

Optimizing MaxWeight For Routing

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MaxWeight: Issues Raised at July Meeting





MaxWeight can be improved once it is better understood

Context of the reported work



Upper bounding techniques		Assumptions
Novel techniques		Realistic
New application of old techniques		Unrealistic but rob
Unconsummated union – applica	able "layers"	
Application layer		Overhead, fee
Transport layer		Ignored
Routing layer		Considered
Scheduling/queuing layer		Topologies
Physical layer		Canonical problem
Dynamics		Infinite / scaling lav
Arbitrary movement	Possible extension	Finite but arbitr
Random waypoint		Other consid
No mobility		Low latency as w optimal throughp
Capacity-achieving techniques a	nd architectures	Proposed networ application: Rela are required to re complexity
Yes, new techniques		Adaptation
No, old techniques		Mobility a potenti extension



Work shows how important global information can be used *if available.* Generally, amount of global information required for approximate optimization is low

Optimizing MaxWeight for Routing



What is the state of the art and what are its limitations?

MW routing *inflexible with respect* to performance improvement

MW corresponds to *h-myopic*, with *h* quadratic. *Key geometric property of quadratic identified by Meyn prior to July meeting.*

KEY NEW INSIGHTS:

• New perturbation technique:

$$\tilde{x}_i = x_i \log(1 + x_i/\theta)$$

 $h(x) = h_0(\tilde{x})$

NEW INSIGHTS

- Application to routing &
 refinements for decentralization
- Heavy traffic optimality
- Taylor series approximation gives interpretation as adaptive MaxWeight - Diagonal matrix adapts to varying congestion

ACHIEVEMENT DESCRIPTION

MAIN RESULT:

h-myopic policy is universally stabilizing

Application to policy synthesis for approximately optimal performance (delay or backlog) in heavy traffic, with log regret

Numerical study underway

Investigate performance and feasibility: 100 nodes, multiple arrivals. Only wireline models investigated to-date.

Excellent performace as predicted by theory

Decentralized implementation appears feasible.

HOW IT WORKS:

- Step 1: Estimation of network cuts
- Step 2: Estimation of congestion on either side
- Step 3: Choice of h_0 piecewise quadratic

Special case: Single dominant destination gives h_0 quadratic function of workload, cost, and effective cost w.r.t. *workload relaxation*

• Decentralized implementation, use of consensus algorithms

GOAL

END-OF-PHASE

CHALLENGE

COMMUNITY

- Wireless models: Apply
- D. Shah's insights on
- maxproduct convergence
- Full analysis of multiple bottlenecks

• Integration with Network Coding projects: *Can we code around network hotspots?*

• Un-consummated union challenge: Integrate coding and resource allocation

 Generally, solutions to complex decision problems should offer insight

Algorithms for dynamic routing: Visualization and Optimization

Simulations for Single Traffic Stream

- Network approximately 100 nodes. Single destination, multiple sources
- MaxWeight compared to policy based on logarithmic perturbation of

$$\begin{aligned} h_0(x) &= c(x) = \sum x_i \\ \text{Greedy} \end{aligned} \qquad \begin{array}{l} h_0(x) &= \hat{J}(w) + \frac{b}{2} [c(x) - \overline{c}(w)]^2 \\ \text{Approximation of DP solution} \end{aligned}$$

• Simulation for high load: 50% improvement over greedy, 25% over MW



Summaries and challenges



CONCLUSIONS: Alignment of workload vector and l_1 cost leads to vastly simplified analysis and implementation

Logarithmic perturbation gives universally stabilizing policies.

For large θ , Taylor series allows interpretation of policy as *adaptive* MW

Performance improvement over MW as expected in simulations

Simulations verify that tighter approximations to the DP solution results in better performance

SCIENTIFIC FOUNDATIONS

Stochastic Lyapunov theory combined with relaxation techniques based on workload to approximate DP solution

HOW BAD IS THE REAL WORLD?

The real world is very bad. Without attention to bottleneck network cuts, a decentralized routing algorithm will create inefficiency through cycling.

PERFORMANCE? Only stability has been established for logarithmic perturbation, though results from simulation studies give optimism.

CAN WE LEARN? Less learning is needed in routing models than first anticipated. Key is the location of bottleneck links. How can this information be shared? Coordination with Ozdgalar and Shah will likely bridge this gap

CAN WE CODE? With the identification of dynamic bottlenecks, it is then evident where the capacity region can be improved

Largest current research bottleneck concerns learning dynamic bottleneck location and workload

References



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• S. P. Meyn. Control Techniques for Complex Networks. Cambridge University Press 2007.

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• *Distributed Subgradient Methods for Multi-agent Optimization* Angelia Nedic and Asuman Ozdaglar. Preprint 2007.

• Polynomial Complexity Algorithms for Full Utilization of Multi-hop Wireless Networks Atilla Eryilmaz, Asuman Ozdaglar and Eytan Modiano. Preprint 2007. Control Techniques FOR Complex Networks



Sean Meyn





Optimizing MaxWeight: From July Meeting



NEW INSIGHT

What is the state of the art and what are its limitations?

Static routing: *ignores dynamics*

MW routing: *inflexible with respect to performance improvement*

Subramanian & Leigh 2007: *MW can be irrational*

KEY NEW INSIGHTS:

 $MW = h-myopic \text{ for a fluid} \\model, with h quadratic$

Fluid model: $\frac{d}{dt}q(t) = \Delta_u(q(t))$ h-myopic policy: $\underset{urgmin}{\operatorname{argmin}} \langle \nabla h(x), \Delta_u(x) \rangle$

Key geometric property of quadratic is identified:

 $\frac{\partial}{\partial x_i}h(x) = 0 \quad \text{whenever } x_i = 0$ Leads to broad new classes
of policies

ACHIEVEMENT DESCRIPTION

MAIN RESULT:

Perturbation technique to generate functions with appropriate geometry

Application to policy synthesis for approximately optimal performance (delay or backlog) in heavy traffic, with logarithmic regret



HOW IT WORKS:

Key analytical tool is Lyapunov theory for Markov processes: The function *h* satisfies Condition (V3) of Meyn & Tweedie 1992; An exponentiated version satisfies (V4)

For approximate optimality, *workload relaxation* Relaxation also provides tool for visualization of high dimensional dynamics. Optimal solutions evolve in region containing *monotone region* for the effective cost.



• Full analysis of multiple bottlenecks

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