

A Growing Neural Gas Network Learns Topologies

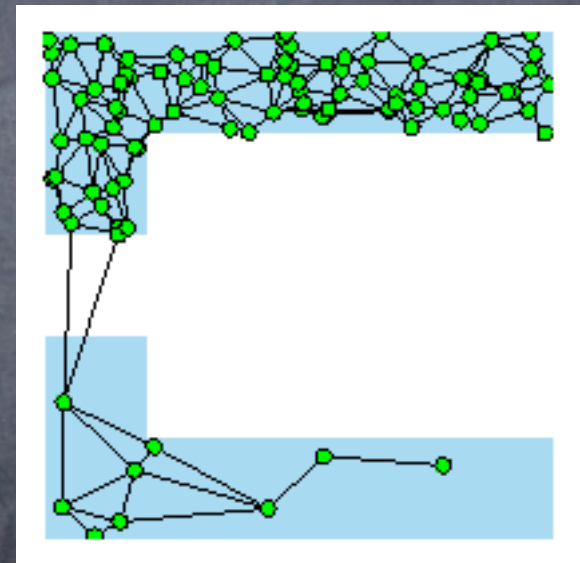
Presentation by Jonathan Beall
on the paper by Bernd Fritzke

Growing Neural Gas

- Adjust units and connections to approximate input data topology.
- Unsupervised learning
 - Dimensionality reduction
 - Topology learning

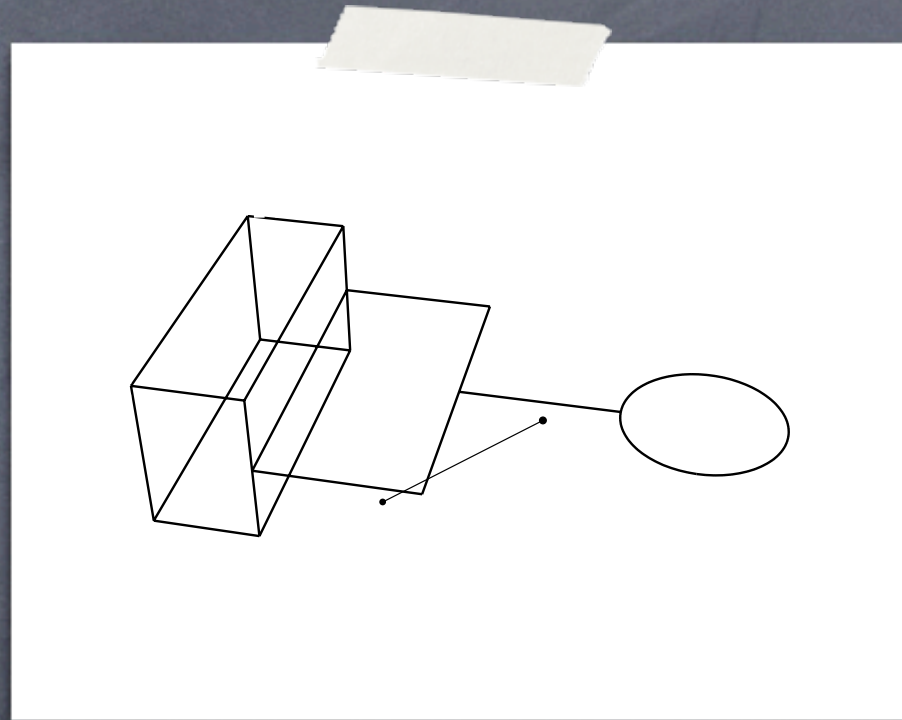
Competitive Hebbian Learning (CHL)

- For each input signal, connect the closest two centers.
- Can add in aging to eliminate now-irrelevant connections
- Creates a subgraph of the Delaunay triangulation

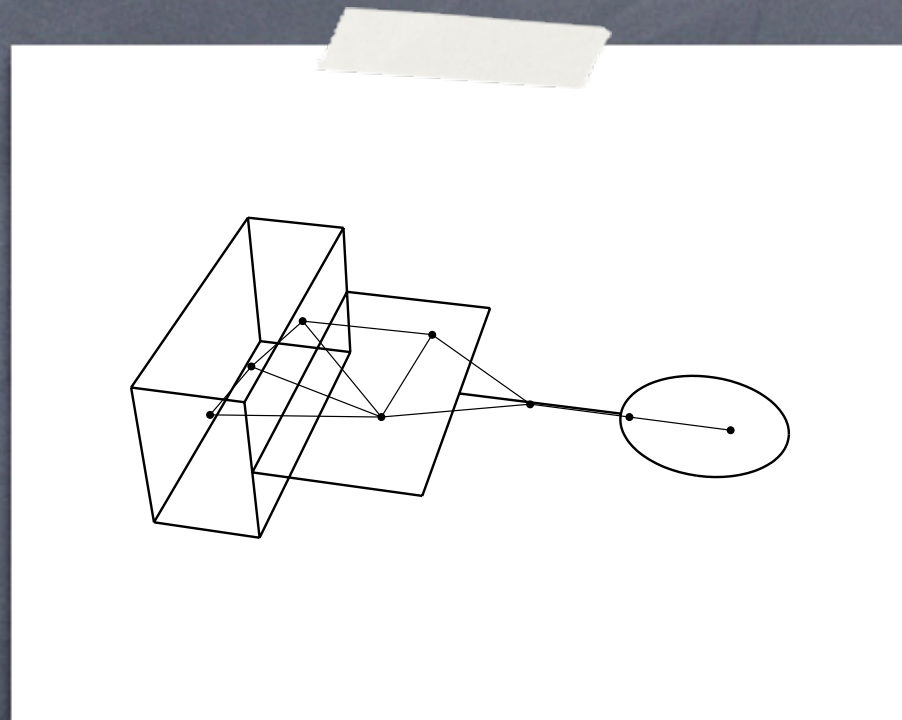


CHL and Neural Gas

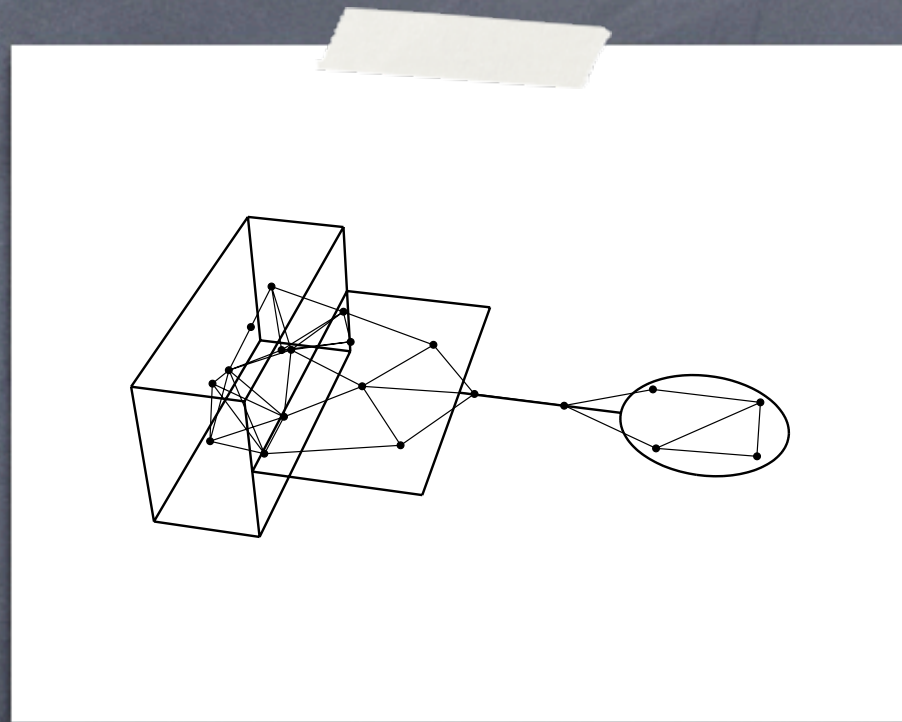
- Place all centers
- For each input
 - Perform CHL
 - Move the k closest centers toward the input
- k decays over time—annealing.



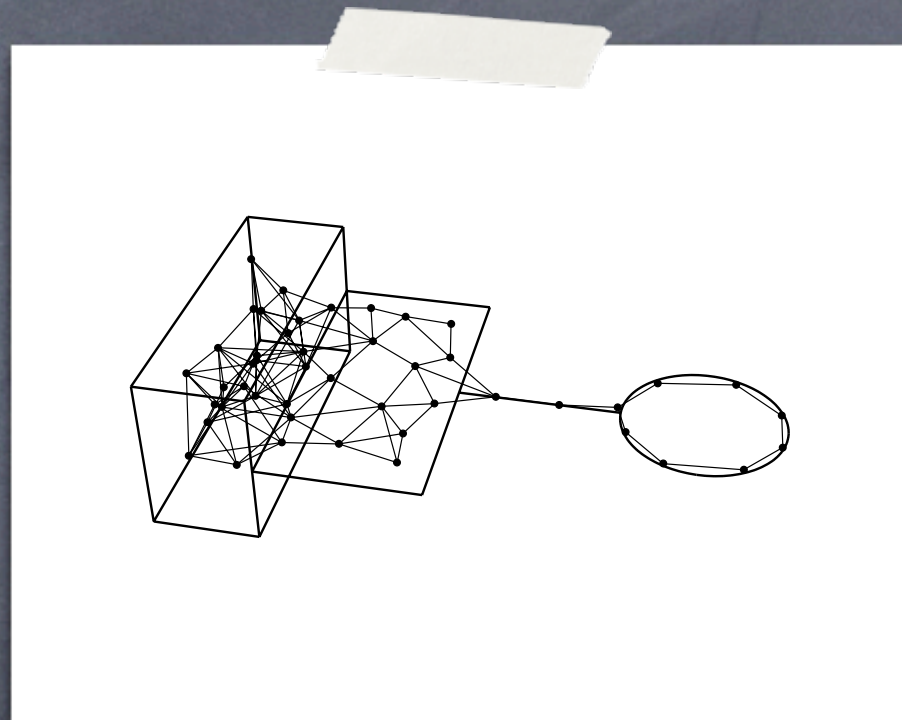
GNG in Action



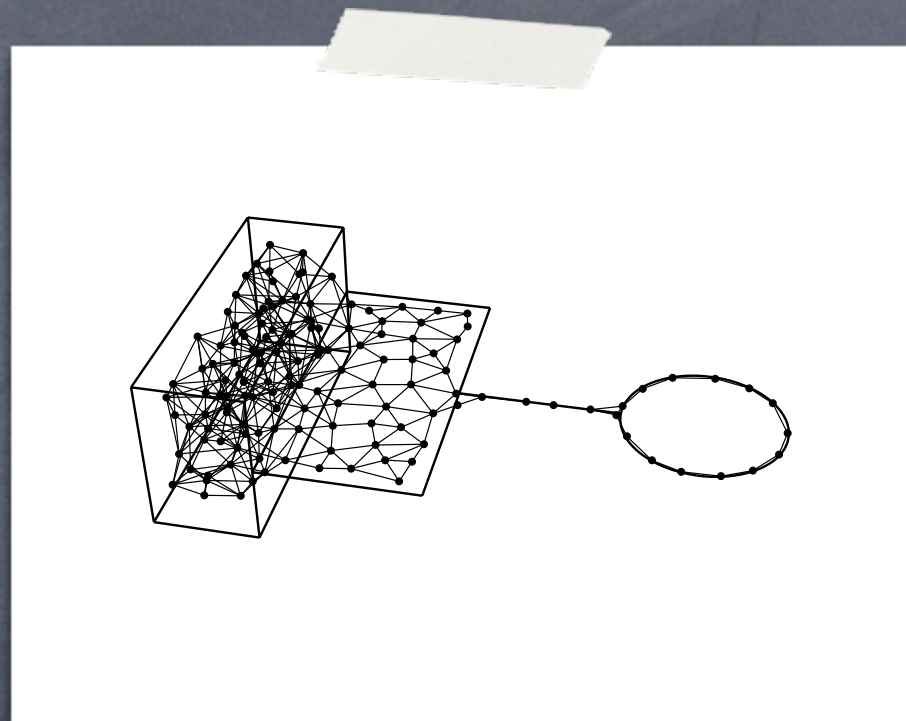
GNG in Action



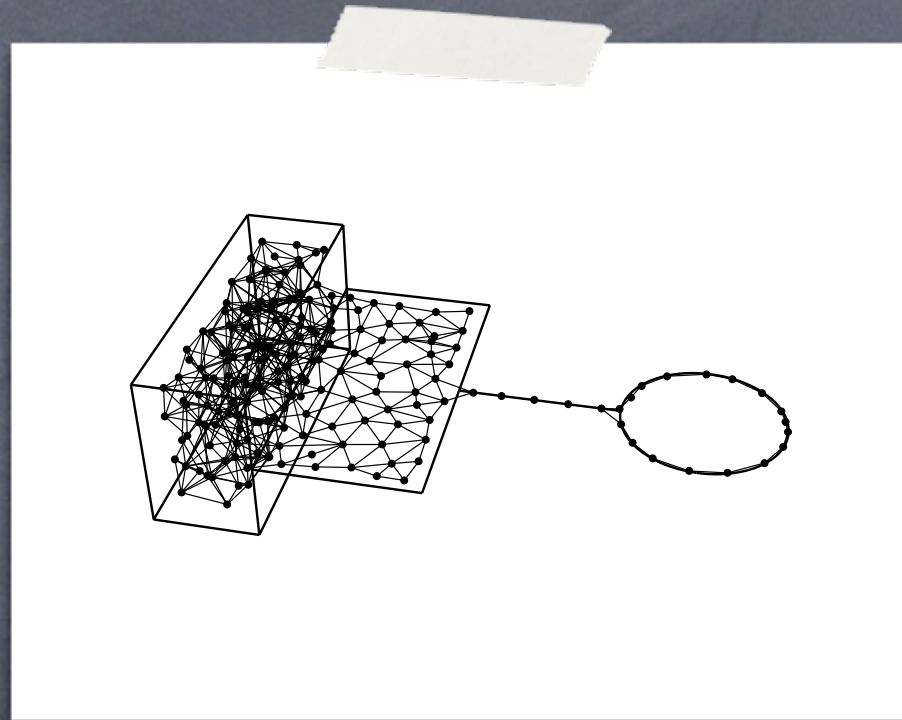
GNG in Action



GNG in Action



GNG in Action



GNG in Action

Growing Neural Gas Algorithm

- Start with 2 random units
- Generate a random signal ξ and find two nearest units, s_1 and s_2 .
- Increment the age of all edges coming from s_1 .
- Add the squared distance $\|s_1 - \xi\|^2$ to s_1 's error counter.

- Move s_1 and its direct neighbors s_n towards ξ by fractions ϵ_b and ϵ_n , respectively, of the total distance.
- If s_1 and s_2 are connected, set the age of the edge connecting them to 0, otherwise create an edge between them.
- Remove any edges with ages larger than a_{\max} . If this leaves any units with no edges, delete them.

- Every λ input signals, insert a new unit:
 - Find the unit q with the maximum error.
 - Find q 's neighbor f with the largest error.
 - Insert a new unit r between q and f
 - Decrease the errors of q and f by multiplying them by a constant α , and then set the error of r to q 's new error.

- Decrease all error variables by multiplying them by a constant d .
- If the network is not yet complete (size, performance...), begin again by generating a new signal, ξ .

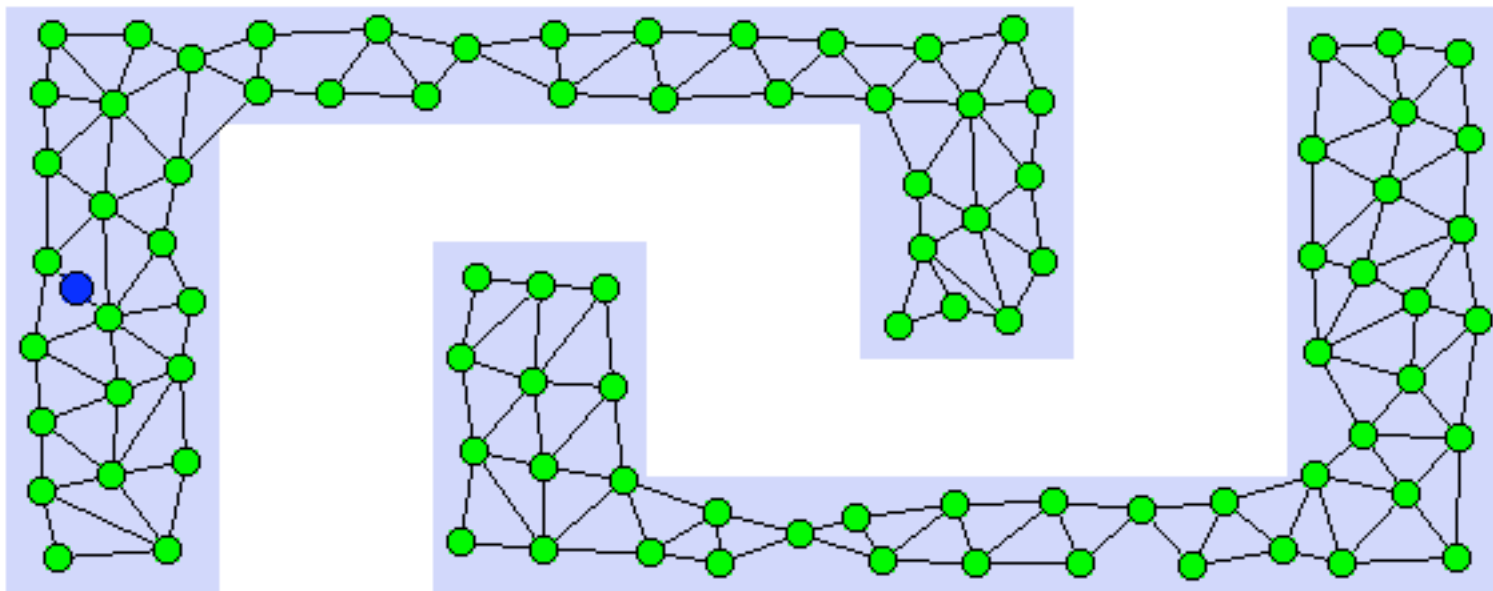
GNG Applet

<http://www.neuroinformatik.ruhr-uni-bochum.de/ini/VDM/research/gsn/JavaPaper/>

Network Model: **Growing Neural Gas / GNG-U**

55000

v1.5



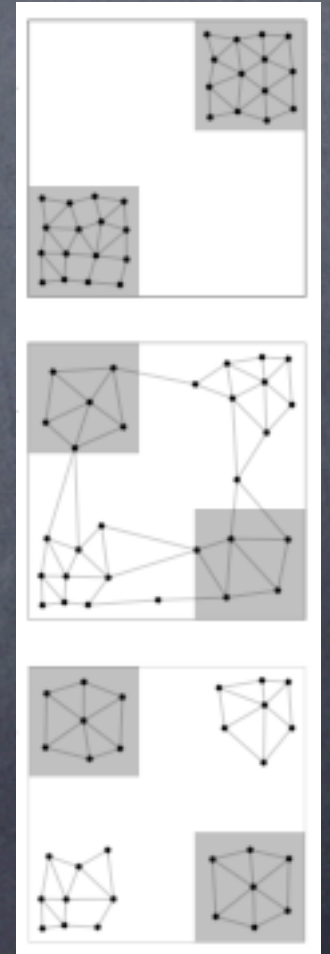
93 Nodes

Benefits of GNG

- Variable dimensionality
- Constant parameters
- No set termination time
- Soft competition helps skip over local minima
- Insertion based on batch accumulated errors helps to avoid overfitting
- Efficient—time increases linearly with the number of links

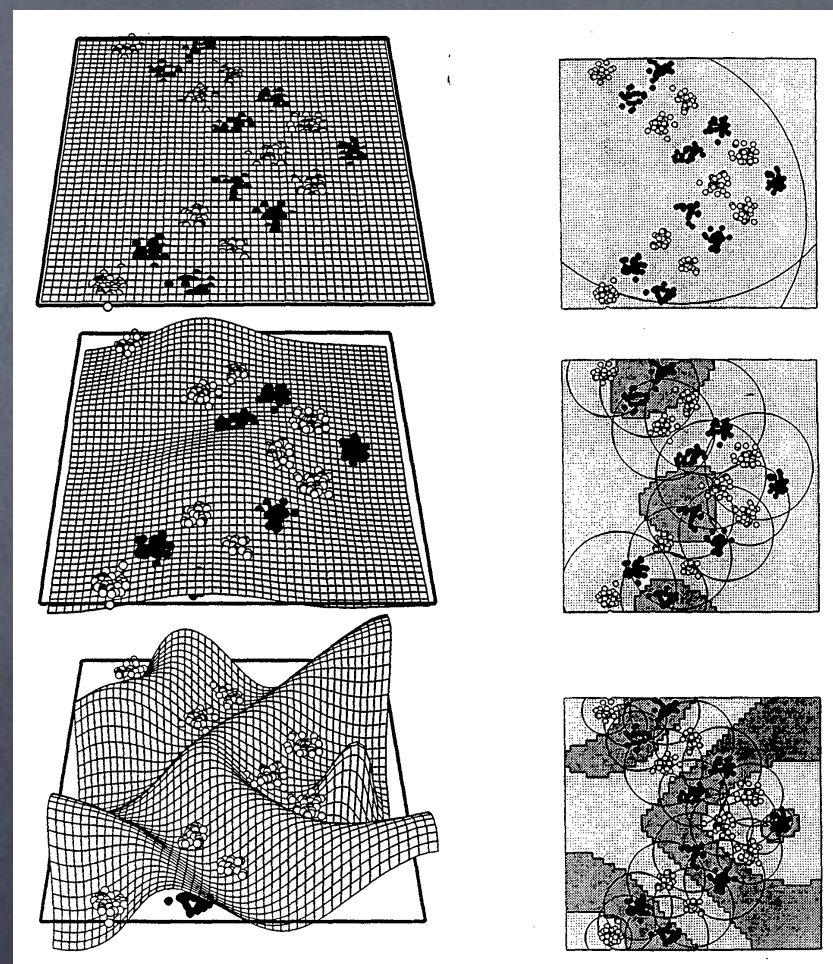
Drawbacks of GNG

- No cooling schedule, so no long term stability
- 7 parameters must be set
- CHL is not batch-driven, is sensitive to noise
- Can leave stranded units when exposed to rapidly changing data (GNG-U)
- Cannot visualize higher dimension GNGs easily.



Applications of GNG

- Incremental radial basis function networks
- 2-stage hybrid classifier (Baraldi and Blonda)
- Stability assessment of electric power systems (Rehtans and Leder)



Sources

- Baraldi and Blonda. "A Survey of Fuzzy Clustering Algorithms for Pattern Recognition—Part II." IEEE Transactions on Systems, Man, and Cybernetics, 1999.
- Fritzke. "A Growing Neural Gas Network Learns Topologies." Advances in Neural Information Processing Systems, 1994.
- Fritzke. "A self organizing network that can follow non-stationary distributions." Proceedings of ICANN, 1997.
- Martinetz and Schulten. "A 'Neural-Gas' Network Learns Topologies." Artificial Neural Networks, 1991.
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