

Whittle Index Approach to Size-aware Scheduling with Time-varying Channels

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ACM Sigmetrics

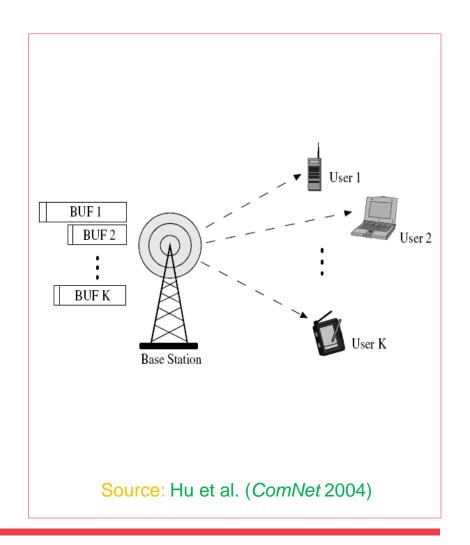
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Outline

- Introduction
- Whittle index approach
- Our contribution
- Illustrations
- Summary

Research problem

- Downlink data transmission in a cellular system
 - traffic = elastic flows
 - file transfers using TCP
 - file sizes known
- Traffic dynamics
 - time scale of seconds+
- Time-varying channels of users
 - time scale of milliseconds
 - channel states known
- Scheduling decisions
 - time scale of milliseconds
- Optimal scheduler for flow-level performance?
 - time scale of seconds+



Two approaches to solve the problem

- Time-scale separation
 - allows to solve the optimization problem exactly
 - applicable for the homogen. case
 - ... but intractable in the general case with heterogeneous users
 - Sadiq and de Veciana (*ITC* 2010)
 - Aalto et al. (Sigmetrics 2011)
 - Aalto et al. (QS 2012)

- Whittle index approach
 - applies restless multi-armed bandits
 - tractable in the general case with heterogeneous users
 - but solves the optimization problem just heuristically
 - Ayesta et al. (PEVA 2010)
 - Jacko (*PEVA* 2011)
 - Cecchi and Jacko (Sigmetrics 2013)
 - Taboada et al. (*ITC* 2014)
 - Taboada et al. (PEVA 2014)

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Multi-armed bandit



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Multi-armed bandit problem

Problem:

- Assume there are K discrete-time bandit processes
- If chosen at time t, the bandit process evolves as a Markov process;
 otherwise its state is frozen until the next time slot t+1
- If process i is chosen when in state x_i , a reward of $r_i(x_i)$ is earned
- Given the states x_i of the bandit processes, choose the optimal bandit i^*

Answer:

- Calculate the Gittins index $G_i(x_i)$ separately for each process i
- Choose the bandit i* with the highest Gittins index
- Gittins and Jones (1974), Gittins (1989)

Note:

 "It was by no means evident that the optimal policy would take the form of such an *index policy*, and certainly not how the index should be calculated" Whittle (*JAP* 1988)

Restless bandit problem (1)

Original problem:

- Assume there are K discrete-time restless bandit processes
- If chosen at time t, the bandit process evolves as a Markov process;
 otherwise its state evolves according to another Markov process
- If process *i* is chosen when in state x_i , a reward of $r_{i,1}(x_i)$ is earned; otherwise another reward of $r_{i,2}(x_i)$ is earned
- Given the states x_i of the bandit processes, choose the optimal bandit i^*

Relaxed problem:

- Given the states x_i of the bandit processes, choose the optimal bandits so that at most one process is chosen per time slot in the long run
- Whittle (*JAP* 1988)

Restless bandit problem (2)

- Answer to the relaxed problem:
 - Consider the separable Lagrangian version of the relaxed problem
 - Show indexability separately for each process i
 - Calculate the Whittle index $W_i(x_i)$ separately for each process i
 - Choose all those bandits with the index greater than a threshold
 - Whittle (*JAP* 1988)
- Heuristic answer to the original problem:
 - Choose the bandit i* with the highest Whittle index
 - Whittle (*JAP* 1988)
- Note:
 - In the multi-armed bandit problem: Whittle index = Gittins index

Opportunistic scheduling problem

Problem:

- Assume there are K jobs with geometric sizes X_i (prob. μ_i)
- Channel states $R_i(t)$ are independent two-state IID variables (good/bad)
- If job *i* with channel state r_i is chosen, it completes with prob. $\mu_i \cdot r_i$
- Holding costs are accrued with rate c_i for any uncompleted job i
- Given the channel states r_i of the jobs, choose the optimal job i^*

Heuristic answer:

- Show indexability separately for each process i
- Calculate the Whittle index $W_i(r_i)$ separately for each process i
- Choose the job i* with the highest Whittle index
- Ayesta et al. (*PEVA* 2010)

Generalizations:

- Jacko (PEVA 2011), Cecchi and Jacko (Sigmetrics 2013)
- Taboada et al. (ITC 2014), Taboada et al. (PEVA 2014)

Whittle index for geometric job sizes

Result:

Primary Whittle index for a job with channel state r is given by

$$W(r) = \begin{cases} \infty, & r = r^{g} \text{ ("good" channel)} \\ \frac{c r^{b}}{P\{R = r^{g}\}(r^{g} - r^{b})}, & r = r^{b} \text{ ("bad" channel)} \end{cases}$$

Secondary Whittle index:

$$\widetilde{W}(r) = \begin{cases} c\mu r^{g}, & r = r^{g} \text{ ("good" channel)} \\ 0, & r = r^{b} \text{ ("bad" channel)} \end{cases}$$

Ayesta et al. (*PEVA*, 2010)

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Size-aware opportunistic scheduling problem

Problem:

- Assume there are K jobs with known sizes x_i
- Channel states $R_i(t)$ are independent two-state IID variables (good/bad)
- If job *i* with channel state r_i is chosen, it completes if $x_i < r_i$
- Holding costs are accrued with rate c_i for any uncompleted job i
- Given the job sizes x_i and the channel states r_i,
 choose the optimal job i*

Our approach:

 Approximate the known size with a discrete-time phase-type distribution (i.e., shifted Pascal distribution)

Phase-type approximation

Definition: Shifted Pascal distribution with J phases and succ. prob. p

$$X = X_1 + ... + X_J$$
 $(X_j \text{ IID})$
 $P\{X_j = n\} = (1-p)^{n-1} p, \quad n = 1,2,...$
 $E[X] = \frac{J}{p}, \quad \text{Var}[X] = \frac{J(1-p)}{p^2}$

• Deterministic job size x approximated by a random variable X with shifted Pascal distribution (J phases, success prob. p = J/x)

$$E[X] = x, \qquad C[X] = \sqrt{\frac{1}{J} - \frac{1}{x}}$$

- For large x and J, the relative variance is small!

Approx. opportunistic scheduling problem

Problem:

- Assume there are K_{jobs} with shifted Pascal sizes X_{i} (J, p_{i})
- Channel states $R_i(t)$ are independent two-state IID variables (good/bad)
- If job *i* with channel state r_i is chosen, the job completes its phase with probability $p_i \cdot r_i$
- Holding costs are accrued with rate c_i for any uncompleted job i
- Given the phases j_i and the channel states r_i of the jobs, choose the optimal job i^*

Heuristic answer:

- Consider the separable Lagrangian version of the relaxed problem
- Show indexability separately for each process i
- Calculate the Whittle index $W_i(j_i,r_i)$ separately for each process i
- Choose the job i* with the highest Whittle index

Relaxed opportunistic scheduling problem

Separable Lagrangian version of the relaxed problem:

$$f_i^{\pi_i} + \nu g_i^{\pi_i} = \min_{\pi_i}! \tag{*}$$

where

$$f_i^{\pi_i} \triangleq E \left[\sum_{t=0}^{\infty} c_i \mathbf{1}_{\{Z_i^{\pi_i}(t) > 0\}} \right], \qquad g_i^{\pi_i} \triangleq E \left[\sum_{t=0}^{\infty} A_i^{\pi_i}(t) \right]$$

Definition:

Optimization problem (*) is indexable if for any j and r there is $W_i(j,r)$ such that

- it is optimal to schedule job i in state (j,r) if $v \leq W_i(j,r)$
- it is optimal not to schedule job i in state (j,r) if $v \ge W_i(j,r)$

Whittle index for shifted Pascal job sizes

Result:

Primary Whittle index for a job with *j* remaining phases and channel state *r* is given by

$$W(j,r) = \begin{cases} \infty, & r = r^{g} \text{ ("good" channel)} \\ \frac{c r^{b}}{P\{R = r^{g}\}(r^{g} - r^{b})}, & r = r^{b} \text{ ("bad" channel)} \end{cases}$$

Secondary Whittle index:

$$\widetilde{W}(j,r) = \begin{cases} \frac{c \ p \ r^g}{j}, & r = r^g \ (\text{"good" channel}) \\ 0, & r = r^b \ (\text{"bad" channel}) \end{cases}$$

Approximative size-aware Whittle index

Result:

Primary approximative Whittle index for a job with remaining size *y* and channel state *r* is given by

$$W(y,r) = \begin{cases} \infty, & r = r^{g} \text{ ("good" channel)} \\ \frac{c r^{b}}{P\{R = r^{g}\}(r^{g} - r^{b})}, & r = r^{b} \text{ ("bad" channel)} \end{cases}$$

Secondary approximative Whittle index:

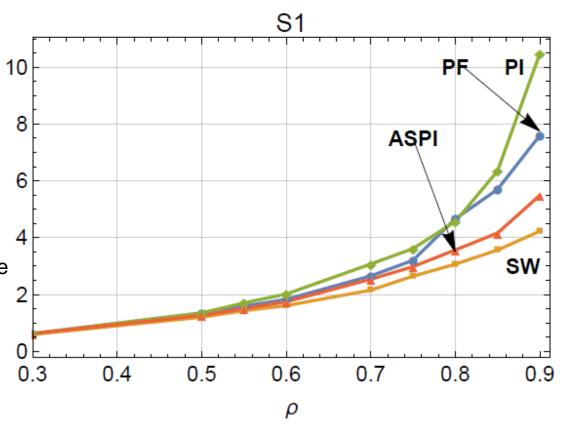
$$\widetilde{W}(y,r) = \begin{cases} \frac{c \, r^g}{y}, & r = r^g \text{ ("good" channel)} \\ 0, & r = r^b \text{ ("bad" channel)} \end{cases}$$

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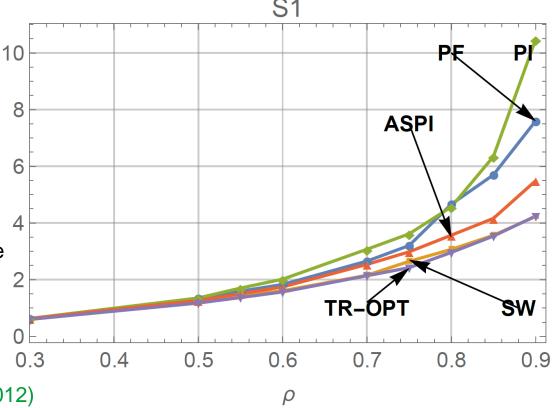
Scenario 1: Homogeneous users

- 1 class
- Poisson job arrivals
- Pareto job sizes
- 2 channel states
- **PF** = Proportional Fair scheduler
- PI = Potential Improv.
 Ayesta et al. (2010)
- ASPI = Attained Service dependent PI Taboada et al. (2014)
- SW = Size-aware
 Whittle index policy



Scenario 1: Homogeneous users

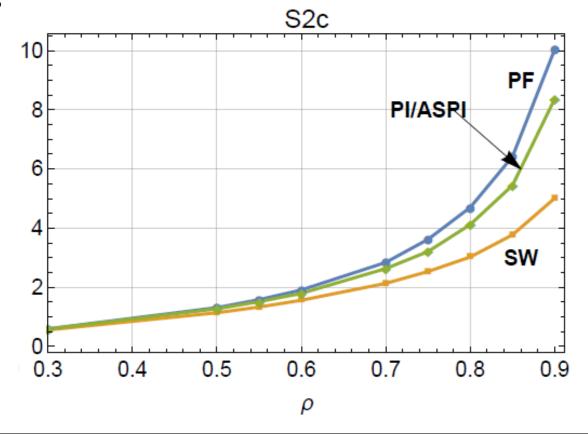
- 1 class
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- **PF** = Proportional Fair scheduler
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- **SW** = Size-aware Whittle index policy
- TR-OPT: Aalto et al. (2012)





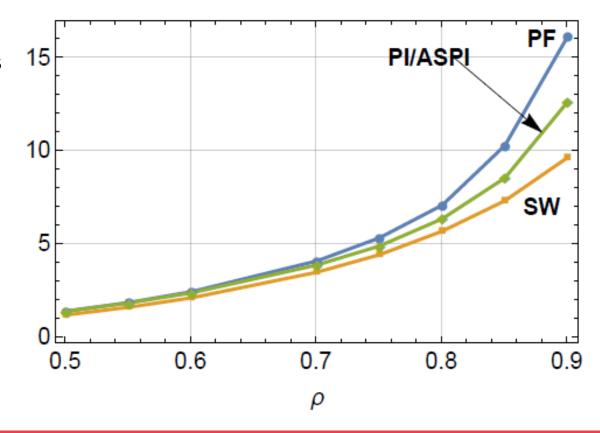
Scenario 2c: Heterogeneous users

- 2 classes with different channels
- Poisson job arrivals
- Exp. job sizes
- 2 channel states



Scenario 3: Multiple channel states

- 2 classes with different channels
- Poisson job arrivals
- Exp. job sizes
- 5/3 channel states



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Summary

- We considered the size-aware opportunistic scheduling problem for elastic downlink data traffic with 2-state time-varying channels
- By the Whittle index approach and a discrete-time phase-type approximation, we were able to derive an approximative size-aware Whittle index
- Primary index:
 - infinite for the good channel state
 - independent of the job size for the bad channel state
- Secondary index:
 - inversely proportional to the remaining size for the good channel state
 - zero for the bad channel state

The End

