

Optimising partner matching for microsimulations of the HIV epidemic

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% Optimising partner matching for microsimulations of the HIV epidemic % Nathan Geffen % SACEMA Research Day, 24 March 2015

Website for this talk

<http://nathangeffen.webfactional.com/partnermatching/partnermatching.html>

- ▶ Too much to cover in 10 minutes
- ▶ This presentation is bird's eye view
- ▶ Webpage has details
- ▶ Perhaps seminar

Aim

- ▶ Microsimulation of HIV epidemic gives rich insights
- ▶ We want:
 - ▶ Monte Carlo simulation to calculate confidence intervals
 - ▶ Convenient high-level language programming
 - ▶ Visualisations on the web using Javascript
- ▶ BUT: It is too slow
- ▶ Partner matching is the bottle-neck
- ▶ **Find faster ways of doing partner matching**

Methodology

- ▶ Define three partner matching algorithms
- ▶ Define two reference algorithms to compare these with:
 - ▶ Quality: One very slow producing nearly ideal matches
 - ▶ Speed: One very fast producing random matches
- ▶ Define measure of quality
- ▶ Analyse mathematically
- ▶ Compare empirically in multiple tests

Typical discrete time microsimulation

```
for each time-step
-----
  for each event E
    for each agent A
      if E should be applied to A
        apply E to A
-----
```

Algorithm efficiency

► $O(n)$ vs $O(n^2)$

Number of agents (n)	n	$(n^2-n)/2$
10	10	45
100	100	4,950
1,000	1,000	499,500
10,000	10,000	49,995,000
100,000	100,000	4,999,950,000
1,000,000	1,000,000	499,999,500,000

Graphically depicted

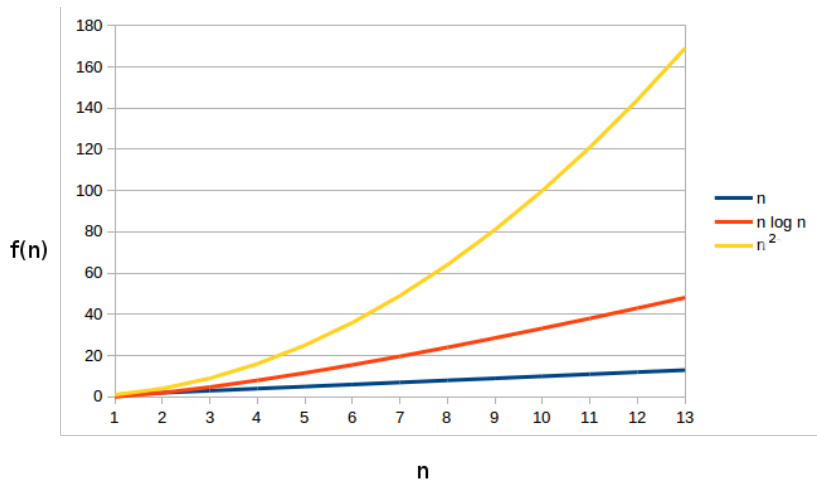


Figure : Three efficiency classes

Typical agent attributes for partner matching

- ▶ age
- ▶ sex
- ▶ desire for new partnership
- ▶ riskiness (including whether agent is a sex worker)
- ▶ relationship status (including whether agent is married)

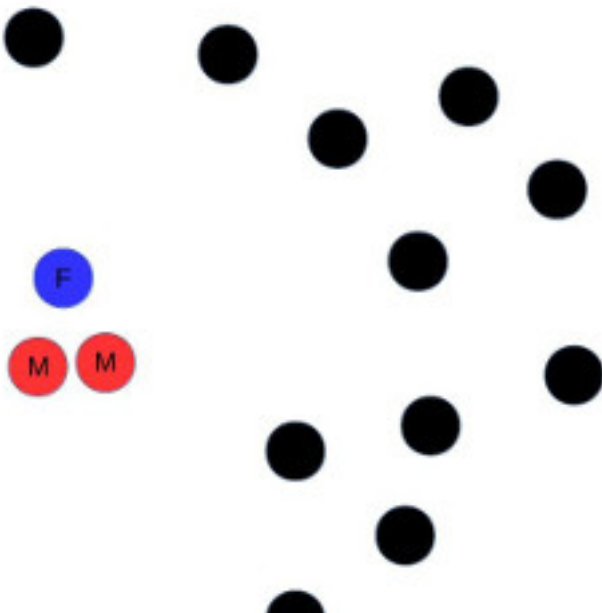
Typical simulation details

- ▶ Time-scale: 1 week
- ▶ All agents examined to see who is forming new partnerships.
- ▶ Agents can acquire new high or low risk partner
- ▶ Marriage affects partner selection

Euclidean plane

- ▶ Mapping agents to Euclidean plane would help
- ▶ Efficient nearest neighbour approximation algorithms
 - ▶ Locality-sensitive hashing
 - ▶ Best bin first
 - ▶ Balanced box decomposition

Euclidean space doesn't work for us



Distance function

- ▶ Example of distance function on webpage
- ▶ Extract from this a **cluster** function

Algorithms

- ▶ Brute force (reference: too slow)
- ▶ Random match (reference: too hopeless)
- ▶ Random match k
- ▶ Weighted shuffling
- ▶ Cluster shuffling

Brute force

```
brute_force_match(Agents):  
    // Agents is an array of agents  
    shuffle(Agents)  
    best = infinity  
    for each agent, a, in Agents  
        for each unmatched agent, b, after a in Agents  
            d = distance(a, b)  
            if d < best  
                best = d  
                best_partner = b  
    make a and best_partner partners
```

This is $O(n^2)$.

Cluster shuffle match

```
cluster_shuffle_match(Agents, c, k)
// Agents is an array of agents
// c is the number of clusters
// k is the number of neighbours to search
// cluster_size = number of agents / c

    calculate cluster values for all agents

    sort agents in cluster_value order

    shuffle each cluster

    for each agent
        find best partner from k neighbours
```

This is $O(n \log n)$.