

Evolutionary Equilibria in Computer Networks: Specialization and Niche Formation

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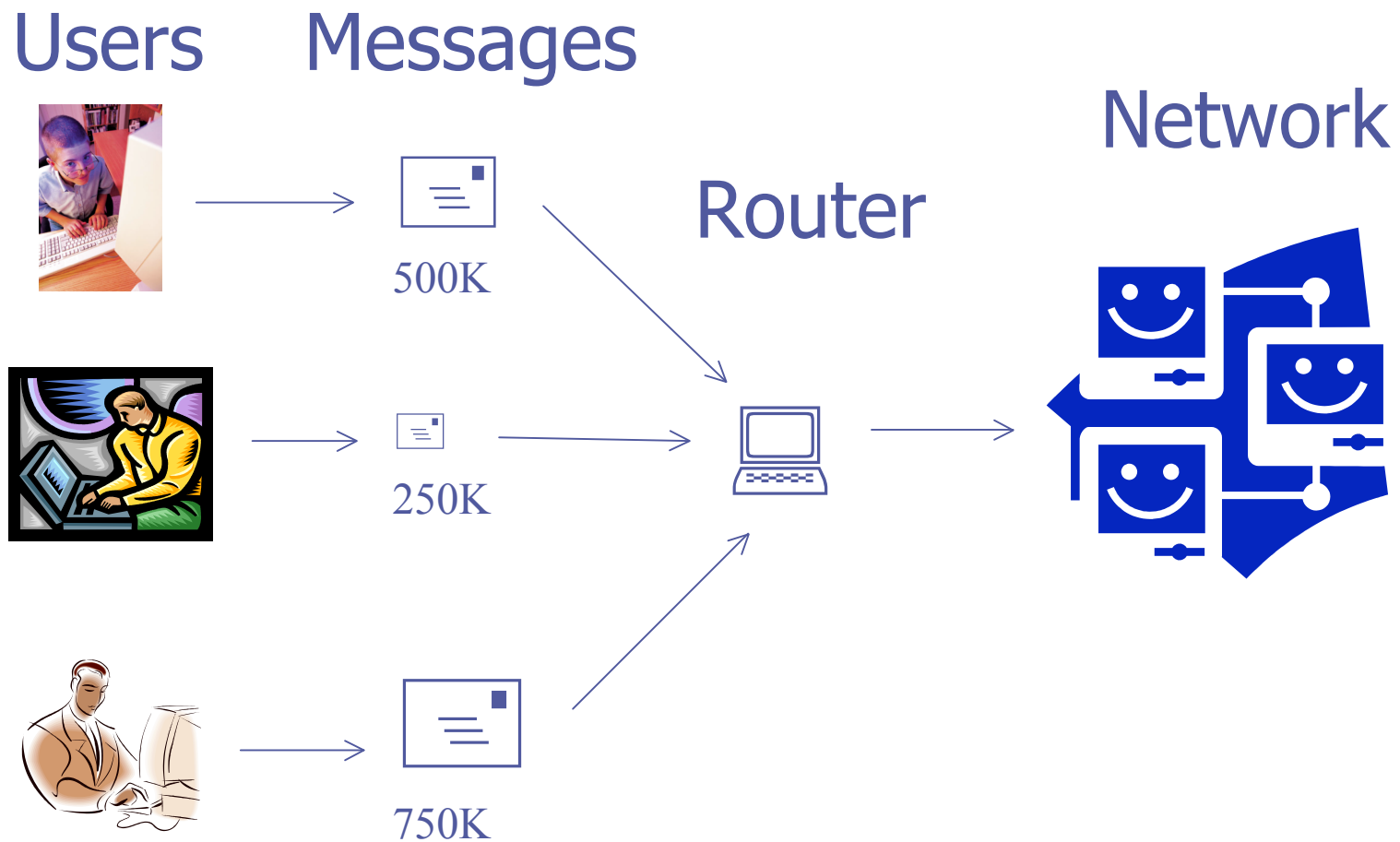
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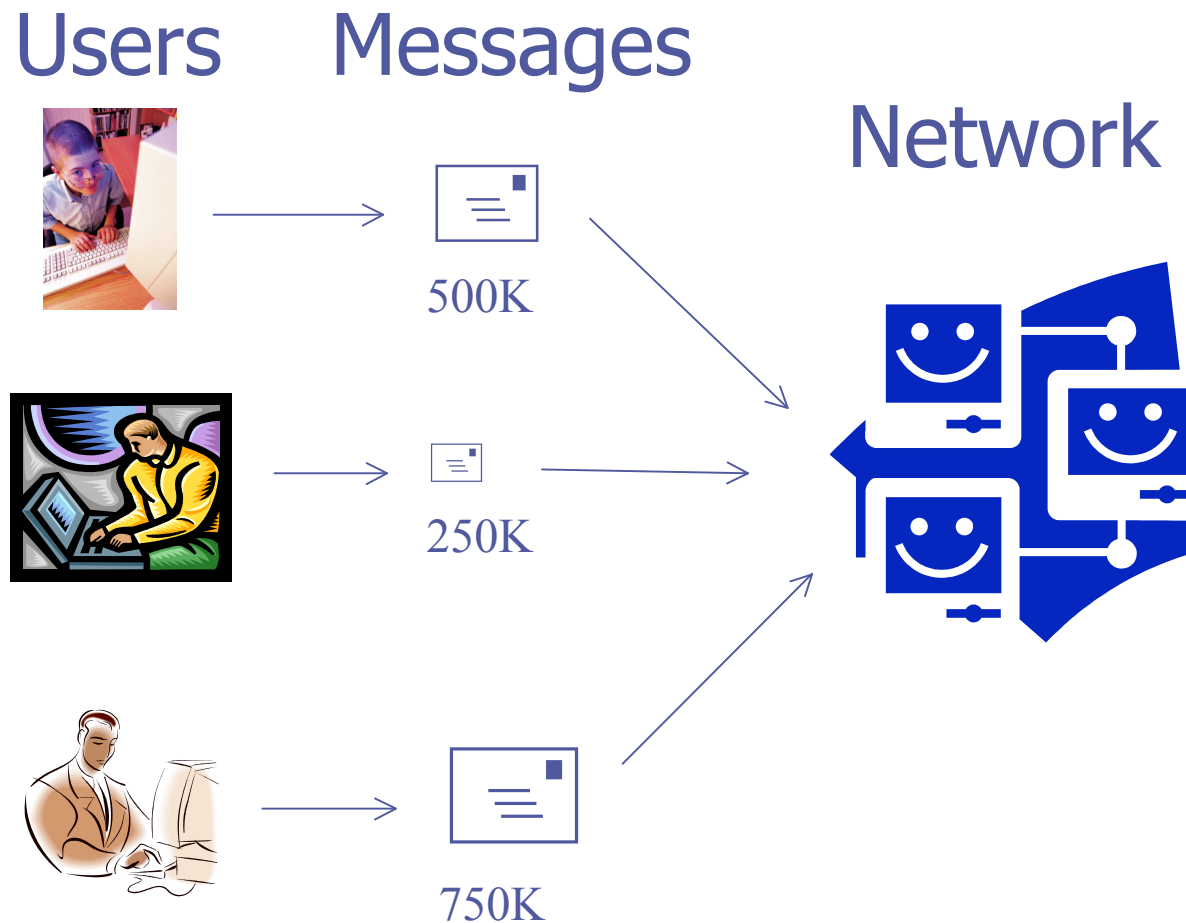
Modelling User Communities

- ◆ A system provides users with access to resources, e.g. a **network**.
- ◆ Centralized planning: gather requests, compute optimal allocation.
- ◆ “Anarchy”: users individually choose resources, e.g. **routes for messages**.
- ◆ Individual choice \rightarrow strategic interactions (\approx traffic models).

Central Allocation



Decentralized Individual Choice



Motivation for Game-Theoretic Modelling

Use game theory to predict outcome of “selfish” user choices (Nash equilibrium)










1. Assess “price of anarchy”
2. Improve network design/protocols

Outline

- Parallel Links Model
- Bayesian Parallel Links Game
- Intro to Evolutionary Stability
- ESS for Parallel Links Game
 - Characterization
 - Structural Conditions

Parallel Links Model

Tasks	Links	Speeds
 100K	  	10
 500K		20
 250K		15
 750K		

delay of task w on link l =
 $w / (\text{speed of } l)$

Parallel Links Model as a Game (Koutsoupias and Papadimitriou 1999)

1. Players $1, \dots, n$ with tasks w_1, \dots, w_n
2. Pure strategy = (choice of) link
3. Fix choices $(w_1, l_1), \dots, (w_n, l_n)$.

\Rightarrow load on link $l = \sum_{i=1..n} w_i$ for $l_i = l$.

\Rightarrow utility u_i for player i =
- load on link l_i
speed of link l_i

Bayesian Routing Game

(Gairing, Monien, Tiemann 2005)

- Agents are uncertain about tasks.
 - common dist. μ over tasks W
 - strategy \sim "program" p for routing tasks
 - $p(l|w)$ = probability that program p chooses link l when given task w .
- $u_i(p_1, \dots, p_n) = \sum_{\text{task assignments}} \langle w_1, \dots, w_n \rangle$
 $\prod_{j=1..n} \mu(w_j) \cdot u_i [(w_1, p_1 | w_1), \dots, (w_n, p_n | w_n)]$

Motivation for Evolutionary Analysis

1. Under “anarchy”, we expect successful strategies to spread → evolutionary dynamics.
2. Highly successful predictions in biology.
3. Distinguishes stable from unstable equilibria.
4. May be useful in network design:
see W. Sandholm’s (2002) pricing scheme for traffic congestion. “evolutionary implementation in computer networks seems an important topic for future research”.

Hawk vs. Dove As A Population Game

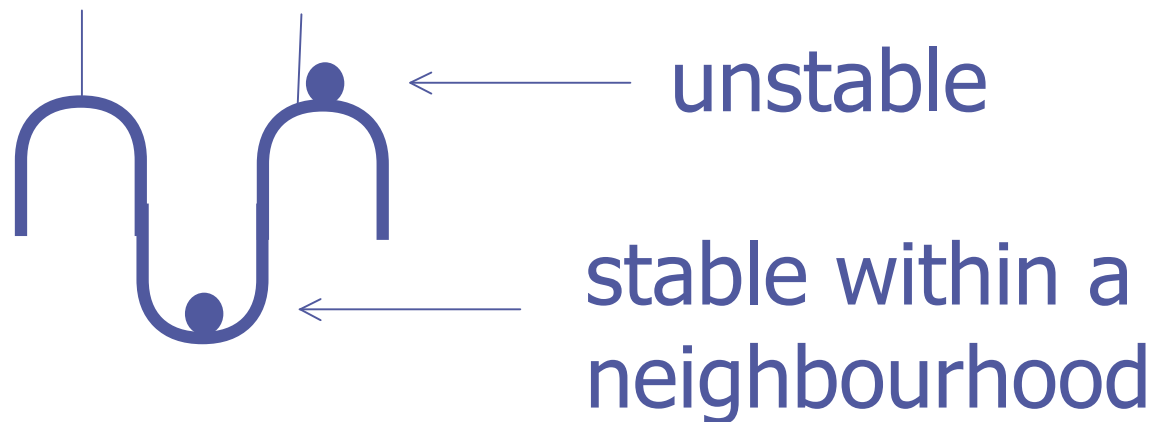
	Hawk (H)	Dove (D)
Hawk	-2,-2	6,0
Dove	0,6	3,3

- Assume a large population of agents.
- Agents are either hawks (H) or doves(D).
- We randomly draw 2 at a time to play.

Population Interpretation of Nash Equilibrium

1. Consider a population of agents with **frequency distribution** π .
e.g. [H,H,H,H,H,H,D,D,D,D]
2. π is in equilibrium
 \Leftrightarrow H does as well as D
 $\Leftrightarrow (\pi, \pi)$ is a **symmetric Nash equilibrium**.
3. (π, π) does **not** represent the choices of 2 players.
4. (π, π) says that both positions are drawn from the same population of agents with distribution π .

Stable vs. Unstable Equilibrium



Evolutionarily Stable Strategies (ESS)

mixed population dist. = $(1-\varepsilon) \pi^* + \varepsilon \pi$

<p>current dist π^* HHHHHH DDDD $10/12 = 1-\varepsilon$</p>	<p>mutant dist π H D $2/12 = \varepsilon$</p>	<p>← mutant plays mutant: $u(1/2, 1/2; \pi)$</p> <p>← incumbent plays mutant: $u(6/10, 4/10; \pi)$</p>
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1. A distribution π^* is an ESS \Leftrightarrow for all sufficiently small mutations π the incumbents in π^* do better in the mixed population than the mutants.
 2. A distribution π^* is an ESS \Leftrightarrow there is an ε^* such that for all sizes $\varepsilon < \varepsilon^*$ $u(\pi^*; (1-\varepsilon) \pi^* + \varepsilon \pi) > u(\pi; (1-\varepsilon) \pi^* + \varepsilon \pi)$ for all mutations $\pi \neq \pi^*$.

Characterization of ESS in Bayesian Routing Game B

Define:

- the load on link l due to strategy p :
 $\text{load}(p, l) = \sum_{\text{tasks } w} \mu(w) \cdot p(l|w) \cdot w$
- the (marginal) probability of using link l :
 $\text{prob}(p, l) = \sum_{\text{tasks } w} \mu(w) \cdot p(l|w)$

Theorem. A strategy p^* is an ESS in $B \Leftrightarrow$
for all best replies $p \neq p^*$ we have
 $\sum_{\text{links } l} [\text{load}(p^*, l) - \text{load}(p, l)] \cdot [\text{prob}(p^*, l) - \text{prob}(p, l)] > 0$

Intuition: to defeat mutation p :

- if load on link increases, use link less (- x -)
- if load decreases, use link more (+ x +)

Necessary Condition: Same Speed, Same Behaviour

Proposition. Let B be a Bayesian routing routing game with ESS p^* . If two links l_1, l_2 have the same speed, then $p^*(l_1|w) = p^*(l_2|w)$ for all tasks w .

Links		Speeds
$w_1:50\%, w_2:50\%, w_3:0$	_____	10
$w_1:50\%, w_2:50\%, w_3:0$	_____	10
$w_1:0, w_2:0, w_3:100\%$	_____	15

Necessary Condition: bigger tasks get faster links

Proposition. Let B be a Bayesian routing game with ESS p^* . Suppose that

1. link 1 is faster than link 2
2. p^* uses link 1 for task w_1 , link 2 for task w_2 .

Then $w_1 \geq w_2$.

	Links	Speeds
$w_2 = 10: 50\%$	_____	10
$w_1 = 20:100\%, w_2: 50\%$	_____	15

Single Task: Unique ESS

Proposition. Let B be a Bayesian network routing game with just one task w .

1. B has a unique ESS p^* .
2. If all m links have the same speed, $p^*(l_j | w) = 1/m$ is the unique ESS.

	Links	Speeds
$w: 1/3$	_____	10
$w: 1/3$	_____	10
$w: 1/3$	_____	10

Strong Necessary Condition: No Double Overlap

- ◆ Fix a Bayesian network game B .
- ◆ Strategy p^* **uses** link l for weight $w \Leftrightarrow p^*(l|w) > 0$.
- ◆ **Proposition.** Let p^* be an ESS in B . Suppose that p^* uses two distinct links $l_1 \neq l_2$ for task w . Then p^* does not use both l_1 and l_2 for any other task w' .

	Links	Speeds
$w_1:70\%, w_2:30\%$	_____	10
$w_1:30\%, w_2:70\%$	_____	20
$w_3:100\%$	_____	15

>2 Tasks, Uniform Speeds: No ESS

Proposition. Let B be a Bayesian network game with >1 link, >1 task, all links the same speed. Then there is no ESS for B .

double overlap ↓	Links	Speeds
$w_1:50\%, w_2:50\%$	_____	10
$w_1:50\%, w_2:50\%$	_____	10

Clusterings are typical ESS's

- ◆ Fix a Bayesian network game B with strategy p^* .
- ◆ A link l is **optimal** for task w given p^* $\Leftrightarrow l$ minimizes $w/\text{speed}(l) + \text{load}(l, p^*)$.
- ◆ A strategy p^* **clusters** \Leftrightarrow if two distinct links $l_1 \neq l_2$ are optimal for task w , then neither l_1 nor l_2 is optimal for any other task $w' \neq w$.
- ◆ **Proposition.** If p^* clusters, then p^* is an ESS.

Tasks

Links



Graph for
Optimality
Relation

Does A Clustered Equilibrium Exist?

◆ Fix an assignment A of links to tasks.

◆ **Proposition.**

1. There is *at most one* clustered ESS p^* whose clustering is A .
2. The candidate p^* can be computed in polynomial time.
3. The question: is there a clustered ESS p^* for a game B ? is in NP.

Future Work

- ◆ Conjecture: if an ESS exists, it's unique.
- ◆ Conjecture: the “no double overlap” condition is sufficient as well as necessary.
- ◆ Computational Complexity and Algorithms for computing ESS's.

Conclusion

- ◆ ESS *refines* Nash equilibrium and defines *stable* equilibria.
- ◆ Analysis of evolutionary stability in Bayesian network games:
 - characterization of *successful mutations*
 - *structure* of stable *task/link allocations*.
- ◆ Finding:
 - evolutionary dynamics leads to formation of “niches” or **clusters** for task/link combinations.
 - Symmetric outcomes tend to be socially suboptimal.