

Welcome back...

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Test out

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Don't worry.

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Don't worry. Be happy.



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Look at instructions.

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Try to get it in then or soon after!

Pareto:

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20% of pods have 80% of peas.

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$i$ th largest city has population  $\frac{P_1}{i}$ .



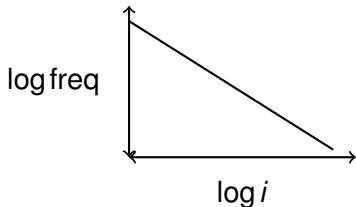
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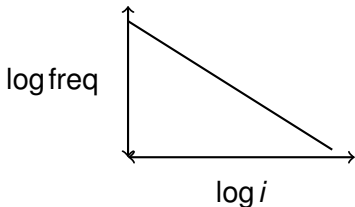
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Zipf's law.

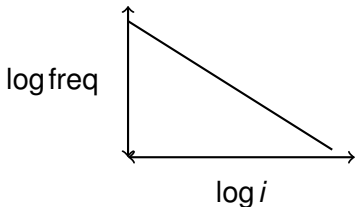
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Zipf's law. Zipf's graph.

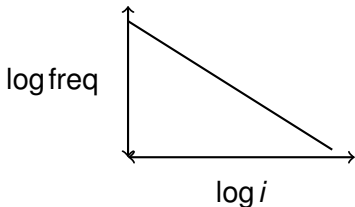
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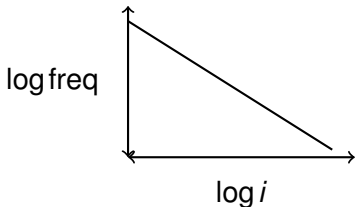
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See Adamic for comment on estimating for real data.

<http://www.hpl.hp.com/research/idl/papers/ranking/ranking.html>

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**MAKE SOME DRAWINGS.**

# Pareto to Zipf

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$$x_i = \frac{1}{i^{1/(1-\alpha)}}$$

$$\text{Relationship: } \beta = \frac{1}{1-\alpha}$$



Self similarity.

Power laws.

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Actually  $\gamma_t \approx (1 + \beta/t)$ .

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$$x_{t+1} = x_t \times \gamma.$$

Actually  $\gamma_t \approx (1 + \beta/t)$ .

Roughly constant for interval of width  $\beta$ .

# Power law and philosophy.

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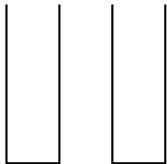
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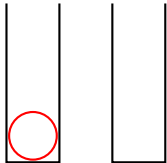
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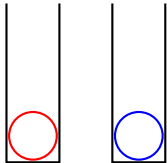
# Polya Urns



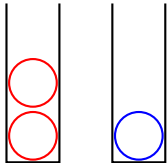
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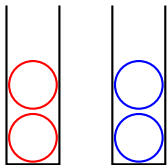
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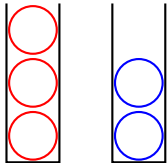
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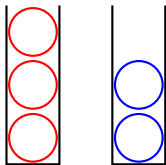
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Choose bin uniformly at random.

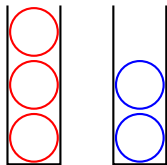


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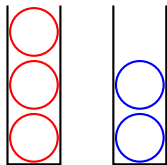
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Load on red bin?

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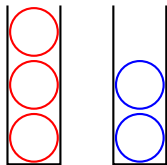
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Expectation?

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Expectation?  $n/2$

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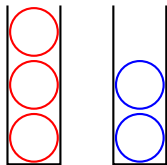
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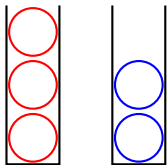
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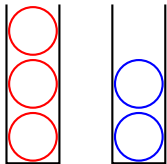
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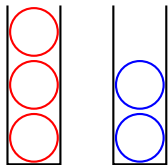
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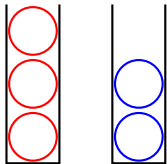
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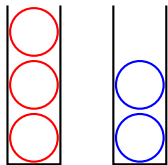
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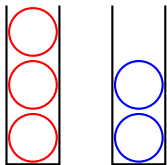
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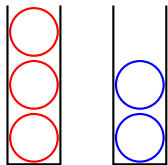
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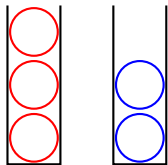
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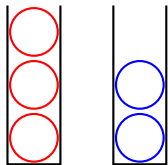
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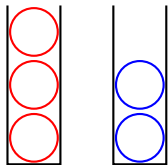
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Start with two balls, insert  $n$  more.

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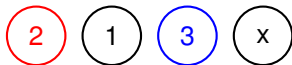
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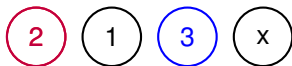
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Where is ball 1?

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Where is ball 1? Position 4.

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Start with two balls, insert  $n$  more.



Where is ball 1? Position 4.

How many red balls?



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Where is ball 1? Position 4.

How many red balls? 3.

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Random permutation. Position  $i$

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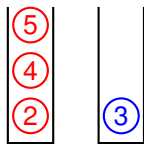
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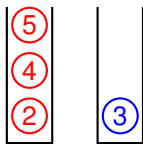
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Red balls have same distribution in two processes.

More bins.

$m$  bins.

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Average degree: 4

Max Degree?



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5% of nodes have degree greater than 20.

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Average degree: 12.

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Internet graph:

Average degree: 12.

Degree  $\geq 100$  with prob.  $\leq 10^{-6}$ .

Actual:

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Internet graph:

Average degree: 12.

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Actual: 1% greater than 100.

Some very large.

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Degrees too large for even that.

Internet: copy links.

Surf.

Internet: copy links.

Surf. Cool page.

Internet: copy links.

Surf. Cool page. Link for mine.



Internet: copy links.

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Model:

# Internet: copy links.

Surf. Cool page. Link for mine.

Model:

Pick a random neighbor.

# Internet: copy links.

Surf. Cool page. Link for mine.

Model:

- Pick a random neighbor.

- Copy all links.

# Internet: copy links.

Surf. Cool page. Link for mine.

Model:

- Pick a random neighbor.

- Copy all links.

Random Graph with average degree 4.

# Internet: copy links.

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Random Graph with average degree 4.

- Plus Copy process

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Plus Copy process  $\rightarrow \sqrt{n}$

# Routers.

Connection Game.

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Process Distance:



# Routers.

Connection Game.

Process Distance:

Arrive randomly at point on unit square.

# Routers.

Connection Game.

Process Distance:

Arrive randomly at point on unit square.

Connect to closest node.

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Generate tree with average degree 1.

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Max degree?

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Process Distance:

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Generate tree with average degree 1.

Max degree?  $O(\log n)$ .

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Process Hops:

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Process Distance/Hops:

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Connect to node with  $\min_{j < i} \alpha d_{ij} + h_j$ .

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Generate tree with average degree 1.

Max degree?  $O(\log n)$ .

Process Hops:

Arrive randomly at point on unit square.

Connect to first node.

Max degree?  $n - 1$ .

Process Distance/Hops:

Arrive randomly at point on unit square.

Connect to node with  $\min_{j < i} \alpha d_{ij} + h_j$ .

Power law if  $c \leq \alpha \leq \sqrt{n}$ ,

# Routers.

Connection Game.

Process Distance:

Arrive randomly at point on unit square.

Connect to closest node.

Generate tree with average degree 1.

Max degree?  $O(\log n)$ .

Process Hops:

Arrive randomly at point on unit square.

Connect to first node.

Max degree?  $n - 1$ .

Process Distance/Hops:

Arrive randomly at point on unit square.

Connect to node with  $\min_{j < i} \alpha d_{ij} + h_j$ .

Power law if  $c \leq \alpha \leq \sqrt{n}$ ,  $\rightarrow$  power law!

See you ...

Thursday.