

The Asymptotic Variance Rate of the Output Process of Finite Capacity Birth-Death Queues

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Setting

Single station queue with finite capacity, represented by a birth-death Markov jump process:

- Birth rates: $\lambda_0, \dots, \lambda_{K-1}$
- Death rates: μ_1, \dots, μ_K
- Steady state distribution: $\pi = (\pi_0, \dots, \pi_K)$
- Assume queue is stationary
- Output process, $D(t)$ = number of jobs that have received service up to time t
- For $K > 1$, $D(t)$ is not a renewal process
- $\mathbb{E}[D(t)] = \lambda^* t$, $\lambda^* = \sum_{i=1}^K \pi_i \mu_i$

Variance grows asymptotically linearly:

$$\text{Var}(D(t)) = \bar{V} t + o(t).$$

\bar{V} is the asymptotic variance rate.

Main Theorem

A formula for the asymptotic variance rate:

$$\bar{V} = \lambda^* + \sum_{i=0}^{K-1} v_i$$

$$v_i = 2\left(M_i + \frac{M_i^2}{\pi_i \lambda_i}\right), \quad M_i = \sum_{j=1}^i \pi_j \mu_j - \lambda^* \sum_{j=0}^i \pi_j.$$

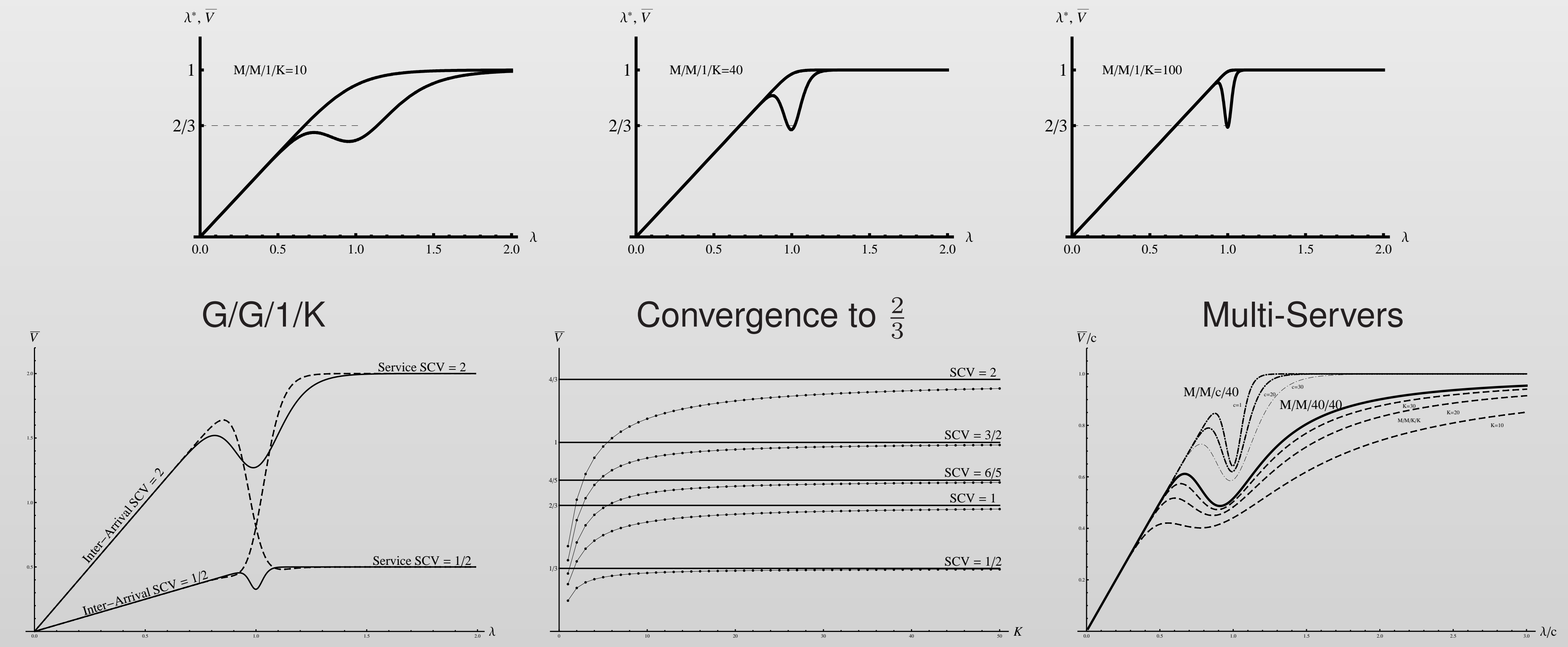
As a consequence:

- When $\lambda_0 \geq \dots \geq \lambda_{K-1}$ and $\mu_1 \leq \dots \leq \mu_K$, then $v_i < 0$, thus $\bar{V}/\lambda^* < 1$
- For $M/M/1/K$ with $\rho = \lambda/\mu$:

$$\bar{V} = \begin{cases} \lambda \frac{2K^2 + K}{3K^2 + 6K + 3} & \rho = 1 \\ \lambda \frac{(1 + \rho^{K+1})(1 - (1 + 2K)\rho^K(1 - \rho) - \rho^{2K+1})}{(1 - \rho^{K+1})^3} & \rho \neq 1 \end{cases}$$

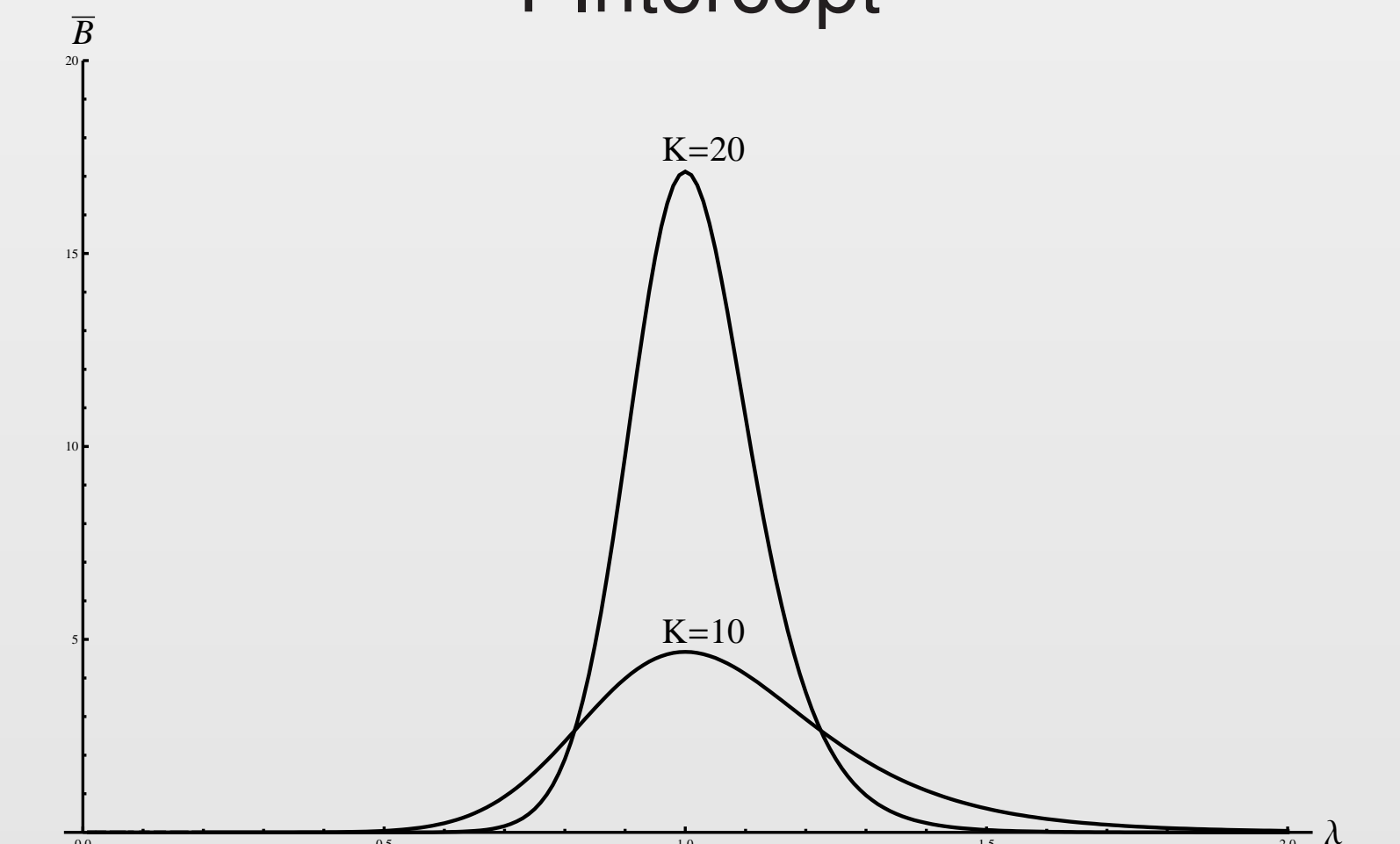
BRAVO Effect – Balancing Reduces Asymptotic Variance of Outputs

\bar{V} is locally minimized when the arrival rate equals the maximal service rate. In single server systems (M/M/1/K and GI/G/1/K) the reduction is by a factor that approaches 2/3 for large K .

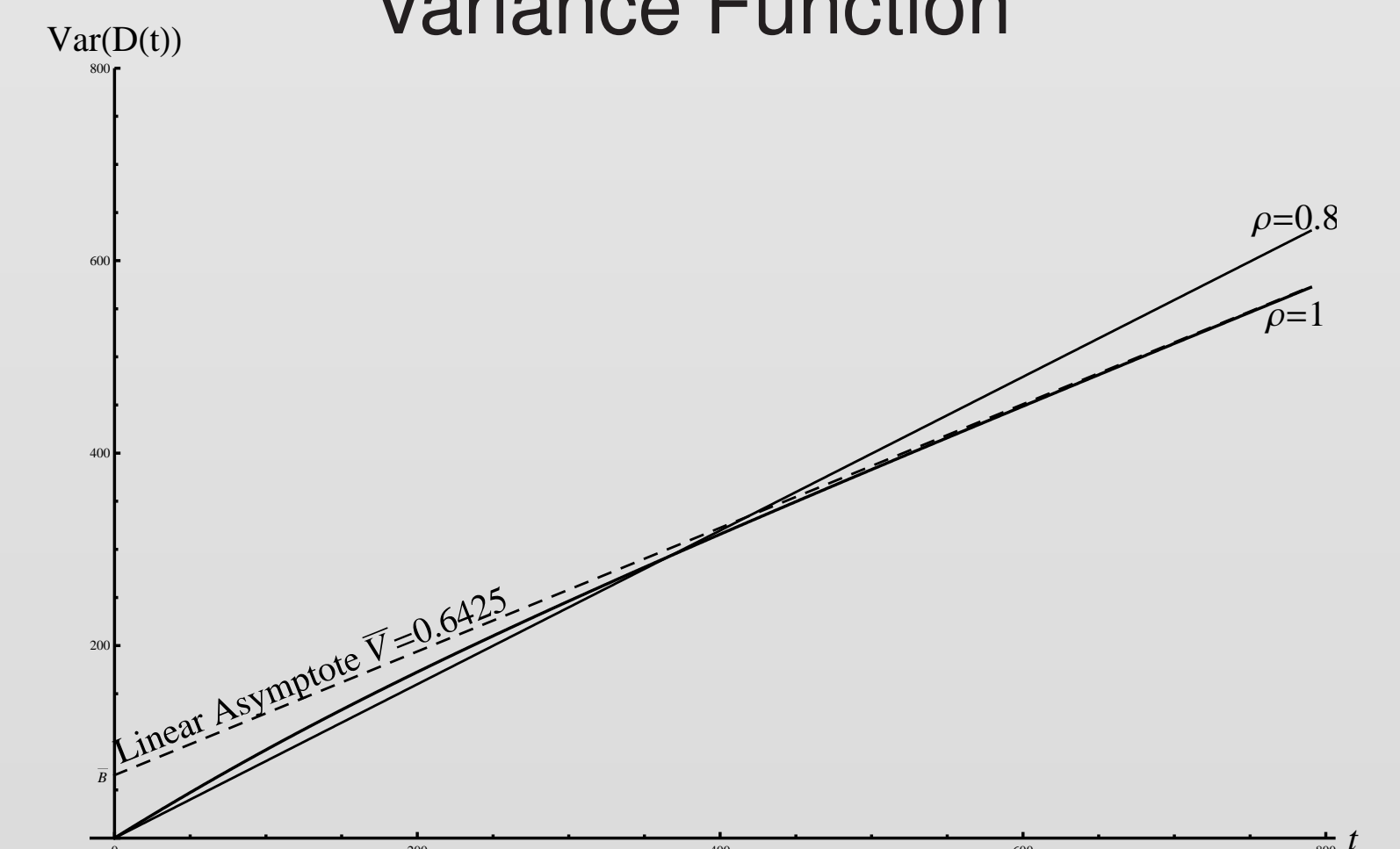


Variance for Finite t

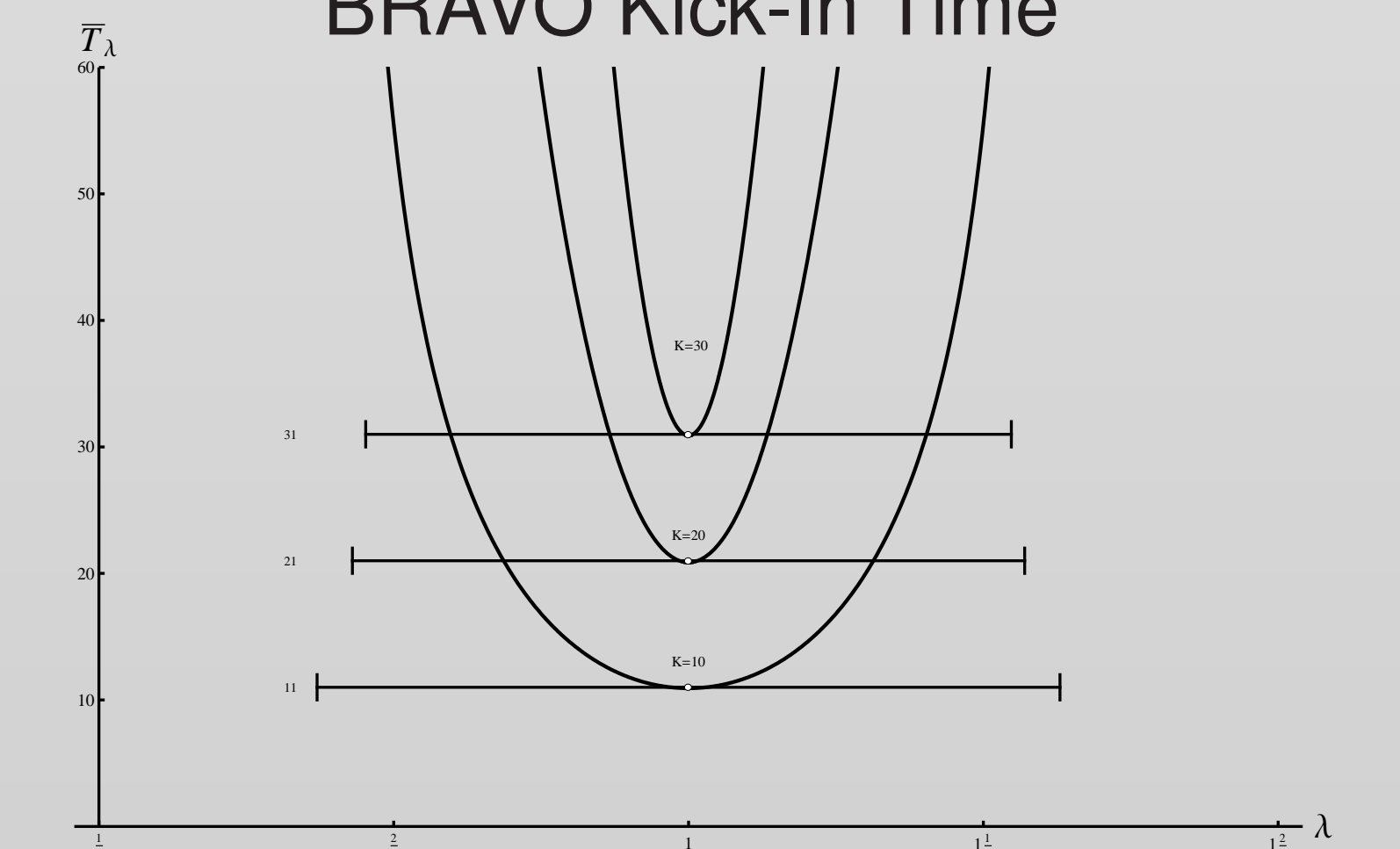
Y-Intercept



Variance Function



BRAVO Kick-In Time



Numerical Computation of \bar{V} using MAPs - Markov Arrival Processes

$$\text{Generator matrix: } \Lambda = \underbrace{\begin{pmatrix} -\lambda_0 & \lambda_0 & & 0 \\ \mu_1 & -(\mu_1 + \lambda_1) & \lambda_1 & \\ & \mu_2 & \ddots & \ddots \\ 0 & & \mu_K & \lambda_{K-1} \\ & & & -\mu_K \end{pmatrix}}_C + \underbrace{\begin{pmatrix} 0 & 0 & 0 \\ \mu_1 & & \\ & \mu_2 & \\ & & \ddots \\ 0 & & & \mu_K & 0 \\ & & & & 0 \end{pmatrix}}_E$$

$$\mathbb{E}[D(t)] = \frac{\pi E \mathbf{1}}{\lambda^*} t \quad \text{Var}(D(t)) = \underbrace{\lambda^* - 2(\lambda^*)^2}_{\bar{V}} - \underbrace{2\pi E \Lambda^{-1} E \mathbf{1}}_B t + 2(\lambda^*)^2 - 2\pi E \Lambda^{-1} \Lambda^{-1} E \mathbf{1} + O(t^{3r+2} e^{-bt})$$

No explicit formula for $\Lambda^{-1} = (\Lambda - \mathbf{1}\pi)^{-1}$.

Main Proof Steps

- $M(t)$ = admissions + outputs. $\bar{V}_M = 4\bar{V}$.
- $M(t)$ (also a MAP) has the same variance as an associated Markov Modulated Poisson Process (MMPP).
- Whitt – a closed formula for calculation of the asymptotic variance of birth-death MMPPs.

Key References

- [1] S. Naryana and M. F. Neuts. The first two moment matrices of the counts for the Markovian arrival process. Stochastic Models, 1992.
- [2] W. Whitt. Asymptotic formulas for Markov processes with applications to simulation. Operations Research, 1992.

Correlation of Outputs and Overflows

