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# Optimizing Performance in Networked Systems

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# Outline

- Problem of improving performance in a network with uncertainty
- Difficulty with repeated simulation
- Using approximation methods to avoid repeated simulation
- Example: Analytical Airfield Model (AAM)
  - Small numerical example that nonetheless gives significant managerial insight
- Conclusion

# The Problem

- Given a network with uncertainty, decide how to invest resources to improve its performance
  - These often arise in logistics, but occur also in many other areas
- Performance measure: often throughput
- Investment: change system parameters to
  - Decrease times for critical operations such as transshipment, refueling, maintenance
  - Increase capacity of intermediate stations that perform these operations

# Measuring Performance

- We usually have to measure performance with simulation models
  - Significant uncertainties in these systems
  - Deterministic models are easy to use but they ignore the uncertainties and may give seriously wrong answers
- Simulation models give “snapshots” of performance of a given system, but are slow
  - More so, if system is large and complex
  - And we need to simulate *every time we change parameter values*, to check improvement

# Problem: Slowness

- These methods simulate the system at each step
  - On complex systems they can be very slow
- How can we get around this?
  - One way: speed up the simulations
    - » Network of workstations (Condor Project)
  - Another: maybe avoid simulating each time
    - » Use approximation methods instead

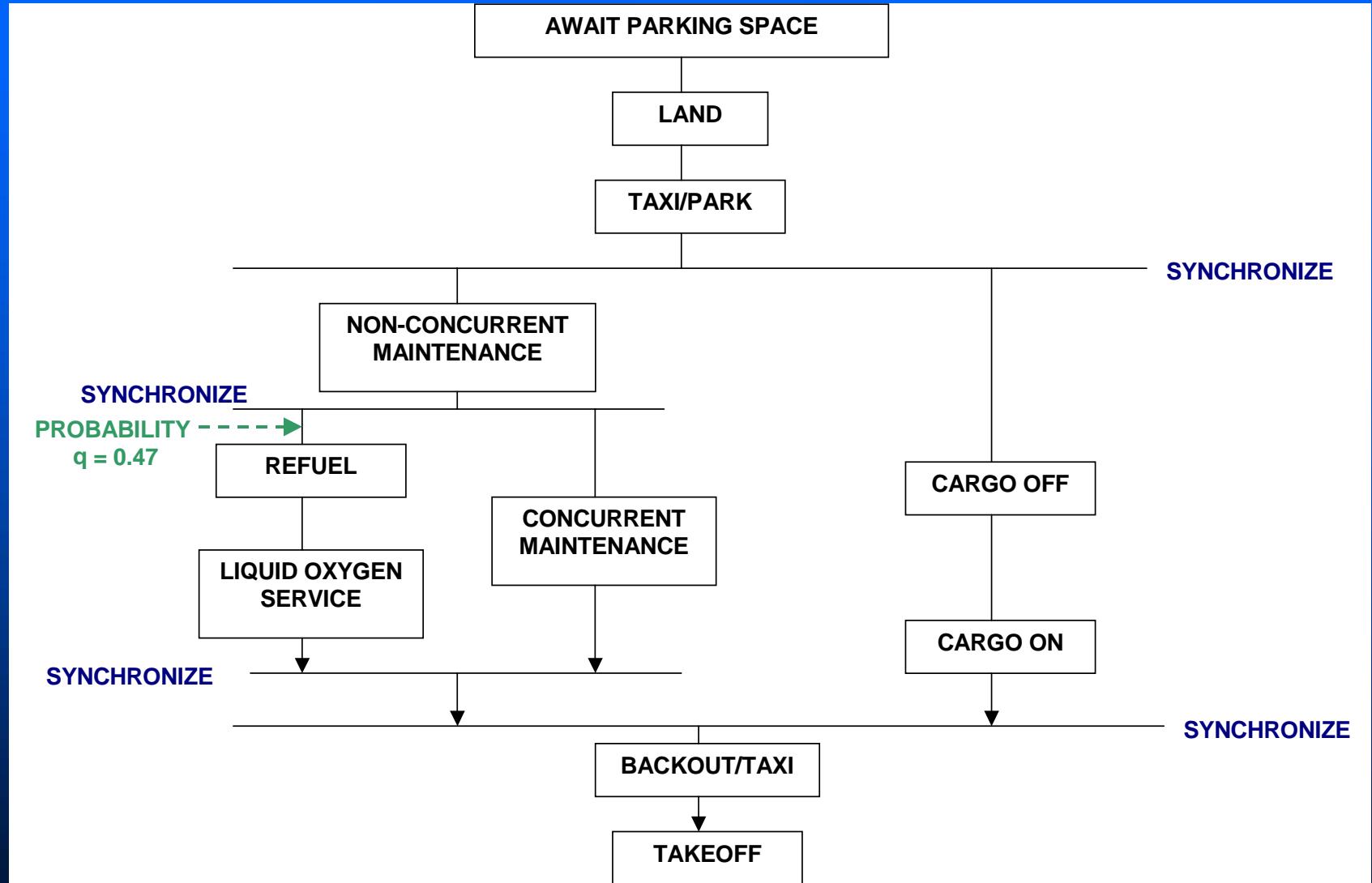
# Network Approximations

- Idea: don't simulate the system exactly
  - Approximation requires *only solving a system of nonlinear equations*
    - » Much faster than simulating
  - On various problems we've tried, ratio of simulation time to approximation time has been in range 16:1 to 80:1 or more
- So: use approximations to improve the system, then simulate once to confirm the results

# Example

- Analytical airfield model (AAM)
  - Ref: Dietz, D. C., *J. Aircraft* (1999) *et seq.*
- Relatively complex network of immediate military interest
  - We'll illustrate how to use optimization to answer managerial questions
  - What's new here is better control methods through use of new 2-moment approximations
  - Tools apply just as well to many other networks

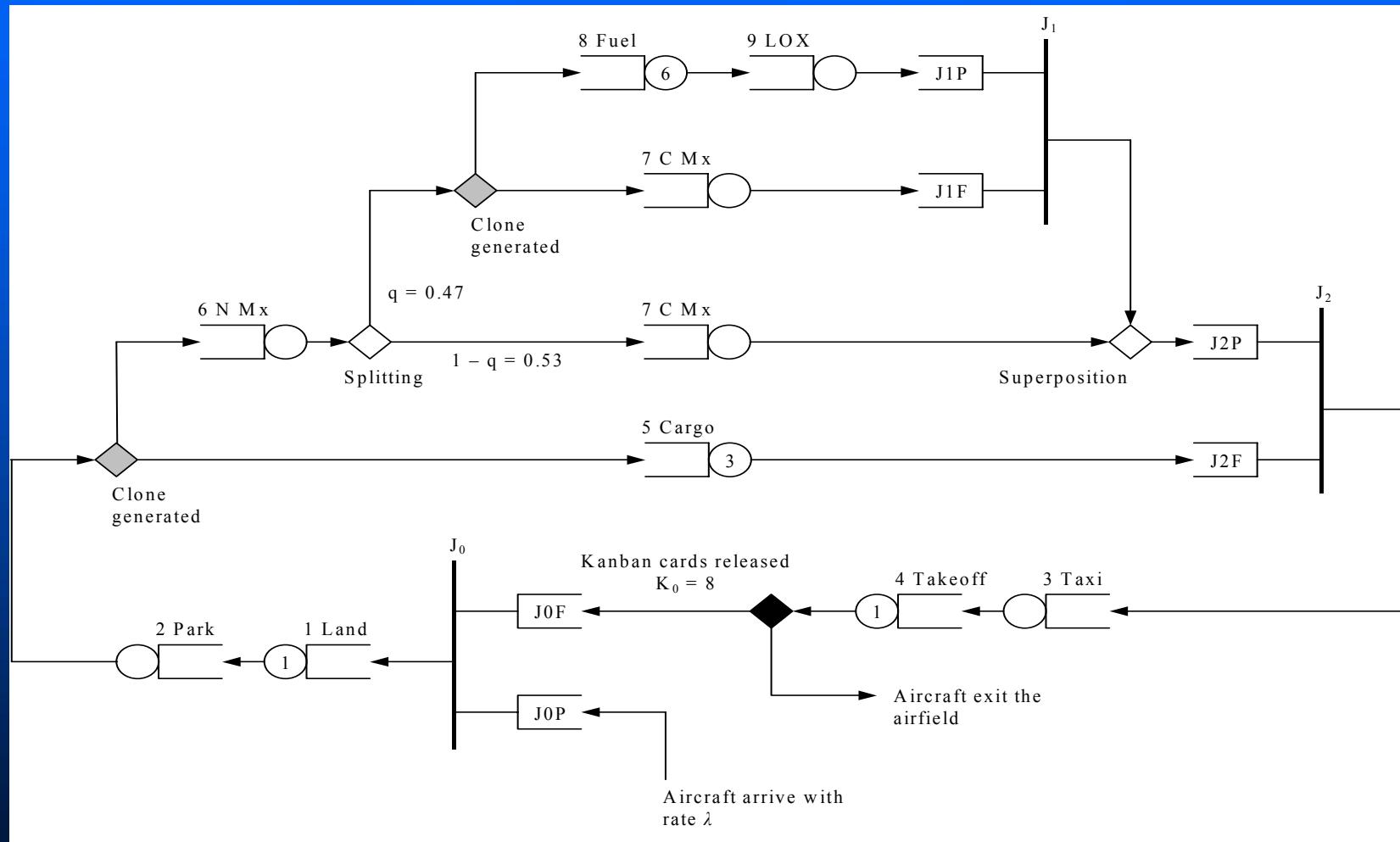
# Task Precedence Graph for the AAM Problem



# Modeling Essential Features

- Concurrent use of resources can be modeled as a fork/join mechanism:
  - Aircraft generate temporary clones with identical attributes
  - Overall waiting time is the maximum over all clone paths
- Aircraft waiting authorization for landing obey a join mechanism governed by Kanban cards
  - Number of Kanban cards equals maximum number of aircraft allowed on ground simultaneously

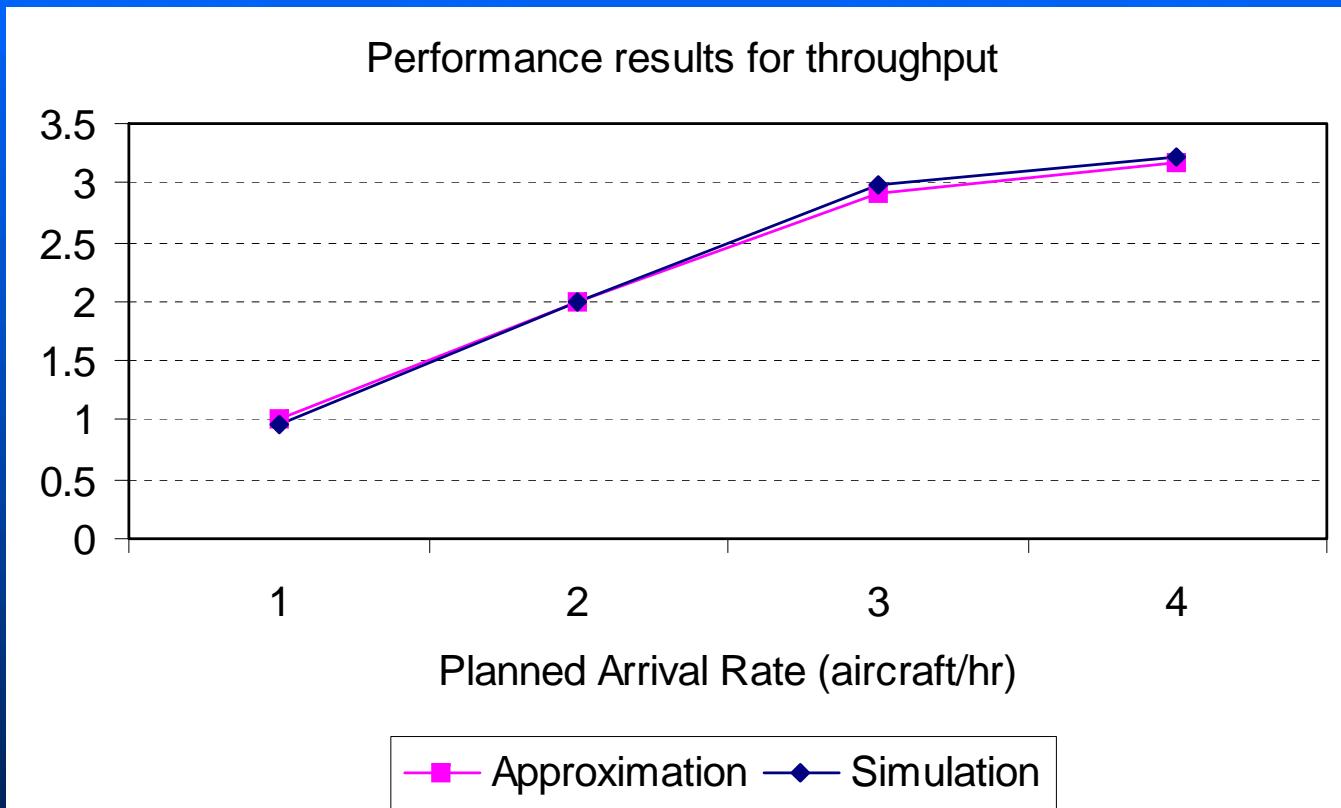
# Layout of Queuing Model



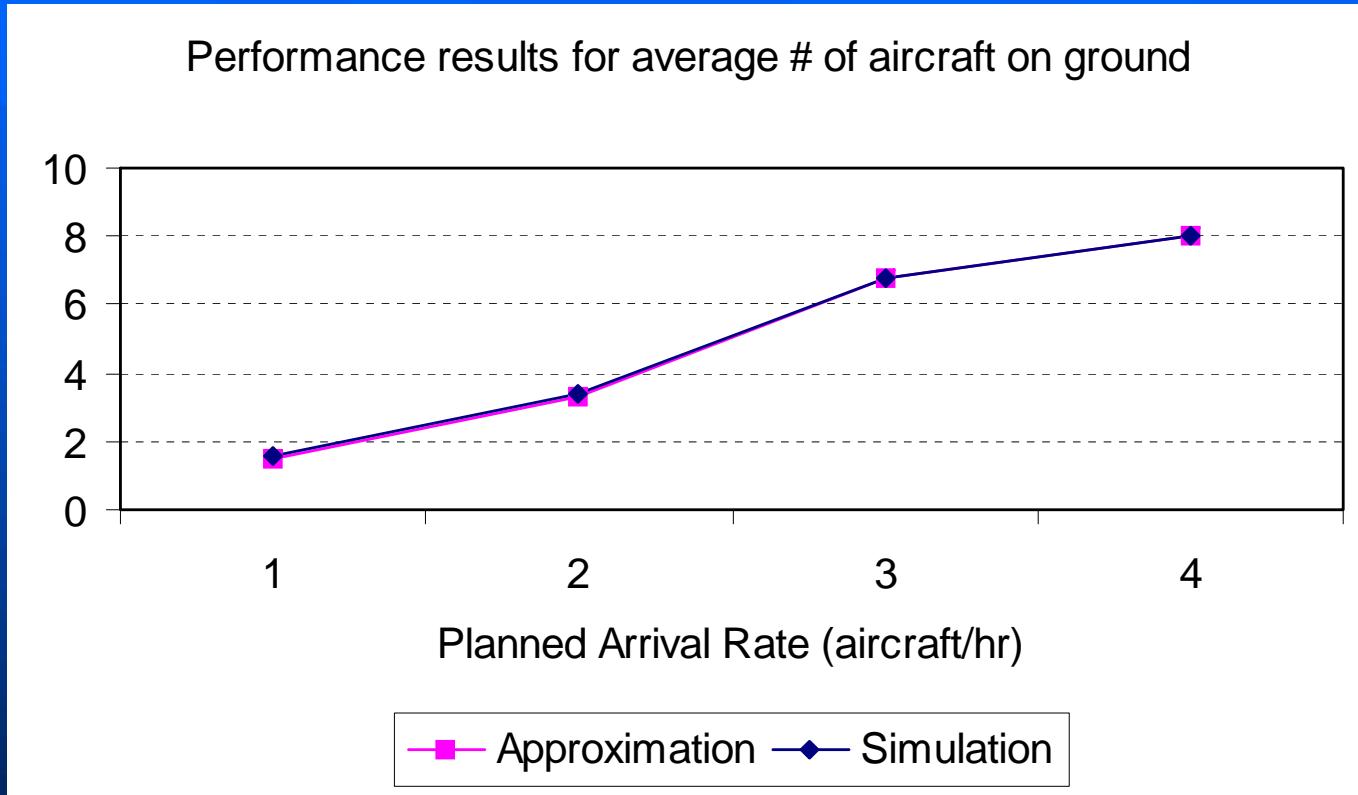
# Approximating the AAM

- Capacity planners need fast answers to changes in the system
  - Performance measures are usually estimated with simulation, which is accurate but costly
- Instead, we do most of the analysis calculations with 2-moment approximations
  - Very fast: solution in < 5 seconds on a PC
  - Relatively accurate (relative error < 10 %)
- Kanban modeling better captures real constraint for landing authorization than current modeling

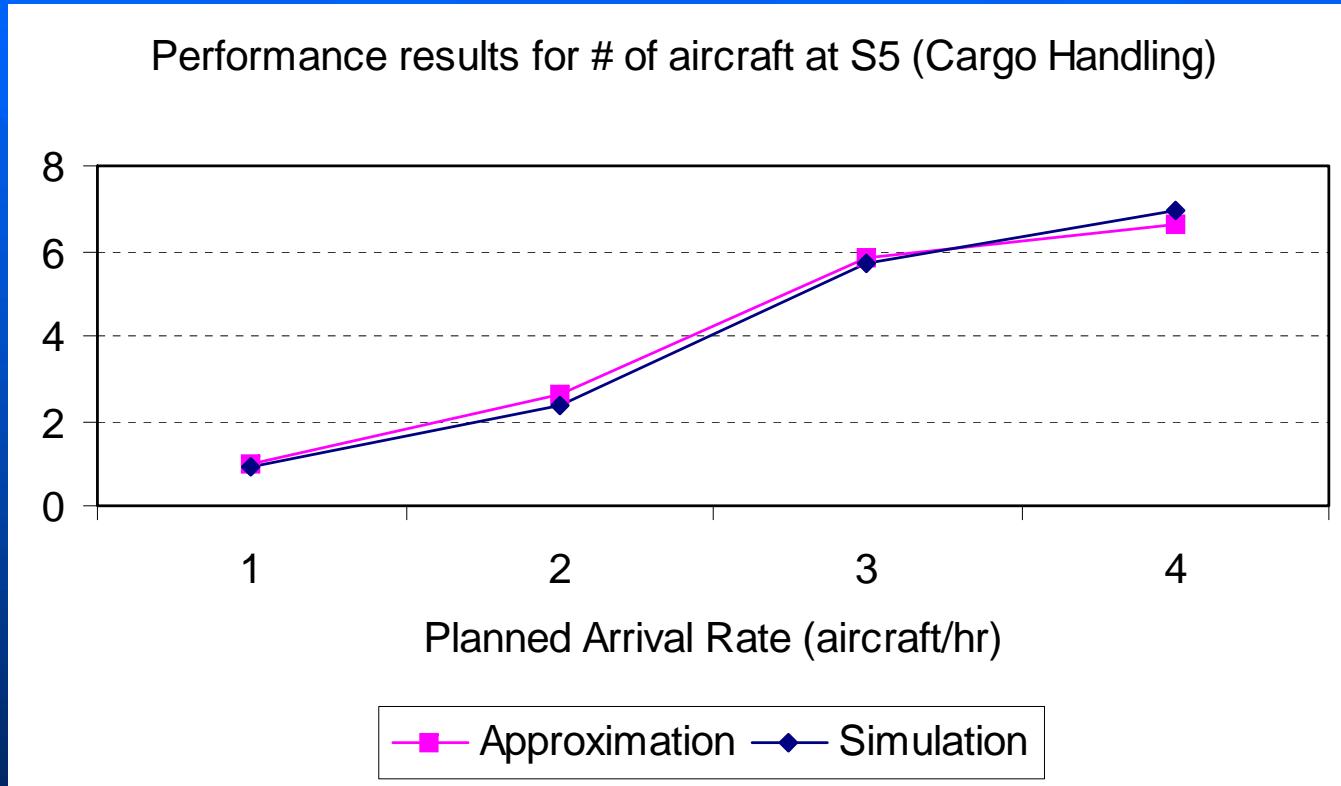
# Example: Throughput



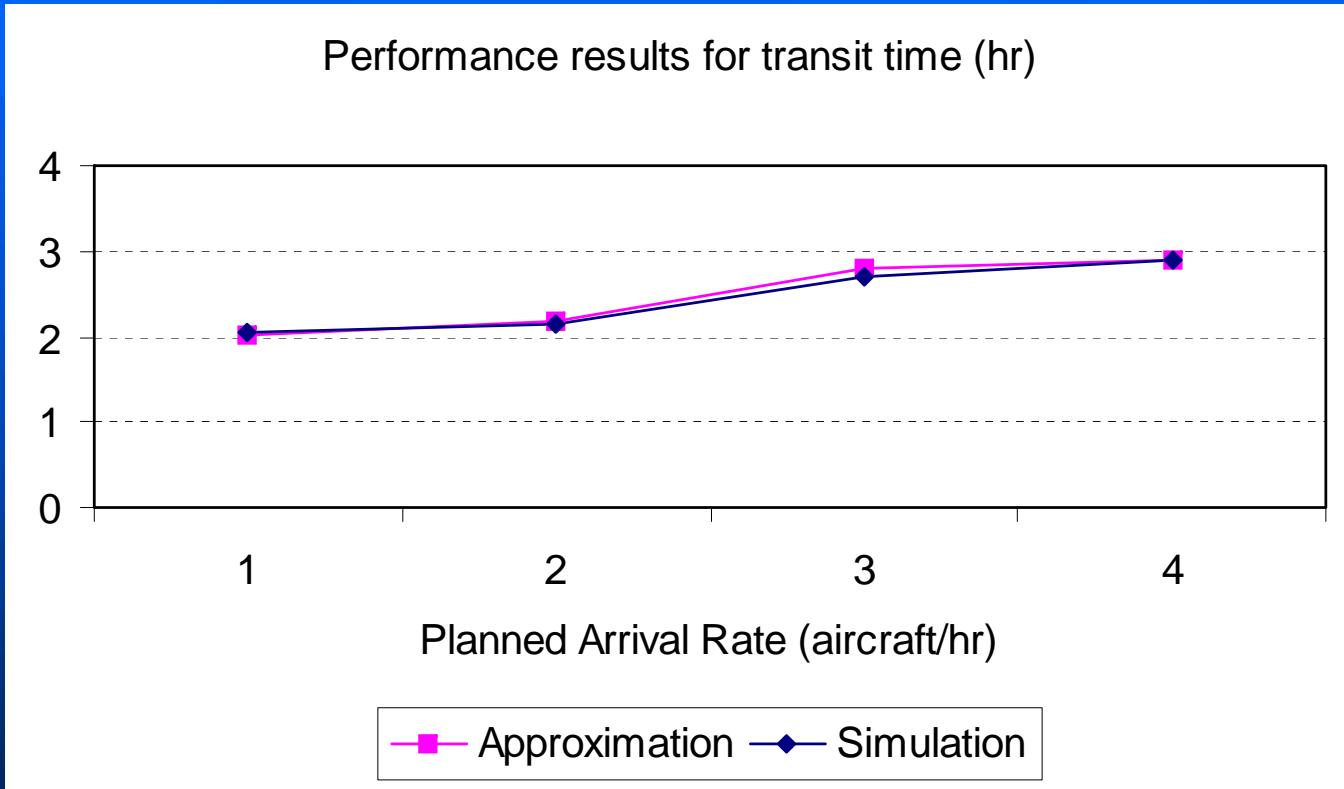
# Example: Average on Ground



# Cargo Handling is a Bottleneck



# Transit Time Up By > 40 %



# Network Improvement

- Initial situation:
  - $\lambda = 3.0$  aircraft/hr
  - $K_0 = 6$  (# allowed on ground)
  - Transit time (Approximation) = 2.482 hr.
- Constraint: transit time should be decreased by 20 %.
- Question: What is the minimum # of capacity units to invest at S5 (Cargo Handling) to have transit time decreased by 20 %?

# Network Improvement

- Answer: 2, so that number of servers at Cargo Handling increases from 3 to 5.

Transit time (hr.)

| Number of servers at<br>Cargo Handling | Approximation | Simulation |
|--|---------------|------------|
| 3                                      | 2.48          | 1.84       |
| 5                                      | 1.89          | 1.49       |



# Network Improvement

- Additional insight: Number of aircraft at Cargo Handling decreases by 31 %.

Number of aircraft at Cargo Handling

| Number of servers at<br>Cargo Handling | Approximation | Simulation |
|--|---------------|------------|
| 3                                      | 4.67          | 4.25       |
| 5                                      | 3.22          | 2.92       |

-31%



# Conclusion

- Fast numerical procedure that gives *quick*, reasonably accurate results
  - Can validate results by using just one simulation run instead of many
- New technical results allow modeling of more complex systems
  - Better modeling of join stations
- Results
  - We can ask and answer managerial questions quickly and easily
  - Improve operations by investing optimally

(Backup slides follow)

# Main Assumptions for 2-Moment Approximations

- Traffic processes are approximated as renewal processes
- Traffic process rate and SCV (squared coefficient of variation) are enough to completely characterize the performance measures of the network
- Example: for a GI/G/1 queue

$$W_q = \left( \frac{c_a^2 + c_s^2}{2} \right) \left( \frac{\lambda \tau^2}{1 - \lambda \tau} \right)$$

$W_q$  : delay in queue

$c_a^2$  : SCV of arrival process

$c_s^2$  : SCV of service process

$\lambda$  : arrival rate

$\tau$  : mean service time

# New Result for a Join Station

- Developed in (Krishnamurthy, 2002)
- Rate and SCV of departure process of a Join station as a function of the 6-tuple:

$$(\lambda_1, c_1^2, K_1, \lambda_2, c_2^2, K_2)$$

- Example:

If  $\rho = \frac{\lambda_1}{\lambda_2} \neq 1$ :

$$\lambda_D = \lambda_1 \left[ \frac{1 - \rho^{K_1 + K_2}}{1 - \rho^{K_1 + K_2 + 1}} \right] \left[ 1 - 0.5 \left( \frac{c_1^2 + c_2^2}{2} \right) \left( \frac{1 - \rho}{1 - \rho^{2(K_1 + K_2) + 1}} \right) \rho^{2(K_1 + K_2)} \right]$$

# References

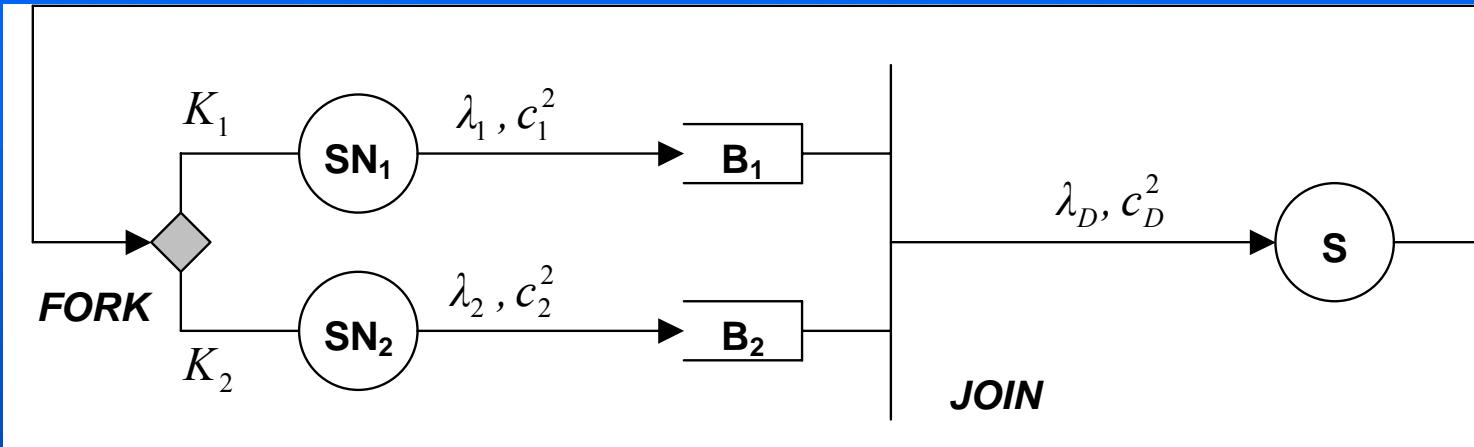
- New results for a Join station:
  - A. Krishnamurthy, Analytical performance models for material control strategies for manufacturing systems, Ph.D. Dissertation, University of Wisconsin-Madison, 2002
- Known results for other types of queues and overall performance evaluation of the network:
  - R. Suri, J.L. Sanders and M. Kamath, Performance evaluation of production networks, in: S.C. Graves et al. (Eds.), *Logistics of Production and Inventory* (Elsevier, Amsterdam, 1993)

# Reference for the AAM Problem

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- Dietz, D. C., Mean value analysis of military airfield operations at an individual airfield, *Journal of Aircraft* 36, No. 5, 1999.
- Start with:
  - Total fleet size equals 20.
  - Maximum number of aircraft allowed on ground  $K_0 = 8$ .
- Vary planned arrival rate from 1.0 to 4.0 aircraft/hr.
- Compute performance measures of interest:
  - Throughput.
  - Transit time on airfield.
  - Average number of aircraft on ground.
  - Queue length at bottlenecks.

# Fork-join Dynamics



**SN<sub>1</sub>** : Sub - network where entity 1 travels independently

**SN<sub>2</sub>** : Sub - network where entity 2 travels independently

**B<sub>1</sub>** : Buffer where entity 1 waits for an arrival of entity 2

**B<sub>2</sub>** : Buffer where entity 2 waits for an arrival of entity 1

**S** : Sub - network where joined entity travels

$\lambda_1, c_1^2, K_1$  : Rate, SCV and shut - down level of arrival process of entity 1

$\lambda_2, c_2^2, K_2$  : Rate, SCV and shut - down level of arrival process of entity 2

$\lambda_D, c_D^2$  : Rate, SCV of departure process of joined entity