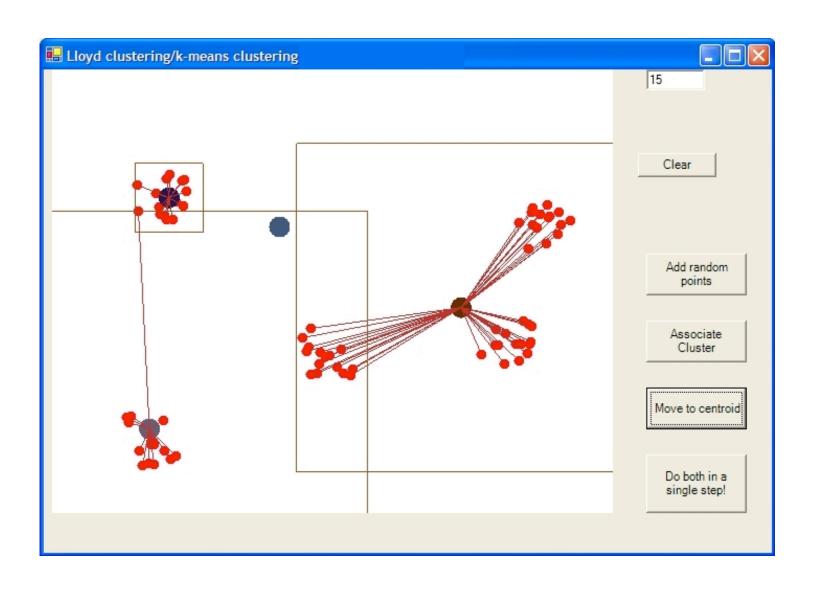
I. Randomly choose points in each cluster and compute centroids.



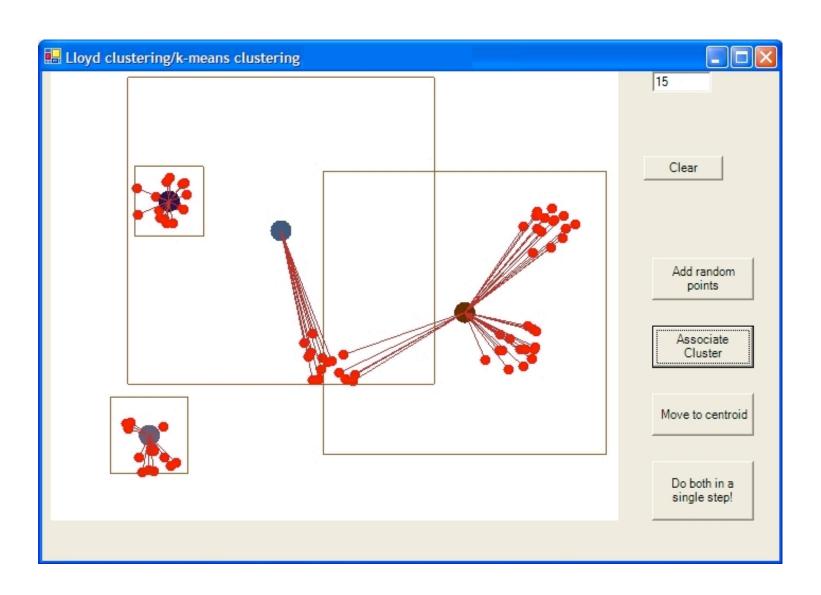
Example from: http://en.wikipedia.org/wiki/K-means_algorithm

2. Organize points by distance to the centroids.

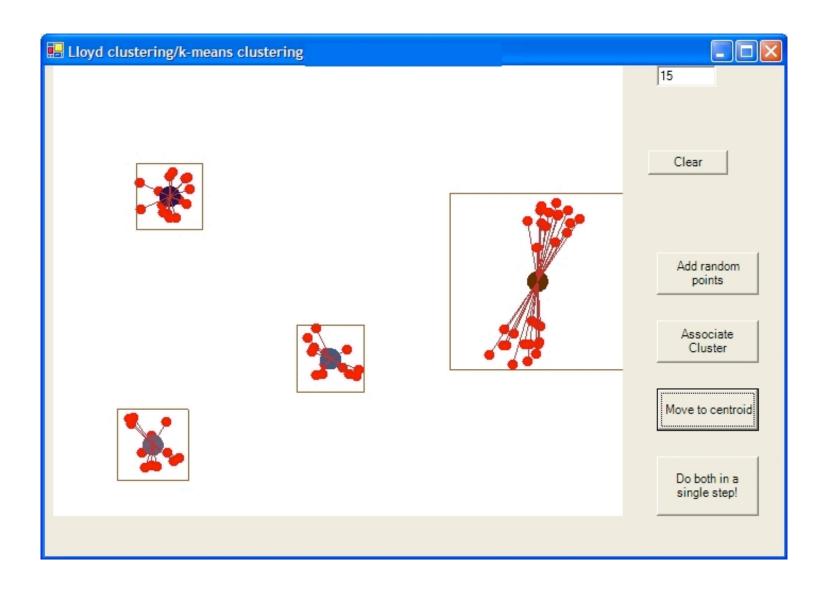
3. Update centroids



4. Repeat...

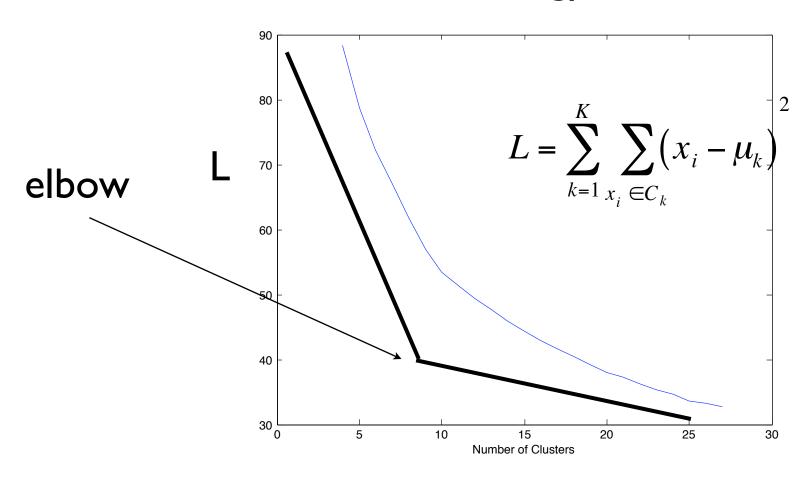


...until stable.



A hard part is choosing K=#(Clusters)

Elbowlogy:



Though is practice this rarely works in a complex multi-scalar system system

Now we can relax the Vagabond Problem and search for a ω -orthonormal set $\{f_i\}_{i=1}^L$ such that

$$\sum_{i}^{L}||f_{i}||_{Dir}^{2}$$

is minimized.

How?

 $< f_i, \Delta_{\omega} f_i >_{\omega} = ||f_i||^2_{Dir}$ is being used as a quadradic form, hence each term in the sum remains the same if we view Δ_{ω} and the Hermitian operator

$$C_{\omega} = \frac{\Delta_{\omega} + \Delta_{\omega}^*}{2}.$$

In other words,

$$< f, \Delta_{\omega} f>_{\omega} = < f, C_{\omega} f>_{\omega}$$

and we have replaced the minimization problem with minimizing

$$\sum_{i} \frac{\langle f_i, C_{\omega} f_i \rangle_{\omega}}{\langle f_i, f_i \rangle_{\omega}}.$$

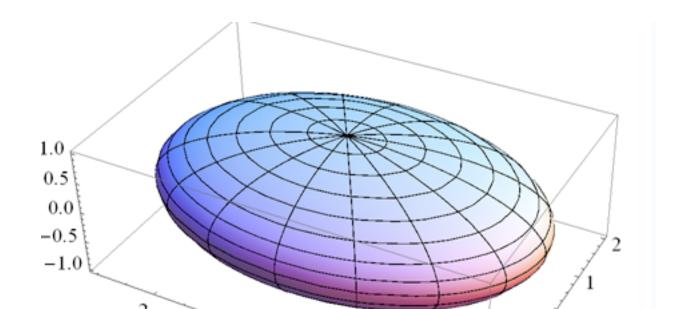
and we have replaced the minimization problem with minimizing

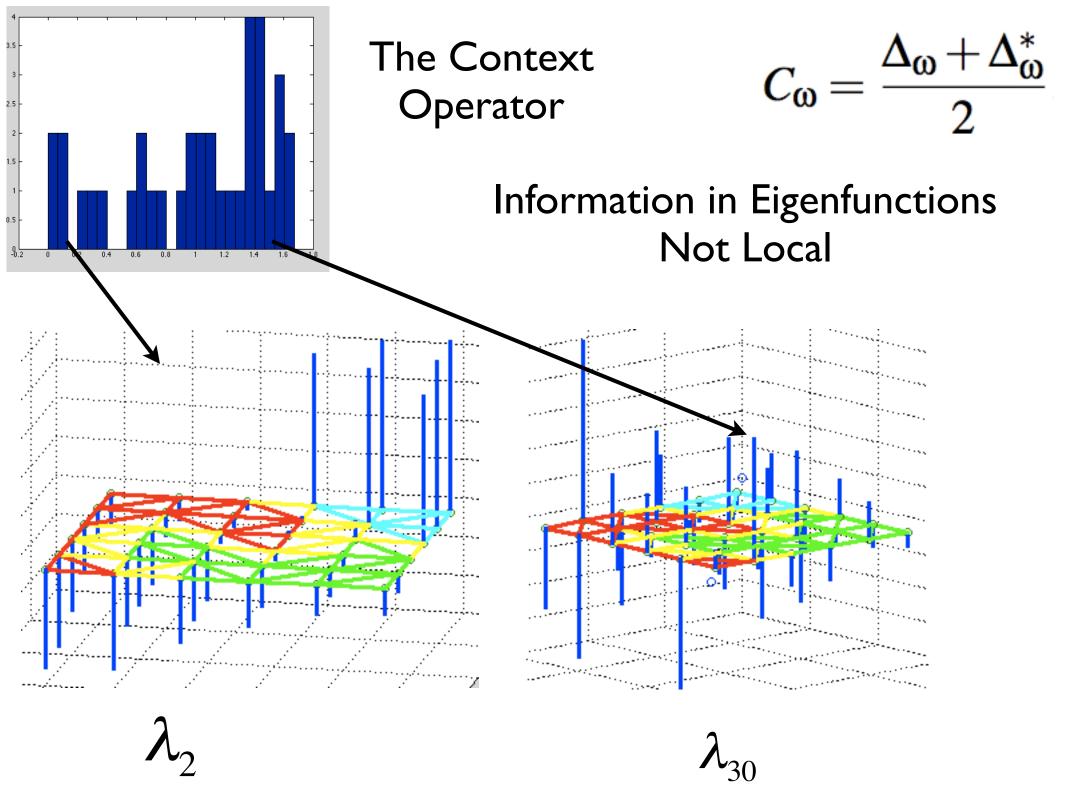
$$\sum_{i} \frac{\langle f_i, C_{\omega} f_i \rangle_{\omega}}{\langle f_i, f_i \rangle_{\omega}}$$
 where $\langle f, C_{\omega} g \rangle = \langle C_{\omega} f, g \rangle$

Why is this good, well it takes two to Tango!

 C_{ω} is our *context operator* and being Hermitian allows us to use the spectral theorem and the Raileigh-Rtiz method to identify our Vagabond Embedding as the ω -unit length eigenfunctions of C_{ω} with the smallest eigenvalues.

Spectral Theorem: If <,> is a Hermitian inner product and <Av,w>=<v,Aw>, then there is an orthonormal basis of A eigenvectors.







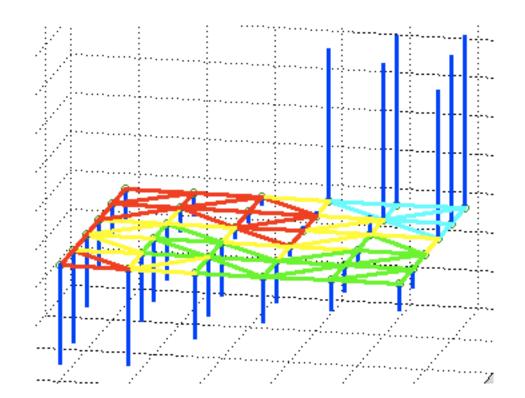
http://www.youtube.com/watch?v=Uu6Ox5LrhJg&feature=response_watch

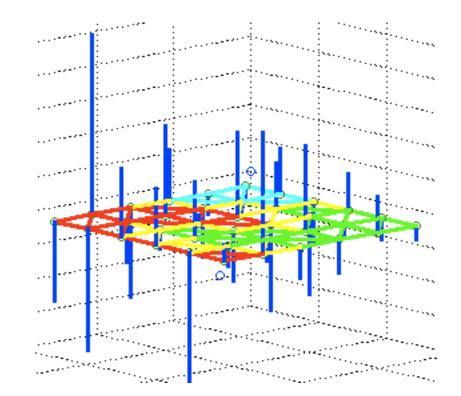
The Green-Kelvin Identity

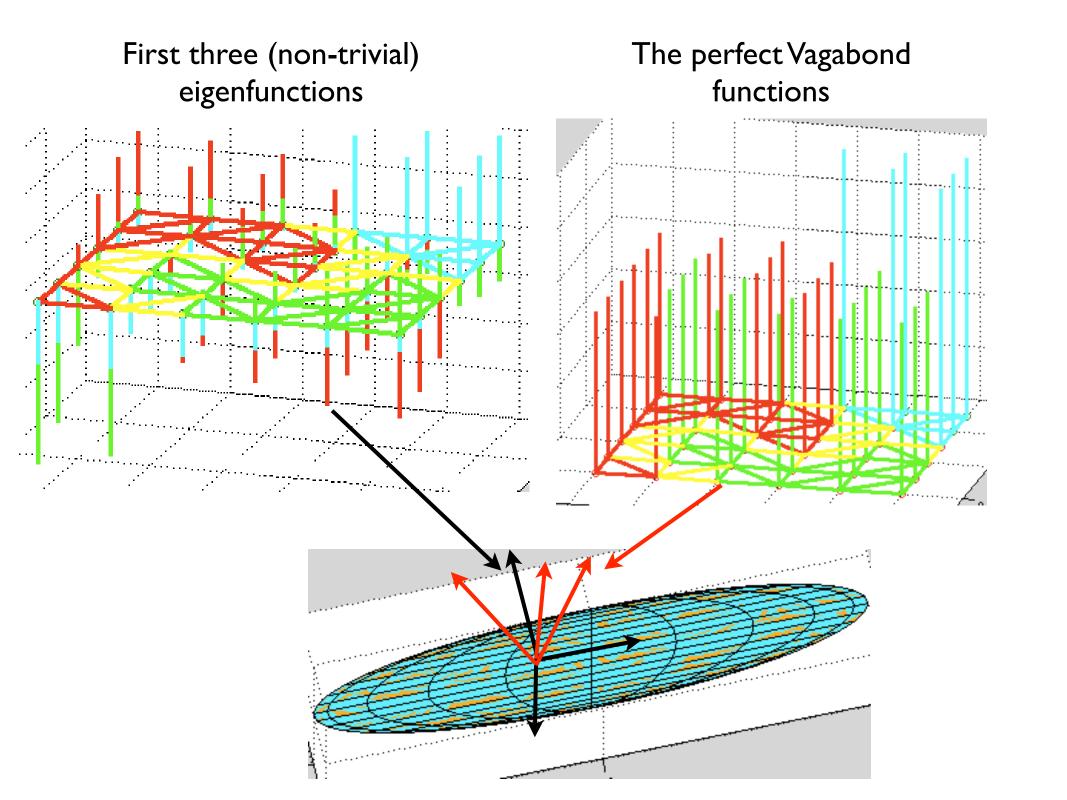
$$\langle f, \Delta f \rangle = \int f \Delta f d\vec{x} = \int |\nabla f|^2 d\vec{x} \qquad \langle f, g \rangle = \int_M f(x)g(x)dVol(x)$$

$$\langle f, g \rangle = \sum_{i} f_{i} w^{i} g_{i}$$

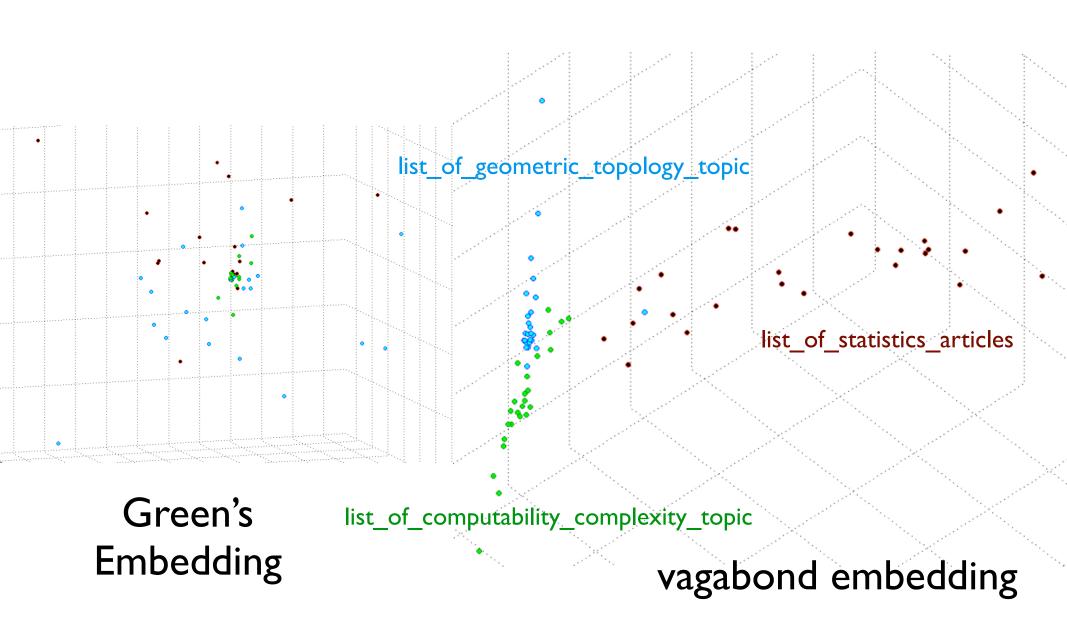
$$\langle f, \Delta f \rangle = \sum_{i,j} f_{j} w^{i} (I_{i}^{j} - P_{i}^{j}) f_{j}^{j} = \frac{1}{2} \sum_{i,j} w^{i} P_{i}^{j} (f_{i} - f_{j})^{2}$$

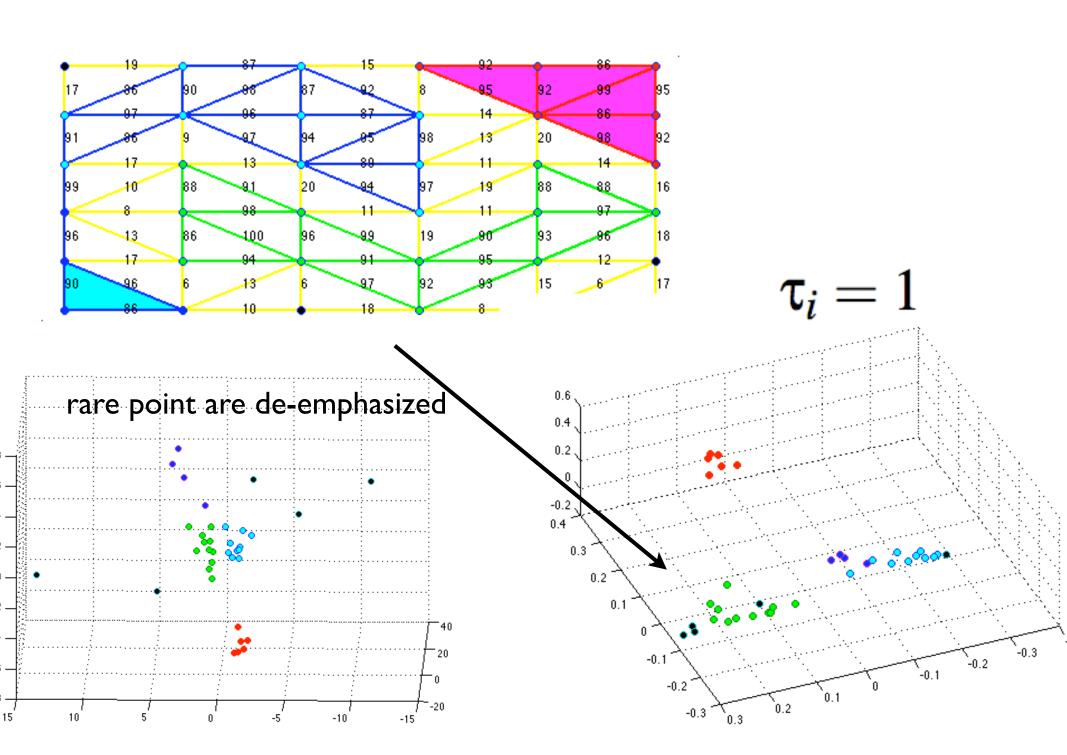




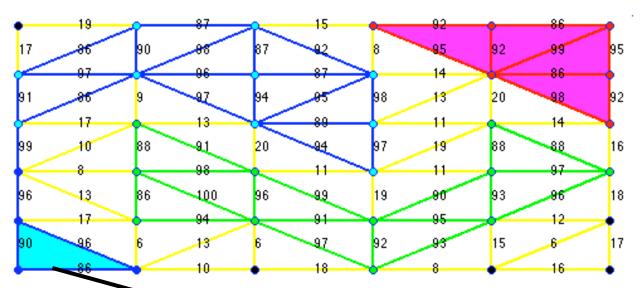


Space of mathematics

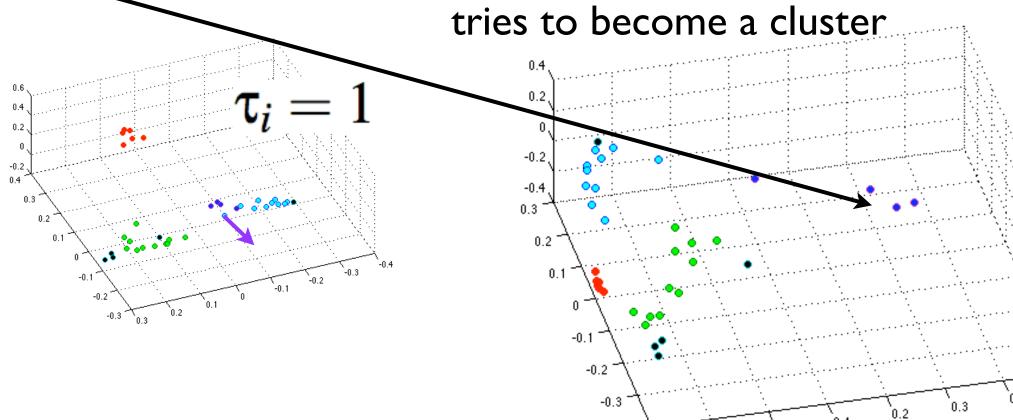




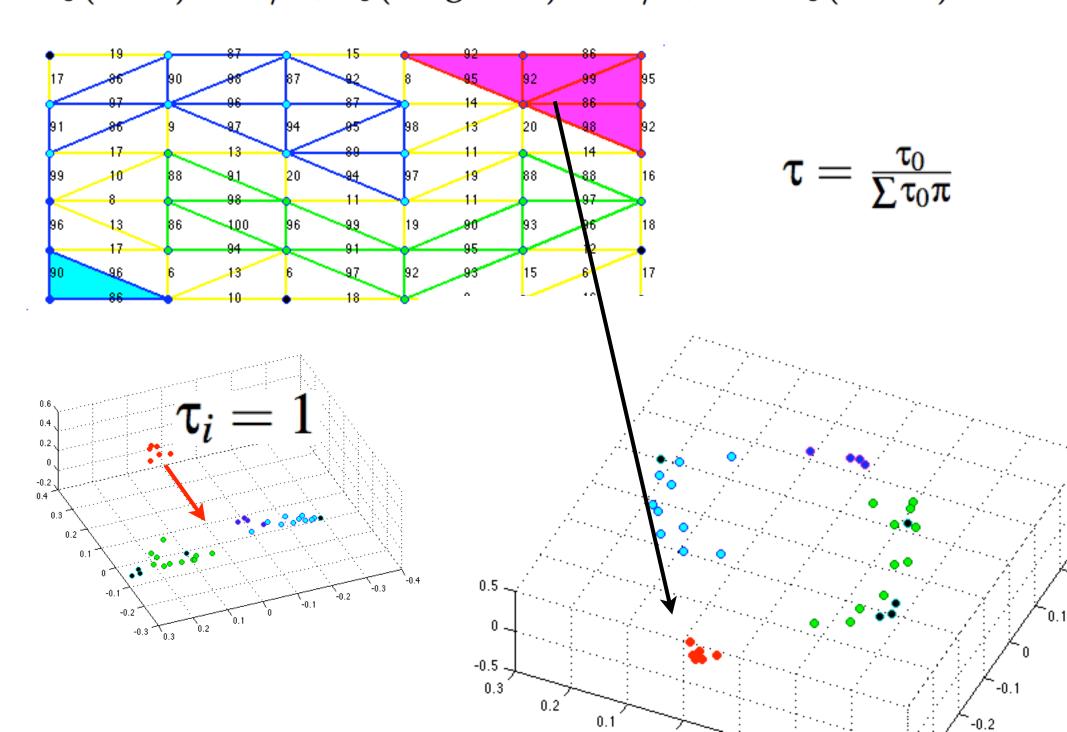
$$\tau_0(blue) = 5$$
, $\tau_0(magenta) = 1/5$, and $\tau_0(other) = 1$



$$au = rac{ au_0}{\Sigma au_0 \pi}$$



 $\tau_0(blue) = 1/5$, $\tau_0(magenta) = 1/5$, and $\tau_0(other) = 1$

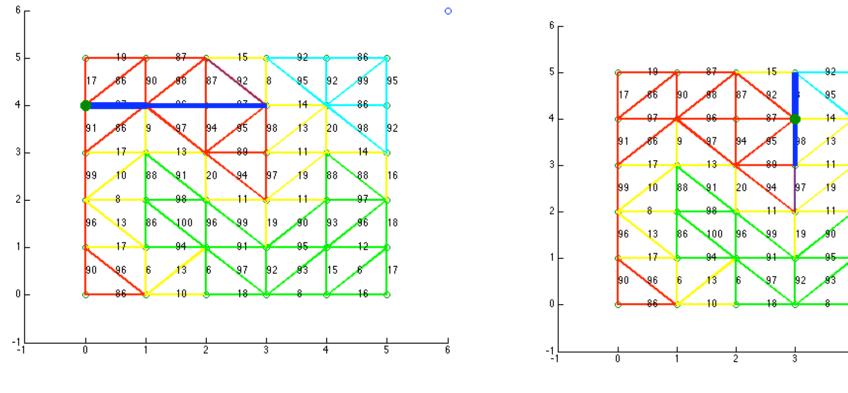


Kakawa's Salted Carmel time! (well, except Kakawa wasn't open)

What is the difference?

figure I

figure 2



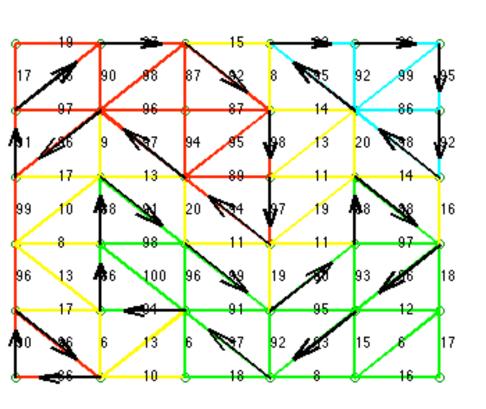
```
close all
Show=0; N=100;
figure(1); [states WR TR]=SnakeRev(N,5,'Vg',Show,1,1,1,1,1,0,0,0,0,0,0);
figure(2); [states WR TR]=SnakeRev(N,5,'Vg',Show,0,1,1,1,1,0,0,0,0,0,0,0);
```

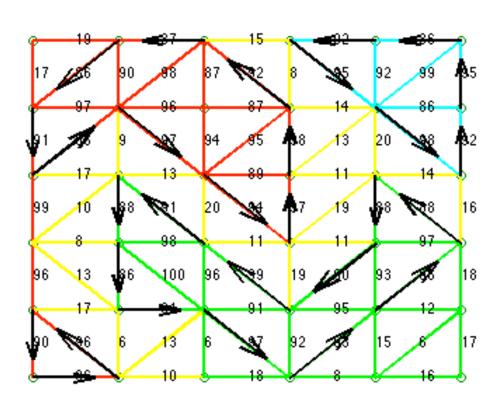
The answer:

0

figure I

figure 2





```
close all
Show=1; N=100;
figure(1); [states WR TR]=SnakeRev(N,5,'Vg',Show,1,1,1,1,1,0,0,0,0,0,0);
figure(2); [states WR TR]=SnakeRev(N,5,'Vg',Show,0,1,1,1,1,0,0,0,0,0,0,0);
```

Reversibility

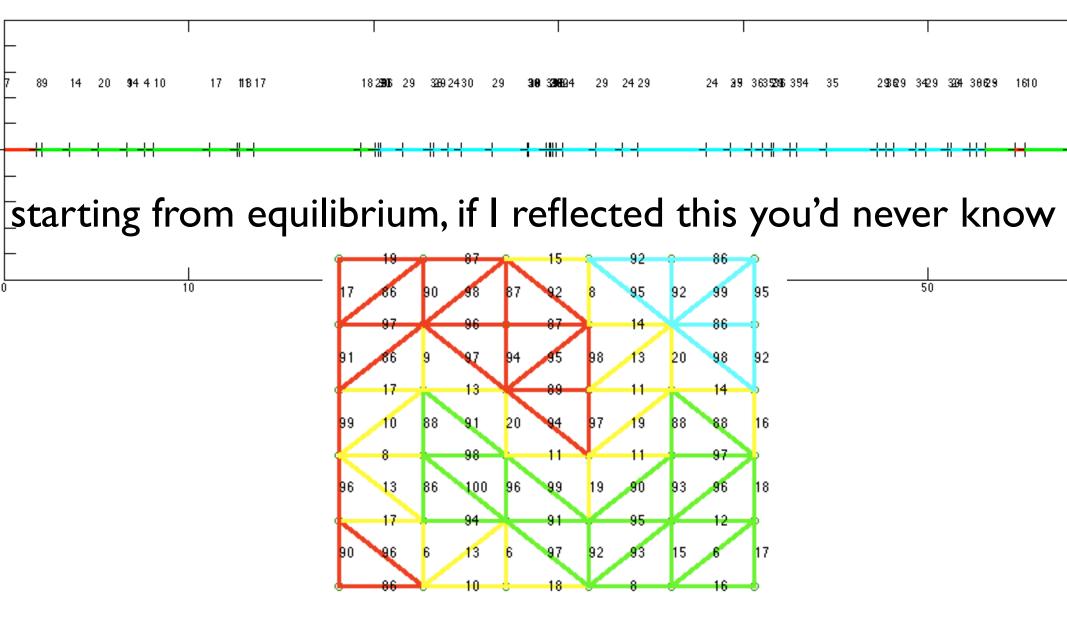
$$P^* = [\pi]^{-1} P^{tr} [\pi]$$

reversible if
$$P = P^*$$

Theorem: Reversible if and only if there is symmetric conductance matrix such that

$$P = [W\mathbf{1}]^{-1}W$$

$$\pi = W1/\left(1^{tr}W1\right)$$



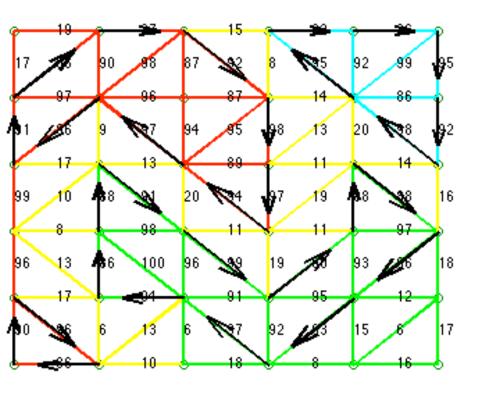
N=200; K=10; [states]=PlotTheState(N,K);

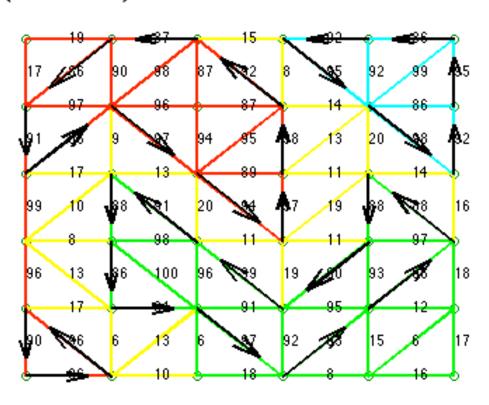
Theorem: For a reversible chain, the vagabond embedding is the PCA of the green's embedding weighted by the equilibrium vector

Theorem: For any chain P, there is a unique (up to a multiplicative constant) conductance W and divergence free, compatible flow F such that

$$P = [W1]^{-1}(W+F)$$
 $P^* = [W1]^{-1}(W-F)$

$$\pi = W1/\left(1^{tr}W1\right)$$





Flow Independence Corollary: The vagabond embedding and C_{ω} depend only on on the Conductance W and and the equilibrium measure ω , and are independent of the flow F.

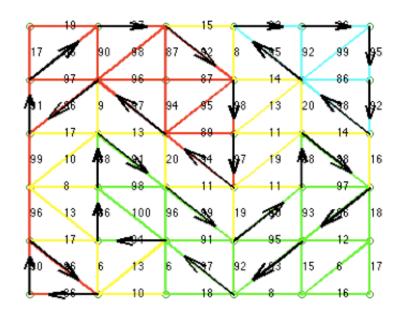
proof: Simply note

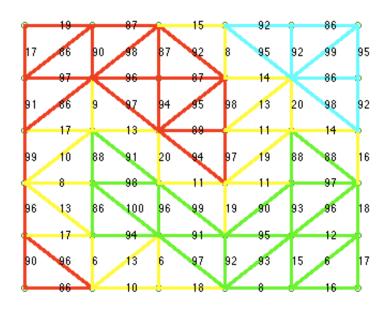
$$C_{\omega} = \frac{\Delta_{\omega} + \Delta_{\omega}^{*}}{2} = [\tau]^{-1} \left(I - \frac{P + P^{*}}{2} \right) = [\tau]^{-1} \left(I - [\pi]^{-1} W \right)$$

we have:

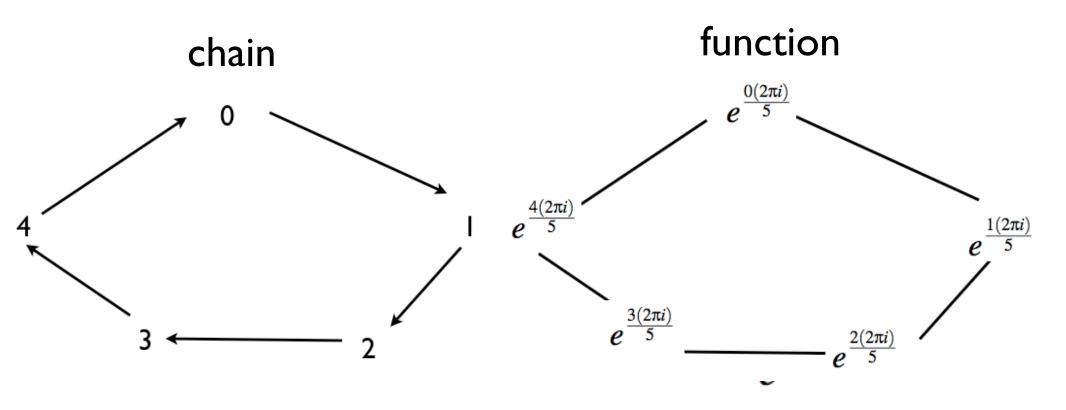
Q.E.D

Clearly the vagabond is missing half the geometry.





Here we are going explore a scenario in our space. This scenario is going to be 'cycle like behavior', and hence we must ask: what does a directed cycle look like from the point of view of the functions on the chain?



Shall we dance?

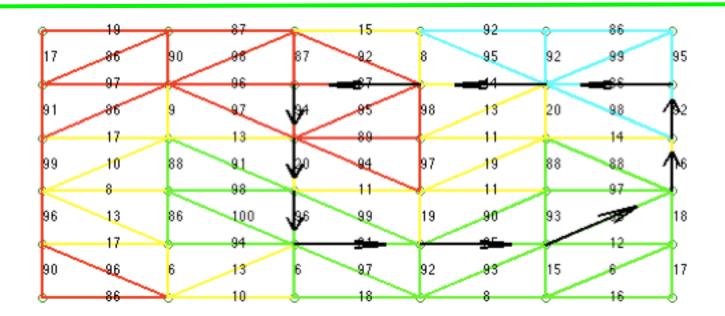
$$e^{-i\left(\frac{K2\pi}{N}\right)n}P^nf=f$$

More generally, cycles will support functions where there is a relatively large frequency κ such that

$$e^{-i\kappa t}e^{-t\Delta}f$$

is pointwise relatively stable. One way to find largish κ such the function f and κ pair locally minimize $\left|\left|\frac{d}{dt}e^{-(i\kappa+\Delta)t}f\right|\right|_{\omega}^2$ at t=0. So we are looking for critical points of

$$F(\kappa, f) = \left| \left| \frac{d}{dt} e^{-(i\kappa + \Delta)t} f \right| \right|_{\omega}^{2} \right|_{t=0}.$$

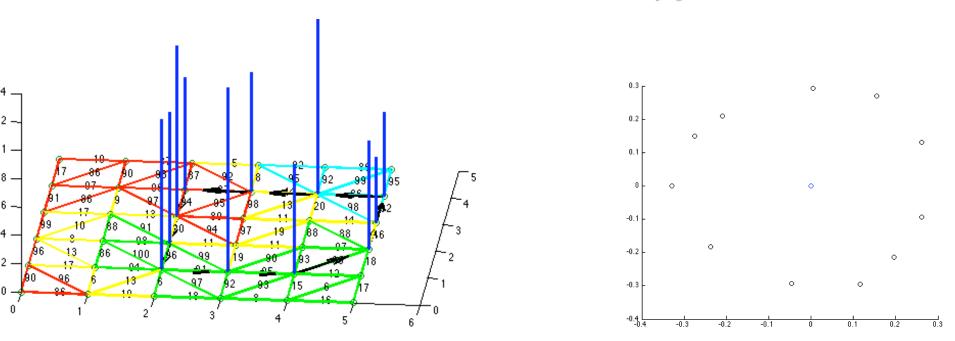


More generally, cycles will support functions where there is a frequency κ such that

$$e^{-i\kappa t}e^{-t\Delta}f$$

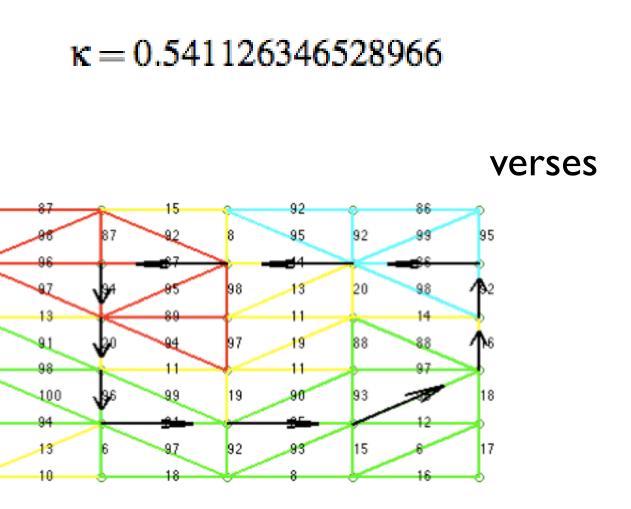
is pointwise relatively stable. One way to find such functions would be to minimize $\left|\left|\frac{d}{dt}e^{-(i\kappa+\Delta)t}f\right|\right|_{\omega}^{2}$ at t=0. Hence we are looking for critical points of

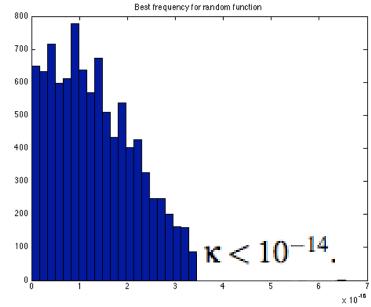
$$F(\kappa, f) = \left| \left| \frac{d}{dt} e^{-(i\kappa + \Delta)t} f \right| \right|_{\omega}^{2} \right|_{t=0},$$



$$\kappa = 0.541126346528966$$

We might call these local minima cycle functions. To get a sense for 'largish', putting a cycle on our toy chain as in figure 1, we can compare the function in the figure 5, which for $\kappa = 0.541126346528966$ has a nearly zero derivative, to a random function (real and imaginary parts random in [0,1] normalized to have norm 1). For a random function we find its optimal κ for minimizing the derivative and after 10000 runs the largest such $\kappa < 10^{-14}$.





How to find them...

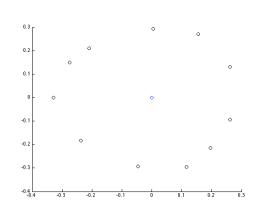
Cycle Detection Lemma: If $B_{\omega}f = \kappa f$ where $S_{\omega} = i\frac{\Delta - \Delta^*}{2}$, then $\frac{d}{d\kappa}F = 0$.

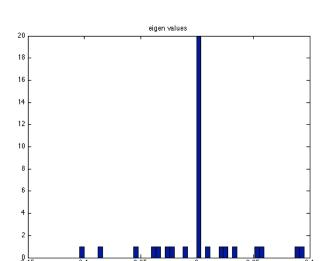
proof:

$$\begin{split} 0 &= \frac{d}{d\kappa} \left| \left| \frac{d}{dt} e^{-(i\kappa + \Delta)t} f \right| \right|_{\omega}^{2} \right|_{t=0} & \text{(definition)} \\ &= \frac{d}{d\kappa} \left| \left| (i\kappa + \Delta) e^{-(i\kappa + \Delta)t} f \right| \right|_{\omega}^{2} \right|_{t=0} & \text{(matrix derivative, sec 1.1ex??)} \\ &= \frac{d}{d\kappa} \left| \left| (i\kappa + \Delta) f \right| \right|_{\omega}^{2} & \text{($e^{At} = I$ at $t = 0$)} \\ &= \frac{d}{d\kappa} \left\langle (i\kappa + \Delta) f, (i\kappa + \Delta) f \right\rangle_{\omega} & \text{(def. $<,>_{\omega}$)} \\ &= \left\langle if, (i\kappa + \Delta) f \right\rangle_{\omega} + \left\langle (i\kappa + \Delta) f, if \right\rangle_{\omega} & \text{(chain rule)} \\ &= \left\langle f, (\kappa - i\Delta) f \right\rangle_{\omega} + \left\langle (\kappa - i\Delta) f, f \right\rangle_{\omega} & \text{(conjugate linear)} \\ &= \left\langle f, ((\kappa - i\Delta) + (\kappa - i\Delta)^{*}) f \right\rangle_{\omega} & \text{(compute adjoint)} \end{split}$$

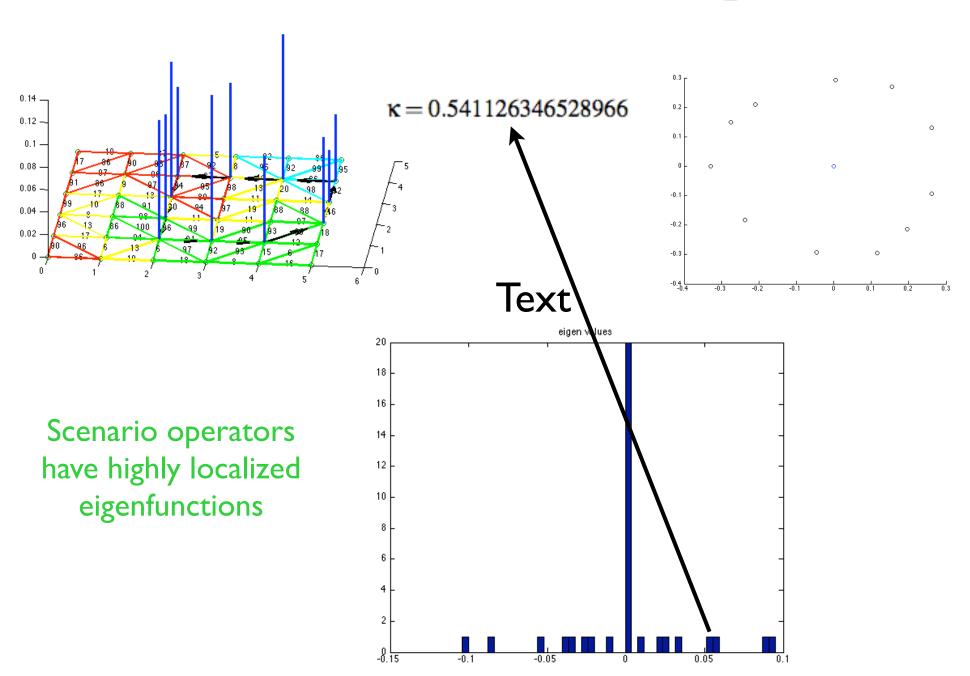
Q.E.D

 $\kappa = 0.541126346528966$





The scenario operator: $S_{\omega} = i \frac{\Delta - \Delta^*}{2}$



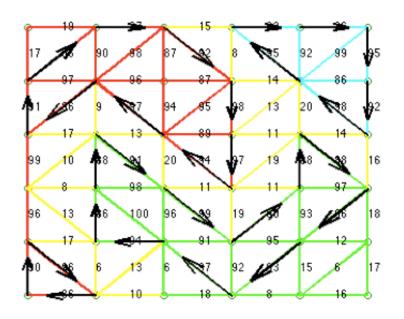
The scenario operator: $S_{\omega} = i \frac{\Delta - \Delta^*}{2}$

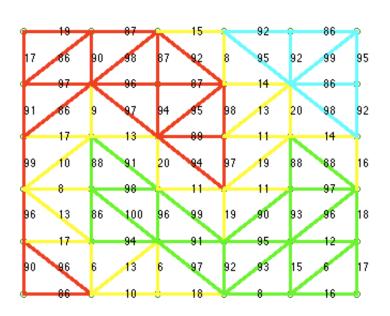
Condctance Independence Corollary: S_{ω} depends only on the flow F and equilibrium vector ω and is independent of conductance C. Furthermore the chain is reversible if and only if $S_{\omega} = 0$.

proof:

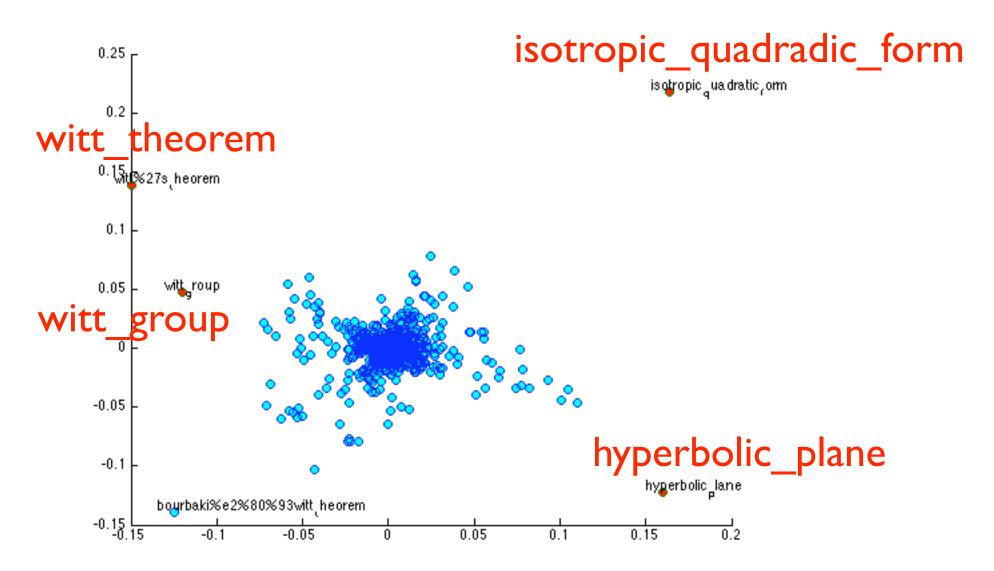
$$S_{\omega} = i \frac{\Delta_{\omega} - \Delta_{\omega}^{*}}{2} = i[\tau]^{-1} \left(\frac{P - P^{*}}{2} \right) = i[\tau]^{-1} [\pi]^{-1} F = i[\omega]^{-1} F$$

Q.E.D



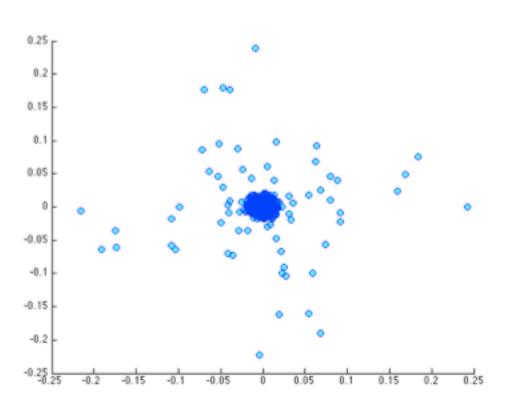


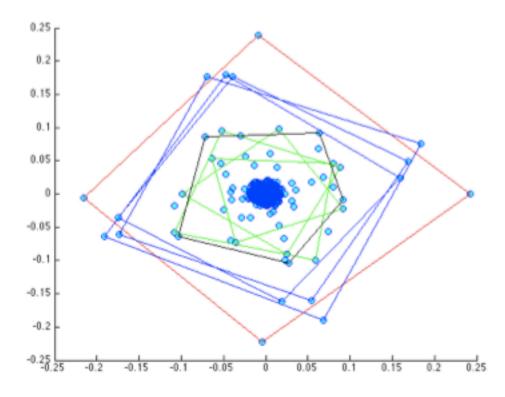
Eigenfunction in the Space of Mathematics



Scenario operators have highly localized eigenfunctions

Reality....





We need something like the vagabond clustering for cycles

The Co-conformal Cycle Hunt



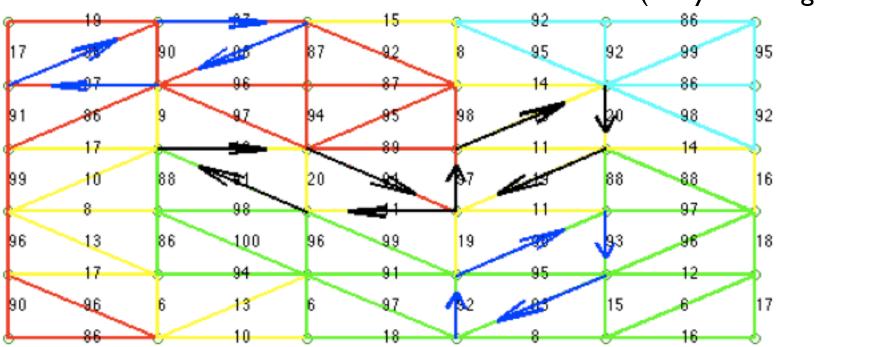
Co-conformal Magnetization

Not all cycles are created equally.... Wy do need this fellow to find a naughty cycle?

Claim: The operator $i[\Delta_{\omega}, \Delta_{\omega}^*] = \Delta_{\omega} \Delta_{\omega}^* - \Delta_{\omega}^* \Delta_{\omega}$

allows us to detect scenarios which are anomalous relative to the context

(all cycles weight 10)

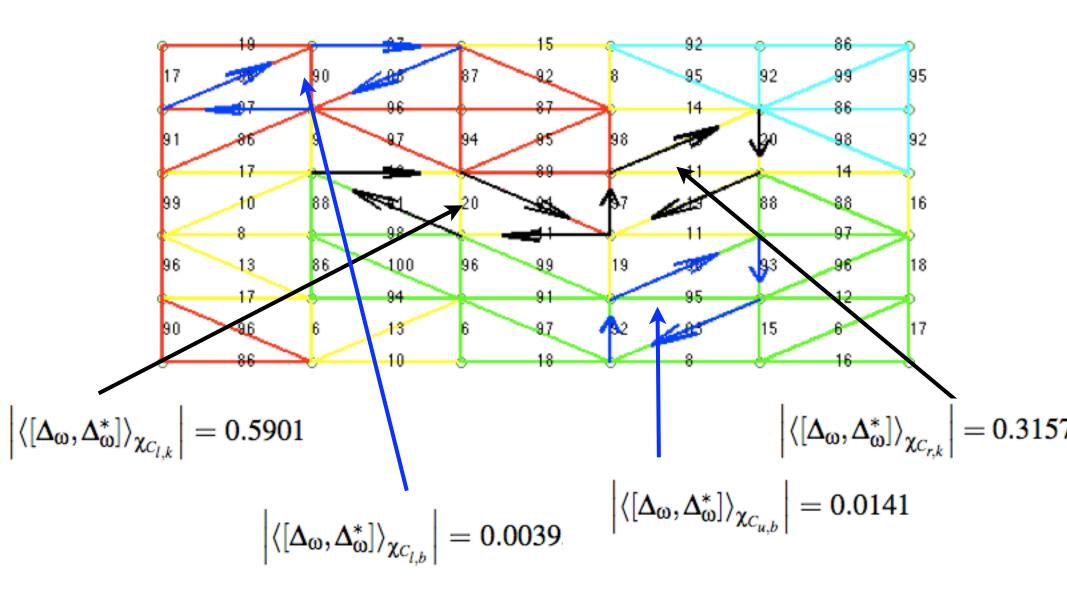


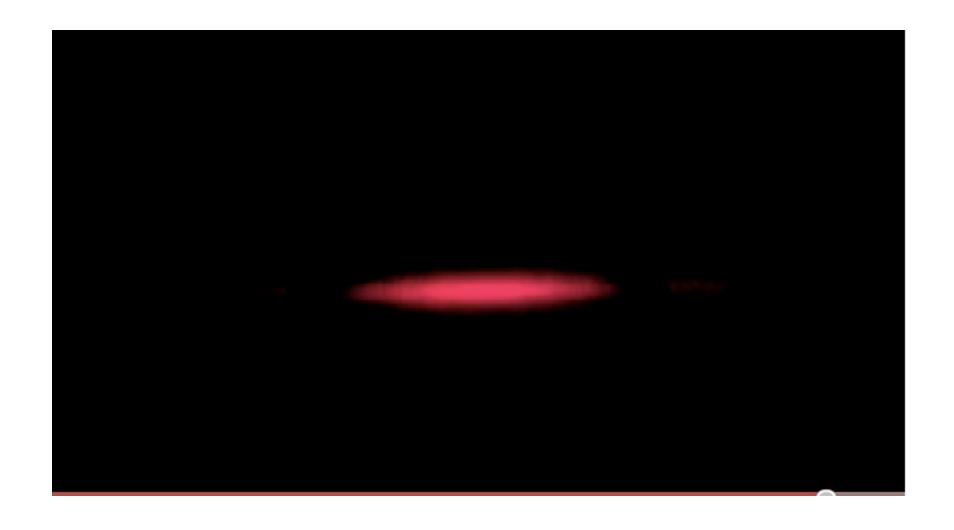
Which cycles are in context and which are not?

Claim: The operator

$$i[\Delta_{\omega}, \Delta_{\omega}^*] = \Delta_{\omega} \Delta_{\omega}^* - \Delta_{\omega}^* \Delta_{\omega}$$

allows us to detect scenarios which are anomalous relative to the context





http://www.youtube.com/watch?v=KT7xJ0tjB4A

Mathematics of Quantum Mechanics

$$< f, g>_{\mathbf{\omega}} = \bar{f}^{tr}[\mathbf{\omega}]g$$

Hilbert Space

$$\langle \psi, \psi \rangle = 1$$

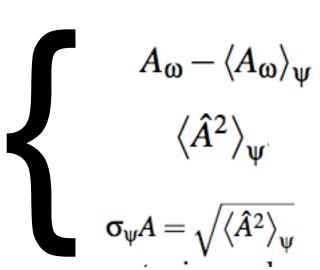
States

$$\langle \psi, A \phi \rangle = \langle A \psi, \phi \rangle$$

Measurements

$$\langle \psi, A \psi \rangle_{\omega} = \langle A \rangle_{\psi}$$

Expectation



Deviation

Why does this work?

The Uncertainty Principle for Markov Chains Lemma:

$$\left(\sigma_{\psi}C_{\omega}\right)\left(\sigma_{\psi}S_{\omega}\right)\geq\left|rac{1}{2}\left\langle\left[\Delta_{\omega},\Delta_{\omega}^{*}\right]
ight
angle_{\psi}
ight|$$

proof: Notice

$$[C_{\omega}, S_{\omega}] = C_{\omega}S_{\omega} - S_{\omega}C_{\omega} = i[\Delta_{\omega}, \Delta_{\omega}^*],$$

SO

$$\begin{vmatrix} \frac{1}{2} \left\langle \left[\Delta_{\omega}, \Delta_{\omega}^{*} \right] \right\rangle_{\psi} \middle| = \begin{vmatrix} \frac{1}{2} \left\langle \left[C_{\omega}, C_{\omega} \right] \right\rangle_{\psi} \middle| & \text{(def. plus foil)} \\ = \begin{vmatrix} \frac{1}{2} \left\langle \left[\hat{C}_{\omega}, \hat{S}_{\omega} \right] \right\rangle_{\psi} \middle| & \text{(differ by constants)} \\ = \frac{1}{2} \left| \left\langle \psi, \hat{C}_{\omega} \hat{S}_{\omega} - \hat{C}_{\omega} \hat{S}_{\omega} \psi \right\rangle \middle| & \text{(definition)} \\ = \frac{1}{2} \left| \left\langle \hat{C}_{\omega} \psi, \hat{S}_{\omega} \psi \right\rangle - \left\langle \hat{S}_{\omega} \psi, \hat{C}_{\omega} \psi \right\rangle \middle| & \text{(self adjoint)} \\ = \left| Im \left\langle \hat{C}_{\omega} \psi, \hat{S}_{\omega} \psi \right\rangle \middle| & \text{(Hemitian Inner-product)} \\ \leq \left| \left\langle \hat{C}_{\omega} \psi, \hat{S}_{\omega} \psi \right\rangle \middle| & \text{(Partial inner-product)} \\ \leq \left| \left\langle \hat{C}_{\omega} \psi, \hat{C}_{\omega} \psi \right\rangle \sqrt{\left\langle \hat{S}_{\omega} \psi, \hat{S}_{\omega} \psi \right\rangle} & \text{(Cauchy-Schwarz inequality)} \\ = \sqrt{\left\langle \psi, \hat{C}_{\omega}^{2} \psi \right\rangle} \sqrt{\left\langle \psi, \hat{S}_{\omega}^{2} \psi \right\rangle} & \text{(Self adjoint)} \\ = \left(\sigma_{\psi} C_{\omega} \right) \left(\sigma_{\psi} S_{\omega} \right) & \text{(defintion)} \end{aligned}$$

Q.E.D