

Phase transitions in network growth

V. Nicosia

¹School of Mathematical Sciences, Queen Mary University of London, London, UK

Sept. 29th 2015

Outline

- 1 Phase transitions in street patterns
- 2 Phase transitions in neural networks
- 3 Summing up

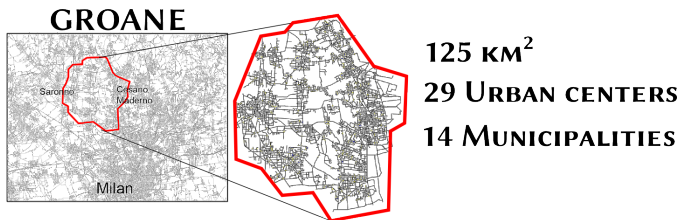
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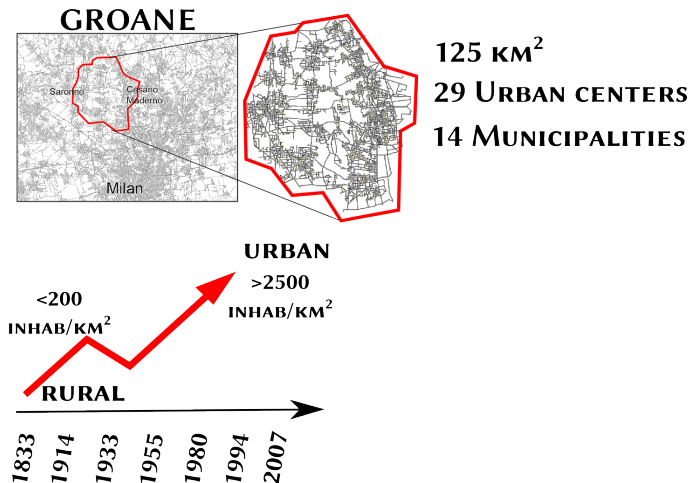
(Sub-)Urban street patterns



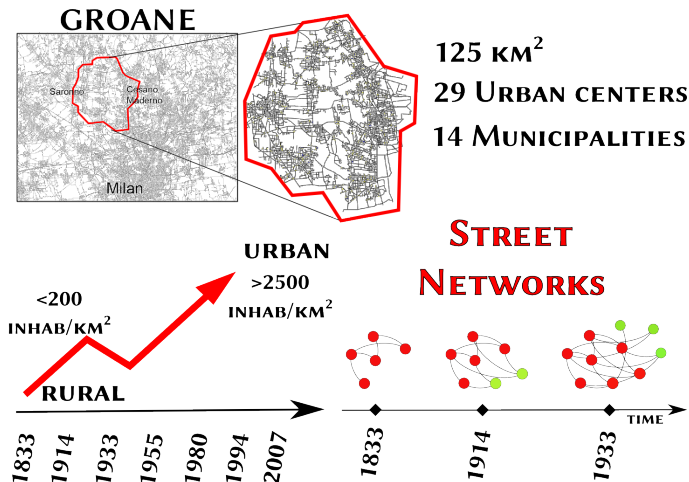
(Sub-)Urban street patterns



(Sub-)Urban street patterns

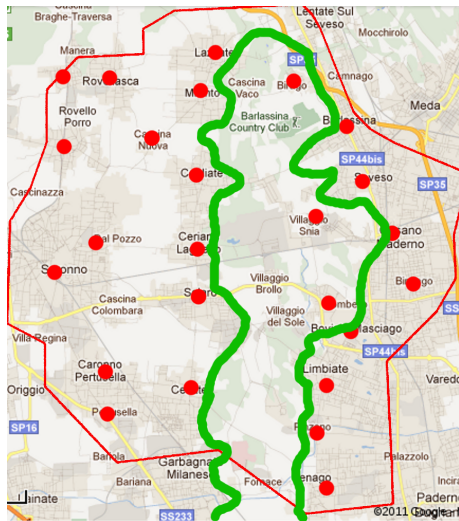


(Sub-)Urban street patterns



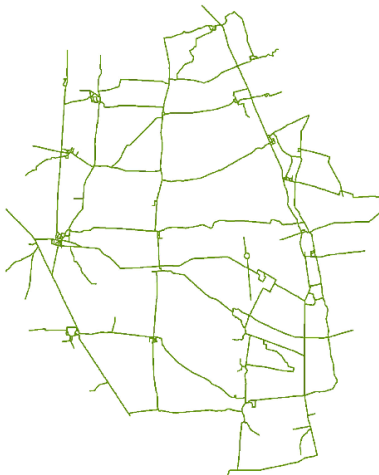
No centralised urban planning

- 29 urban centers within 14 municipalities



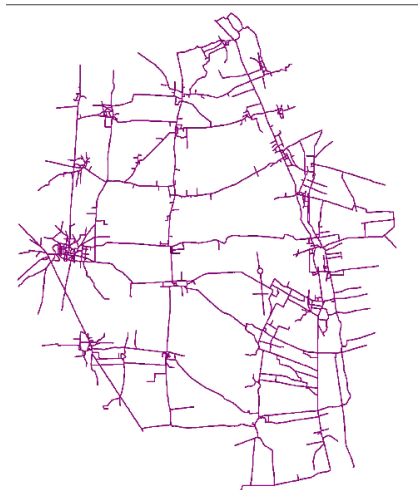
Shadows of a lost empire

- Backbone: Roman and Medieval roads.



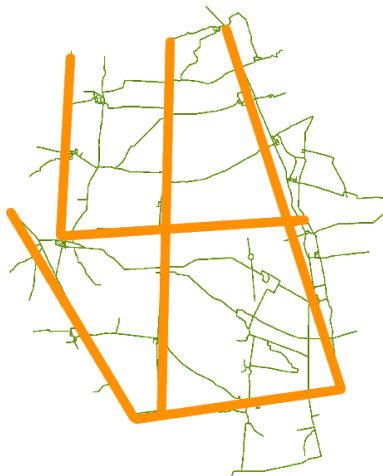
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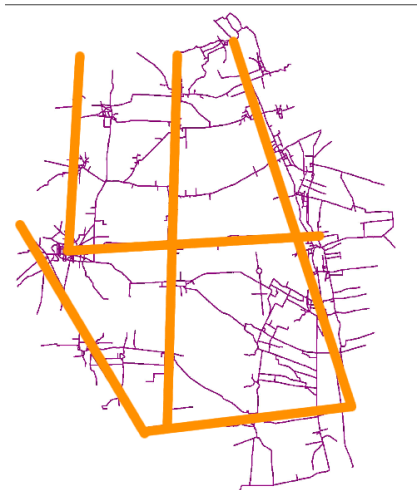
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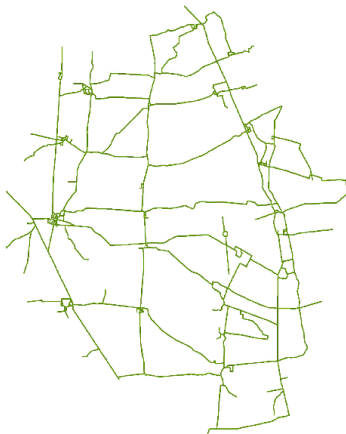
Shadows of a lost empire

- Backbone: Roman and Medieval roads.



New towns just out of the blue

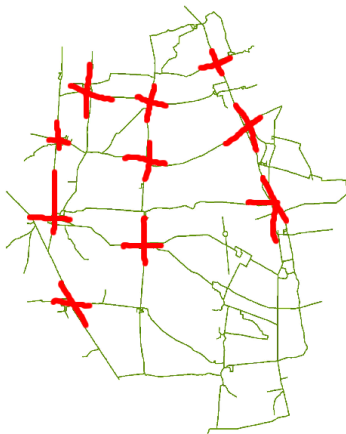
- New villages and towns appear and grow around crossings of old roads.



1833

New towns just out of the blue

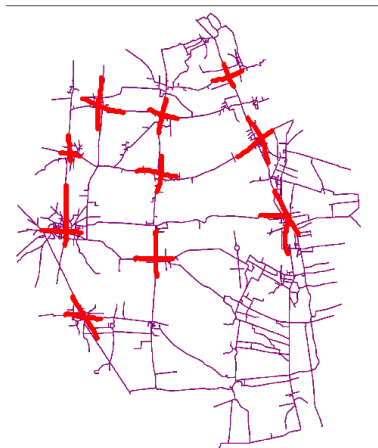
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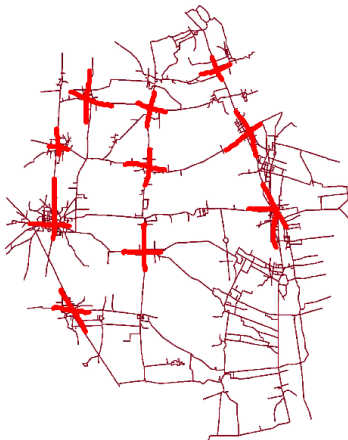
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1914

New towns just out of the blue

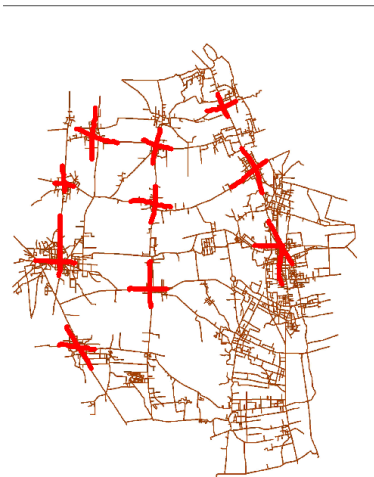
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1933

New towns just out of the blue

- New villages and towns appear and grow around crossings of old roads.



1955

New towns just out of the blue

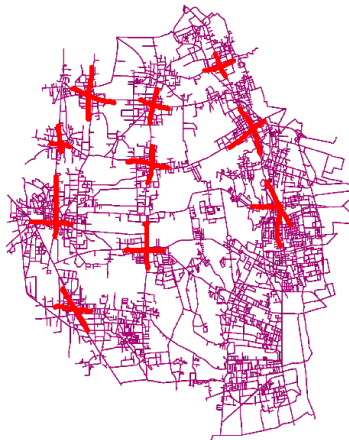
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1980

New towns just out of the blue

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1994

New towns just out of the blue

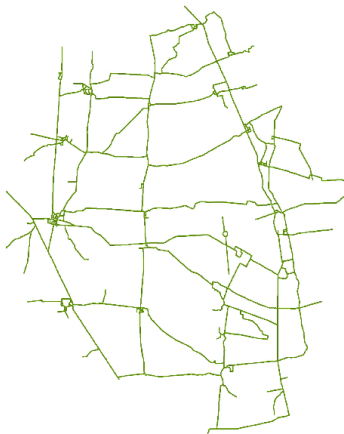
- New villages and towns appear and grow around crossings of old roads.



2007

The big squeeze

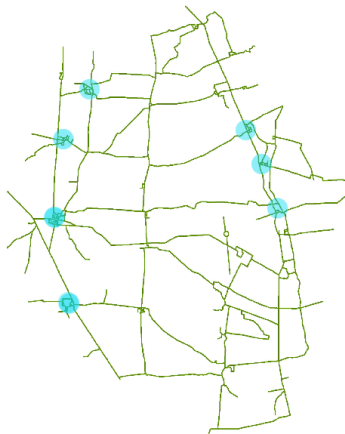
- Small villages become larger and are merged together into a compact urban area



1833

The big squeeze

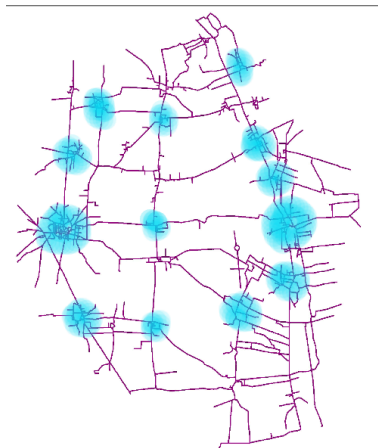
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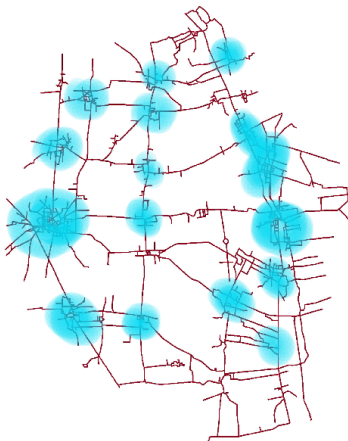
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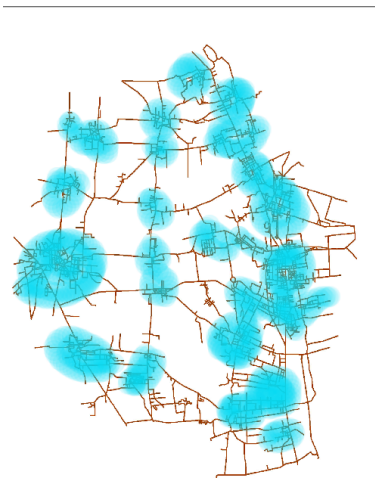
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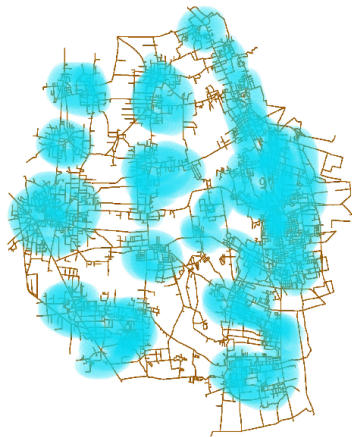
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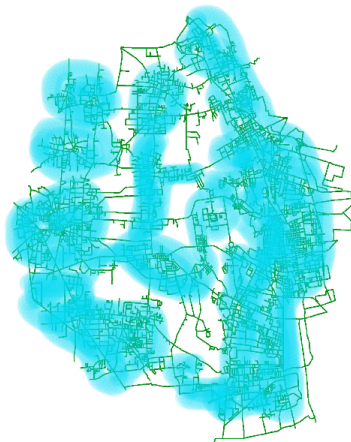
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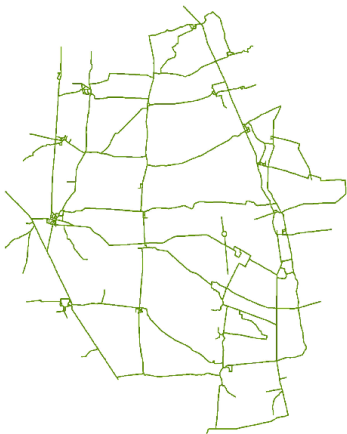


2007

Denser $\implies$ More fragmented

- The land is subsequently fragmented into smaller cells
- The shape and size of some cells remain constant over time

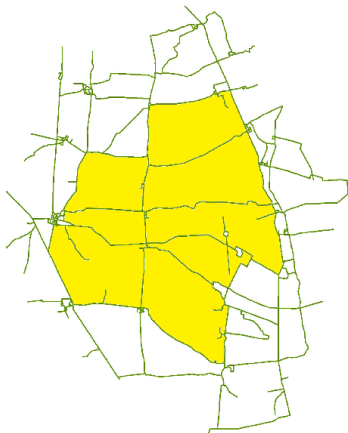
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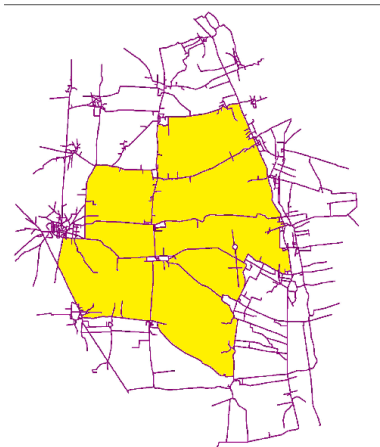
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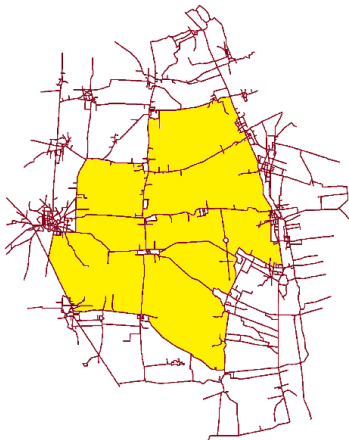
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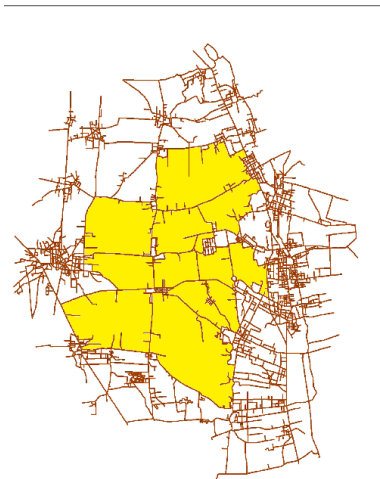
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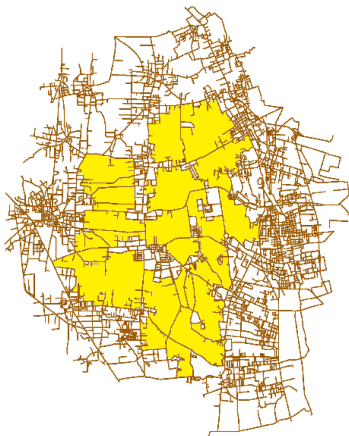
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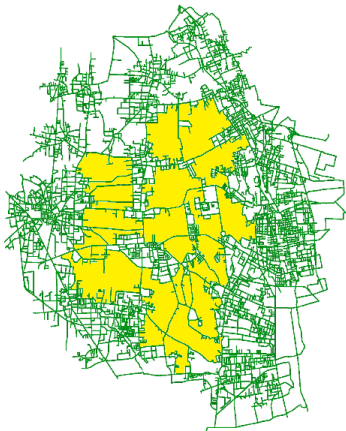
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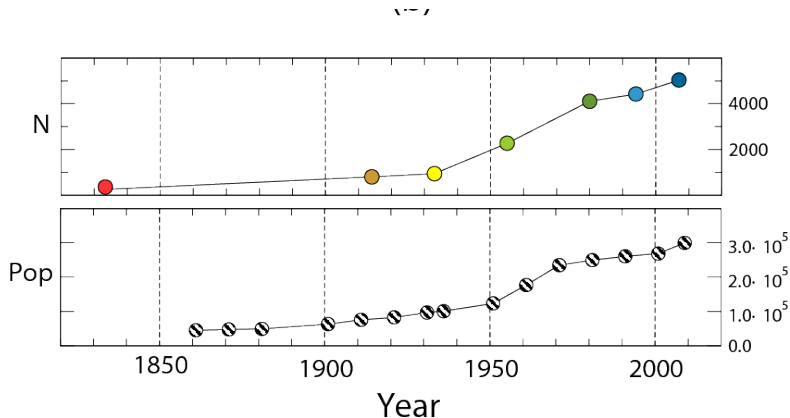
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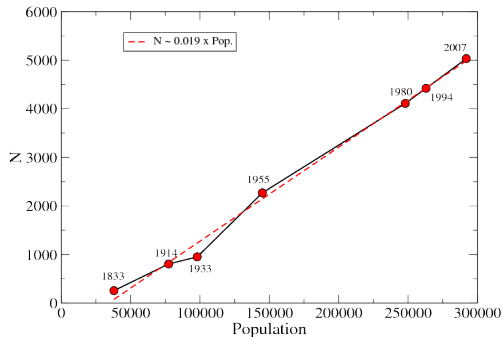
2007

- The land is subsequently fragmented into smaller cells
- The shape and size of some cells remain constant over time

A quantitative approach (1)

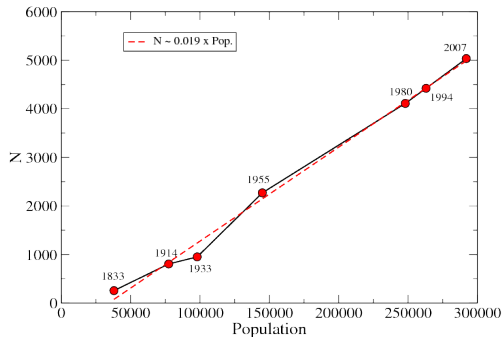


A quantitative approach (1)



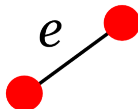
- N grows linearly with the number of inhabitants

A quantitative approach (1)



- N grows linearly with the number of inhabitants
- There are on average 51 inhabitants per node (!)

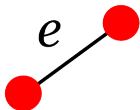
A quantitative approach (2)



EDGE BETWEENNESS

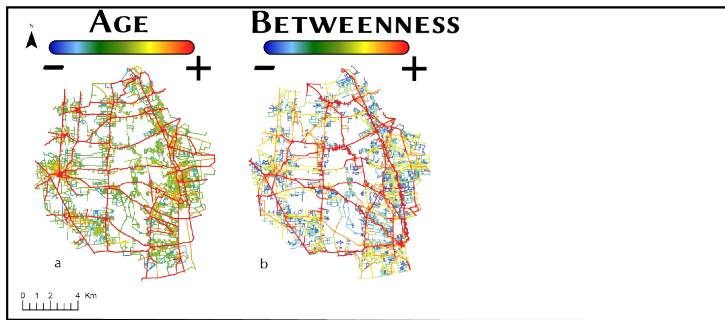
$$b(e) = \frac{\text{\# of shortest paths through } e}{\text{total \# of shortest paths}}$$

A quantitative approach (2)

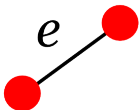


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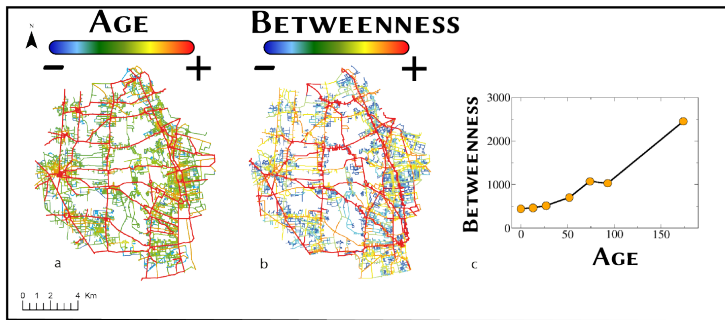


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A quantitative approach (3)

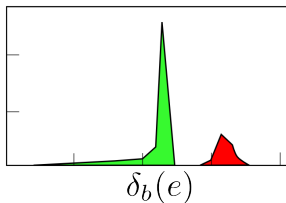
IMPACT ON
BETWEENNESS

$$\delta_b(e) = \frac{[\bar{b}(G_t) - \bar{b}(G_t \setminus \{e\})]}{\bar{b}(G_t)}$$

A quantitative approach (3)

IMPACT ON
BETWEENNESS

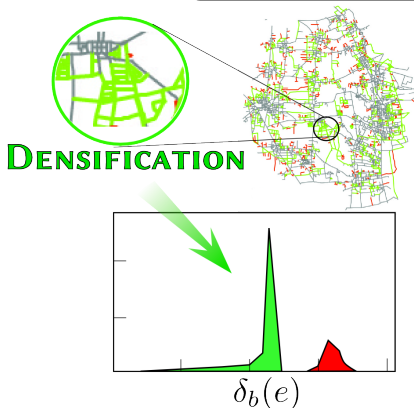
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A quantitative approach (3)

IMPACT ON
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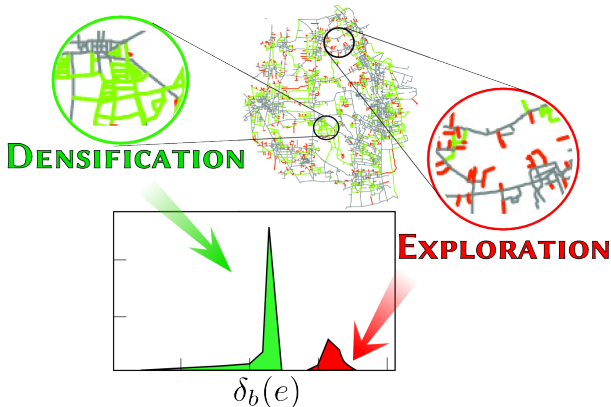
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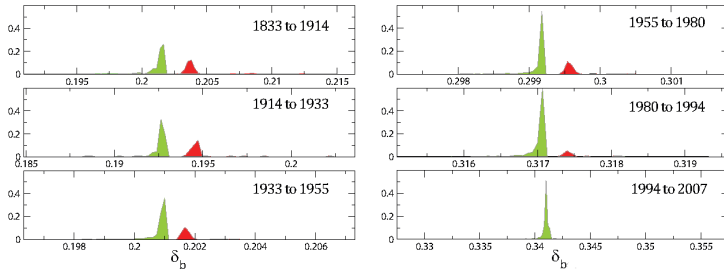


A quantitative approach (3)

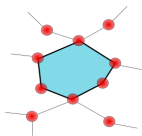
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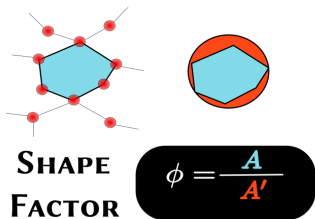


A quantitative approach (4)

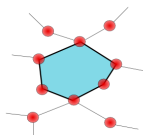


E. Strano, V. Nicosia, V. Latora, S. Porta, M. Barthélemy, *Nature Sci. Rep.* 2, 296 (2012)

A quantitative approach (4)



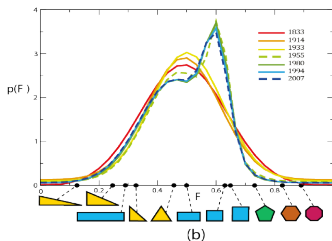
A quantitative approach (4)



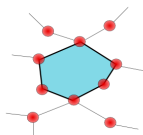
**SHAPE
FACTOR**



$$\phi = \frac{A}{A'}$$

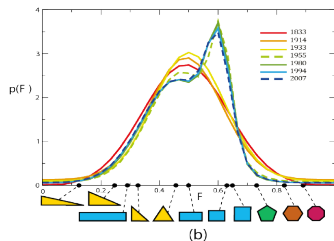


A quantitative approach (4)



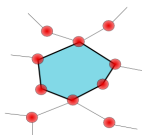
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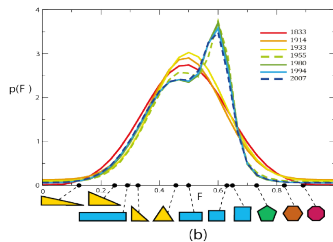
(c)

A quantitative approach (4)



**SHAPE
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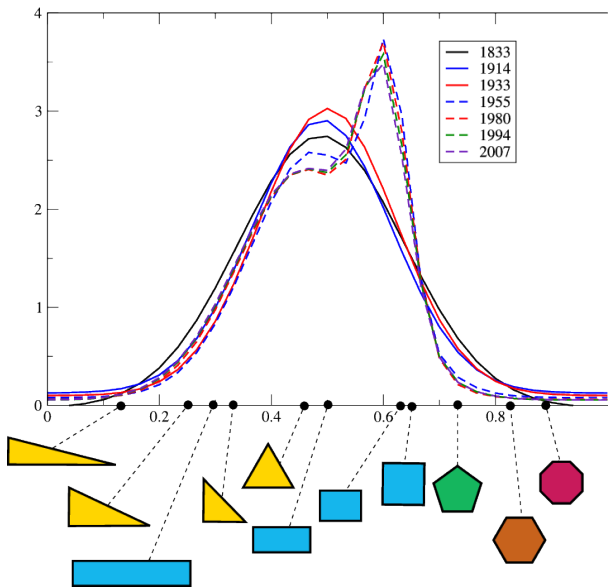
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(c)



(d)



Outline

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- 2 Phase transitions in neural networks
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Animal neural networks

C. ELEGANS



V. Nicosia, P. Vertes, W. Schafer, V. Latora, E. Bullmore, *Proc. Natl. Acad. Sci. USA* 110 (19), 7880-7885 (2013)

Animal neural networks

C. ELEGANS



NEURAL NETWORK

~300 NEURONS

(NODES)

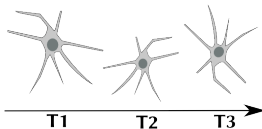
~2400 SYNAPSES

(EDGES)



Animal neural networks

C. ELEGANS



NEURAL NETWORK

~300 NEURONS

(NODES)

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Animal neural networks

C. ELEGANS



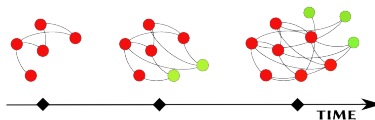
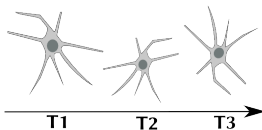
NEURAL NETWORK

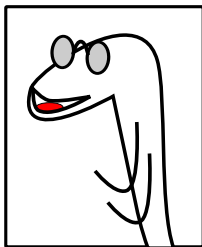
~300 NEURONS

(NODES)

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(EDGES)





Name: Caenorhabditis

Surname: Elegans

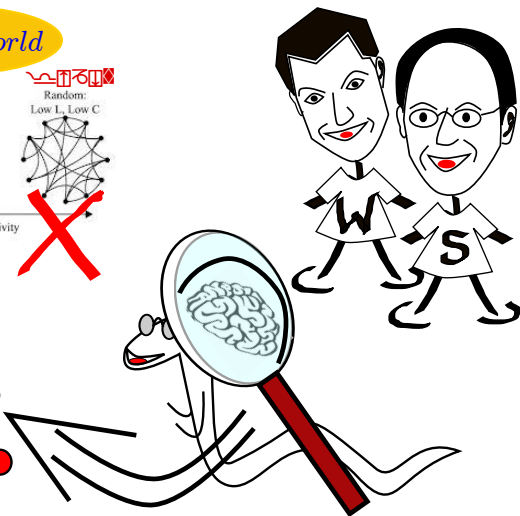
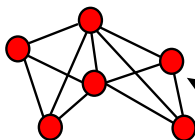
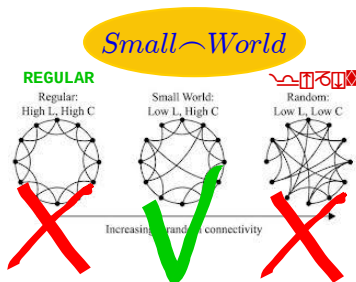
Length: 1.1 mm

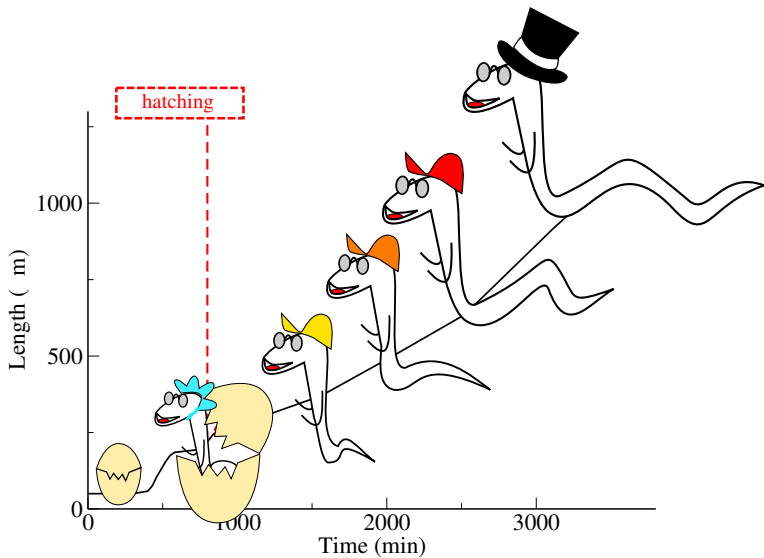
Occupation: lab worm

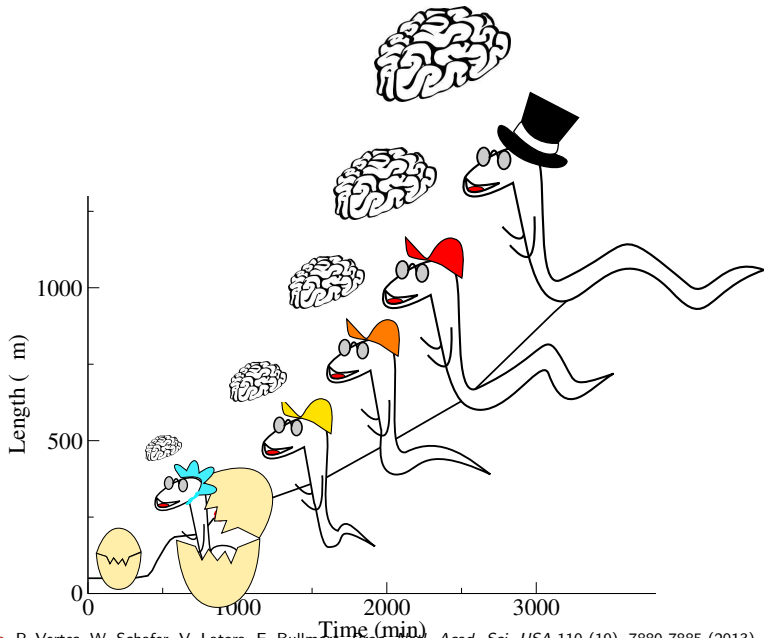
Particular features: transparent, blind

Abilities: Moving , eating, having sex
with just about 300 neurons

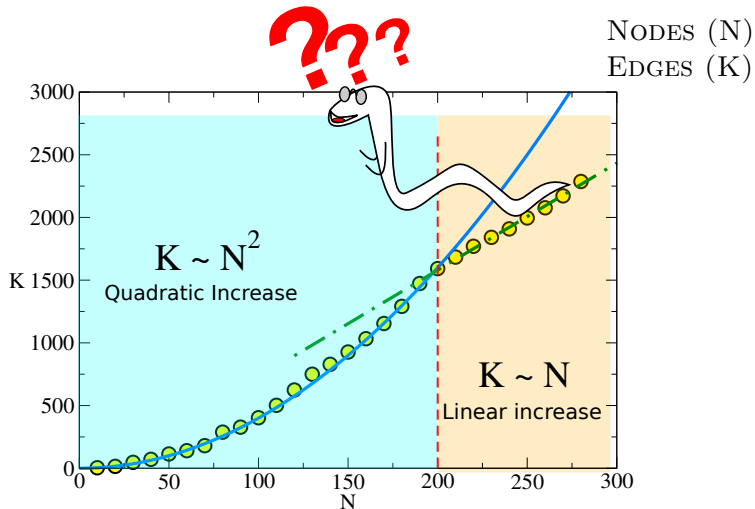
Featured on: Nature, Science, PNAS, PRL

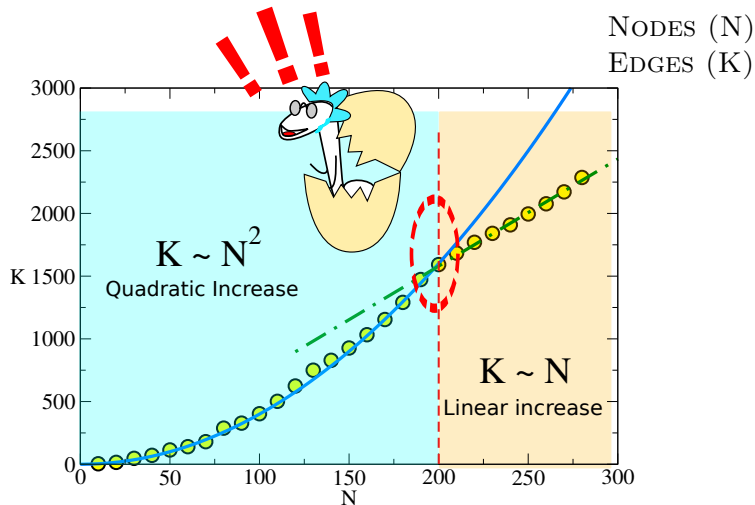


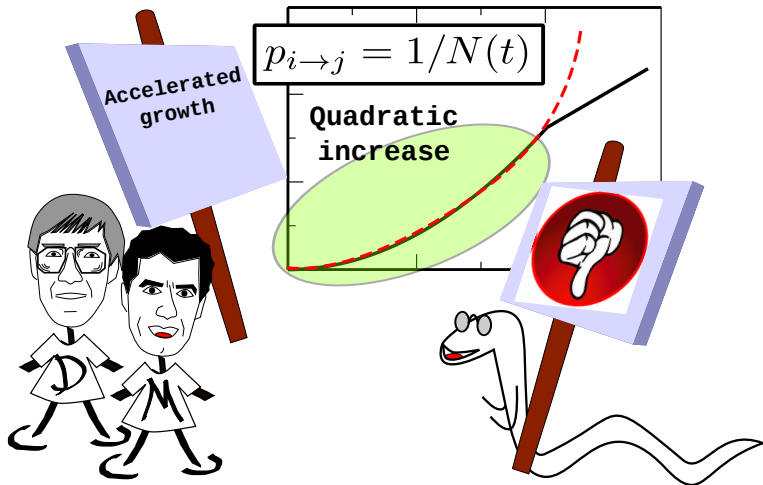


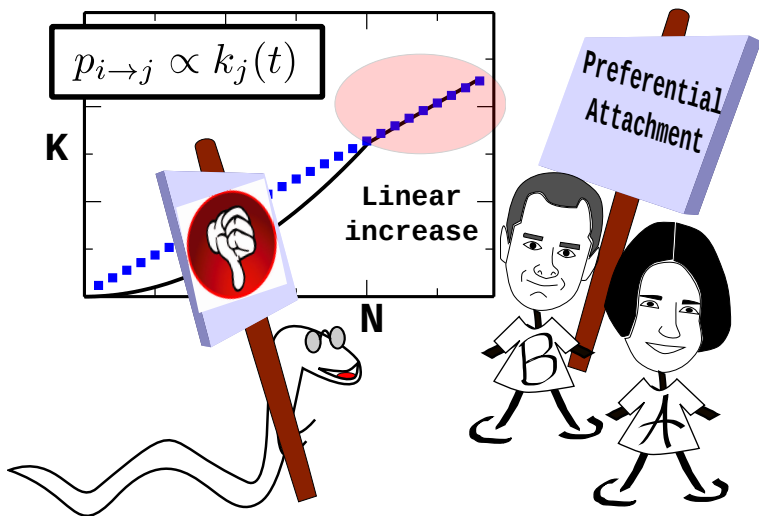


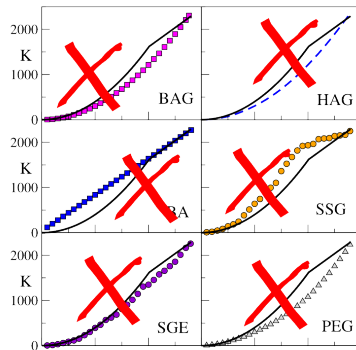
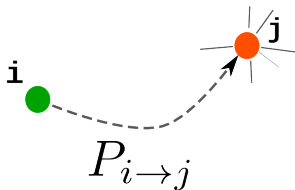
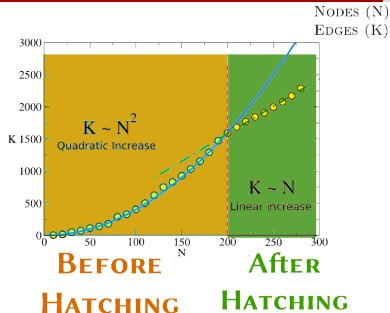
V. Nicosia, P. Vertes, W. Schafer, V. Latora, E. Bullmore, *Proc. Natl. Acad. Sci. USA* 110 (19), 7880-7885 (2013)



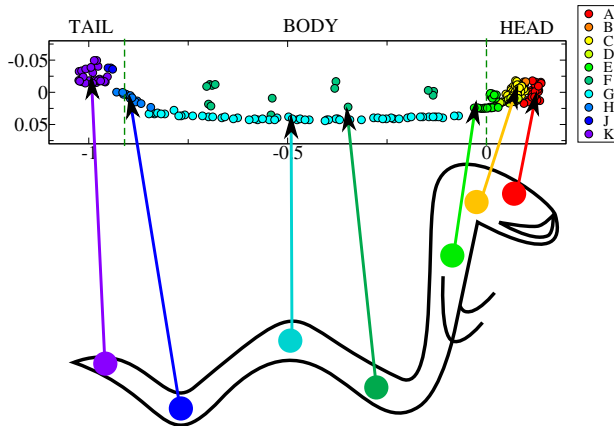




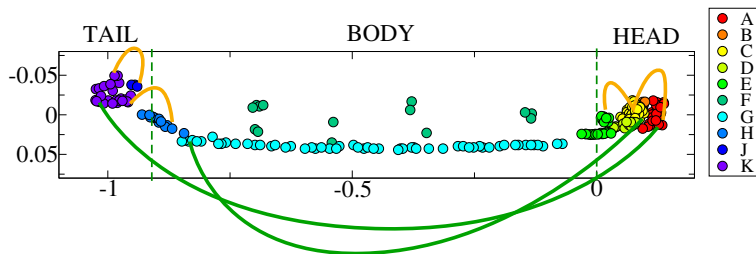




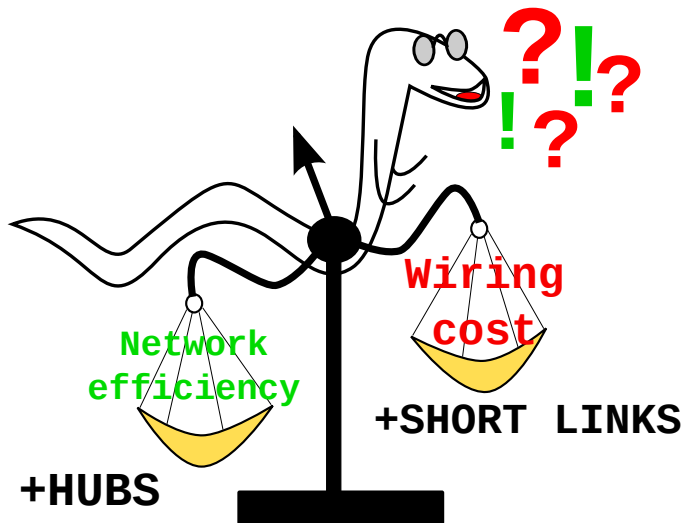
The brain is spread throughout the body!

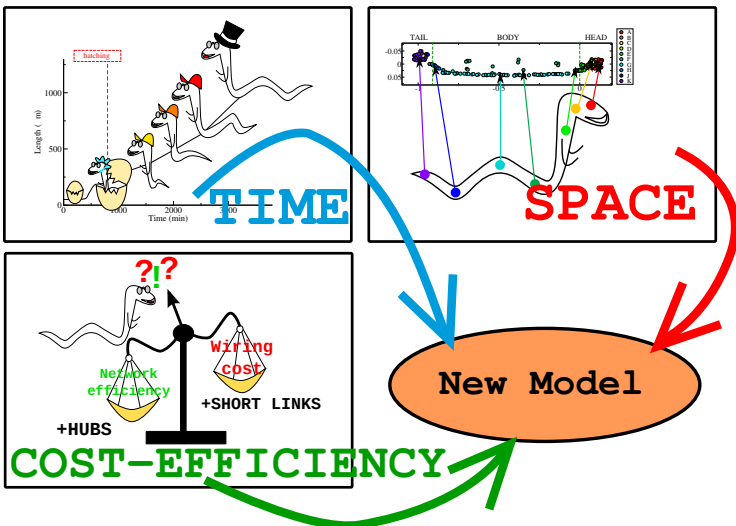


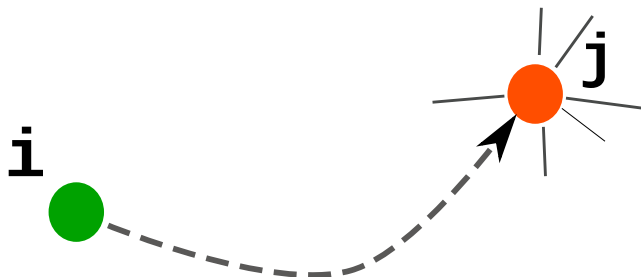
Short-range connections = small wiring cost



Long-range connections (to hubs) = high efficiency

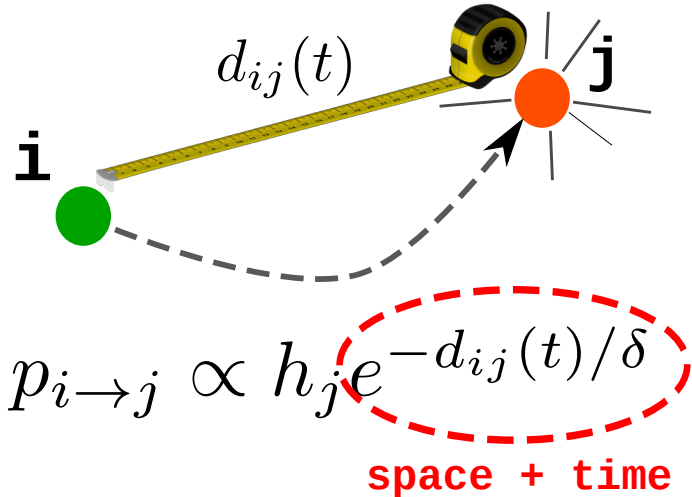


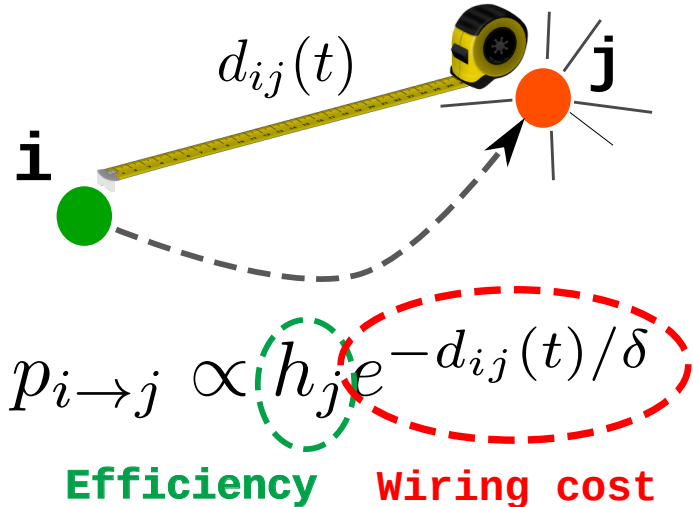




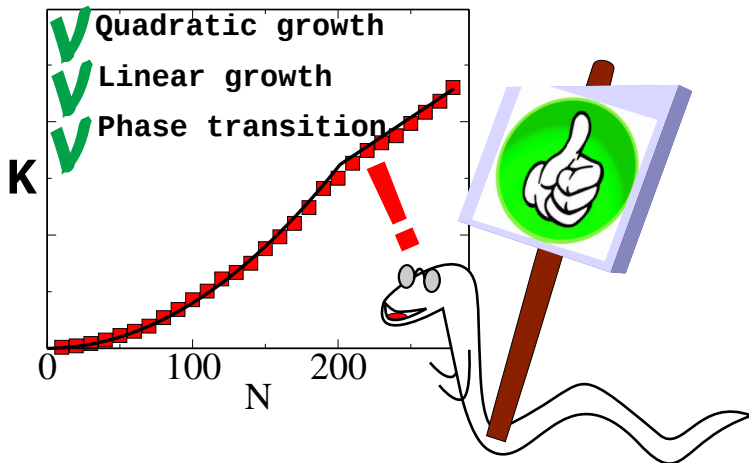
$$p_{i \rightarrow j} \propto h_j e^{-d_{ij}(t)/\delta}$$

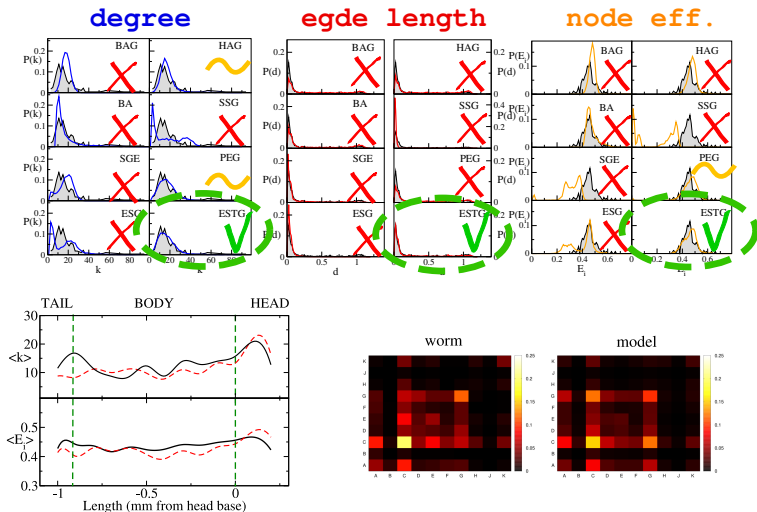
degree in the adult worm





Economical Spatio-temporal growth





Outline

- 1 Phase transitions in street patterns
- 2 Phase transitions in neural networks
- 3 Summing up



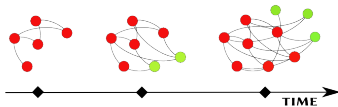
GROANE



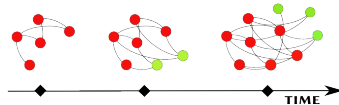
C.ELEGANS



GROANE

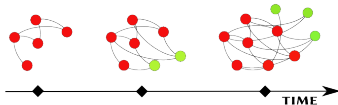


C.ELEGANS





GROANE



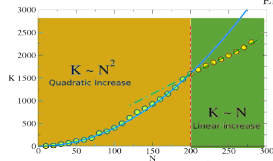
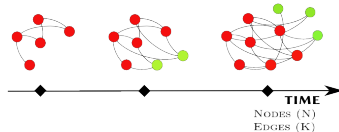
BEFORE WWII



AFTER WWII



C.ELEGANS



**BEFORE
HATCHING**

**AFTER
HATCHING**

References

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