

# On the Representation of Correlated Exponential Distributions by Phase Type Distributions

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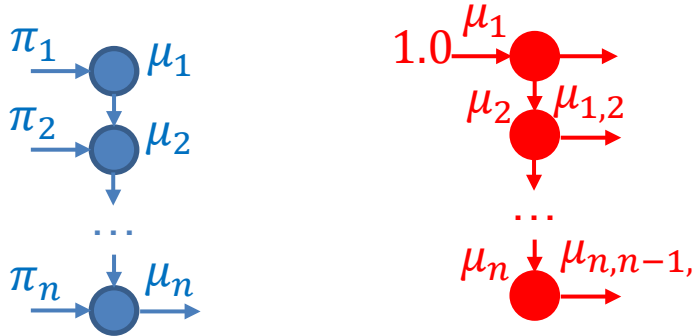


# Outline

- Representation of exponential distributions by acyclic phase type distributions (APHDs)
- Correlation between APHDs
- Optimal representations
- Application

## Acyclic Phase Type Distributions (APHDs)

Canonical representations



For exponential distribution with  $E(X) = \mu^{-1}$  we have Laplace transform

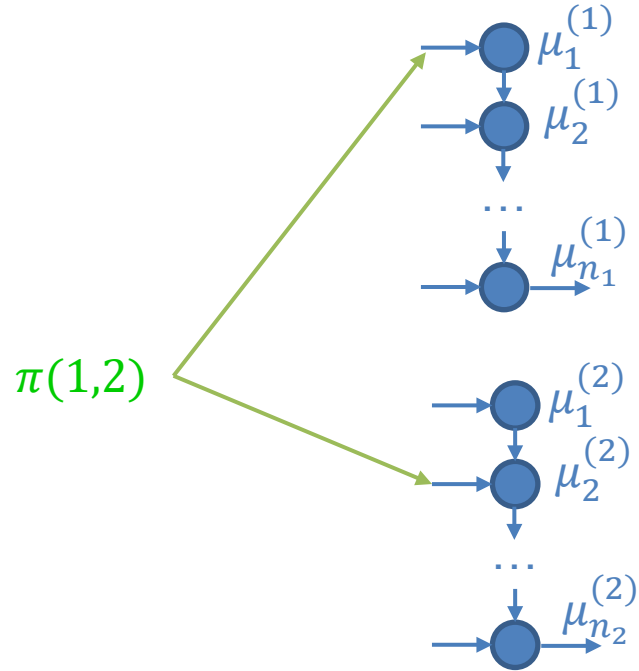
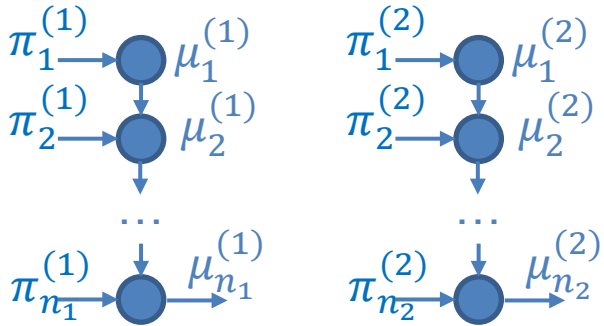
$$L(s) = \frac{\mu}{\mu + s}$$

we assume  $\mu = 1$  (normalized)

- $\mathbf{m}(i)$  sojourn time depending on entry state  $i$
- $\mathbf{a}(i)$  sojourn time depending on exit state  $i$
- $\boldsymbol{\pi}$  entry probabilities,  $\boldsymbol{\psi}$  exit probabilities

# Correlated APHDs

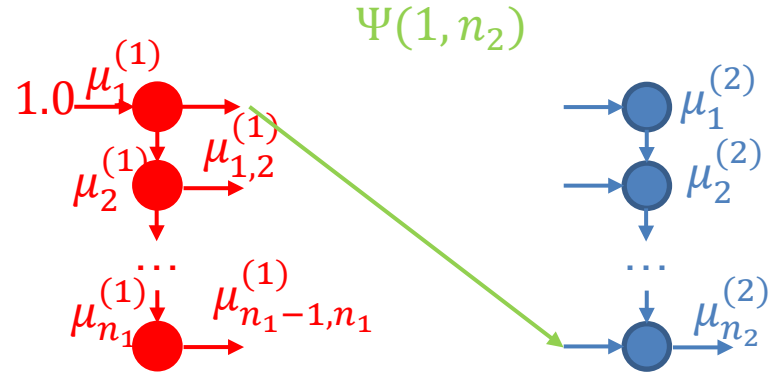
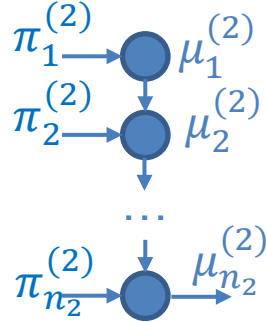
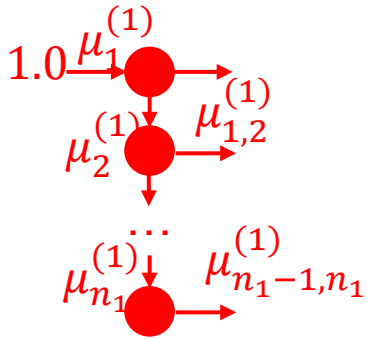
## Parallel composition



$$\sum_{j=1}^{n_2} \pi(i, j) = \pi_i^{(1)} \text{ and } \sum_{i=1}^{n_1} \pi(i, j) = \pi_j^{(2)}$$

# Correlated APHDs

## Sequential composition



$$\sum_{i=1}^{n_2} \psi_i^{(1)} \Psi(i, j) = \pi_j^{(2)} \text{ and } \sum_{j=1}^{n_1} \Psi(i, j) = 1$$

## Minimal/maximal correlation for given APHDs

Linear program (here sequential composition, similar for parallel)

$$\rho_{X,Y}^{+/-}(n_1,n_2) = \min/\max \sum_{i=1}^{n_1} \sum_{j=1}^{n_2} \Psi(i,j) \mathbf{a}_1(i) \mathbf{m}_2(j)$$

$$\text{s. t. } \sum_{i=1}^{n_2} \psi_i^{(1)} \Psi(i,j) = \pi_j^{(2)} \text{ and } \sum_{j=1}^{n_1} \Psi(i,j) = 1$$

**But what are optimal APHD representations for given  $n_1$  and  $n_2$ ?**

## APHD used by Bladt/Nielson 2010 for sequential composition

- Defined for arbitrary  $n$  with  $\mu_i = i$  and  $\pi_i = 1/n$  or  $\mu_{i,i+1} = 1$
- We obtain

$$\rho_{X,X}^{+(n)} = 1 - \frac{1}{n} \sum_{i=1}^n \frac{1}{i} \text{ such that } \lim_{n \rightarrow \infty} \rho_{X,X}^{+(n)} = 1 \text{ (optimal)}$$

$$\rho_{X,X}^{-(n)} = 1 - \sum_{i=1}^n \frac{1}{i^2} \text{ such that } \lim_{n \rightarrow \infty} \rho_{X,X}^{-(n)} = 1 - \frac{\pi^2}{6} \text{ (optimal)}$$

Can we do better?

## Positive correlation

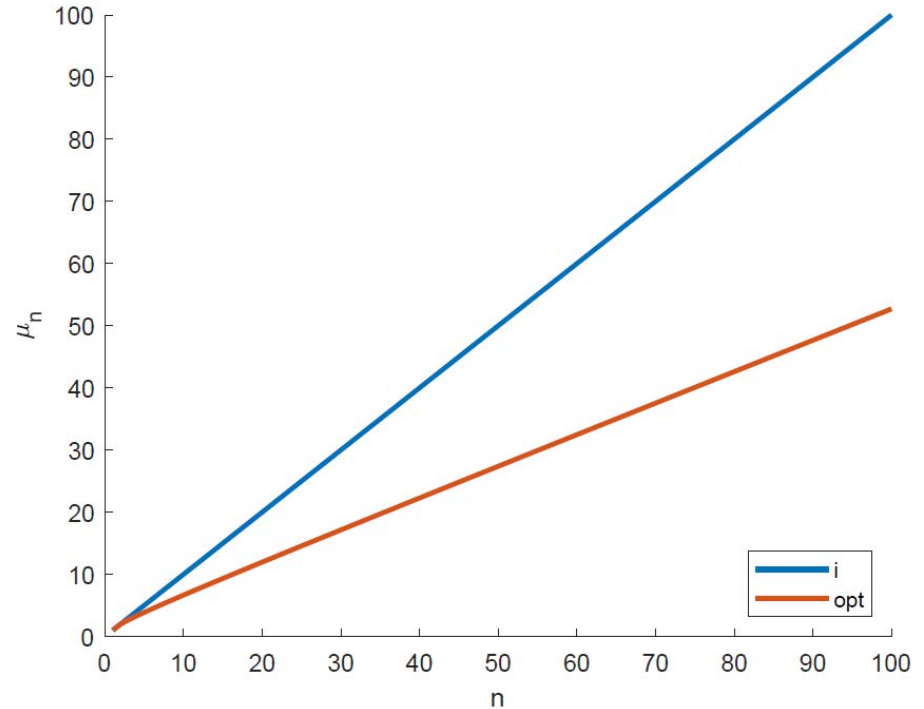
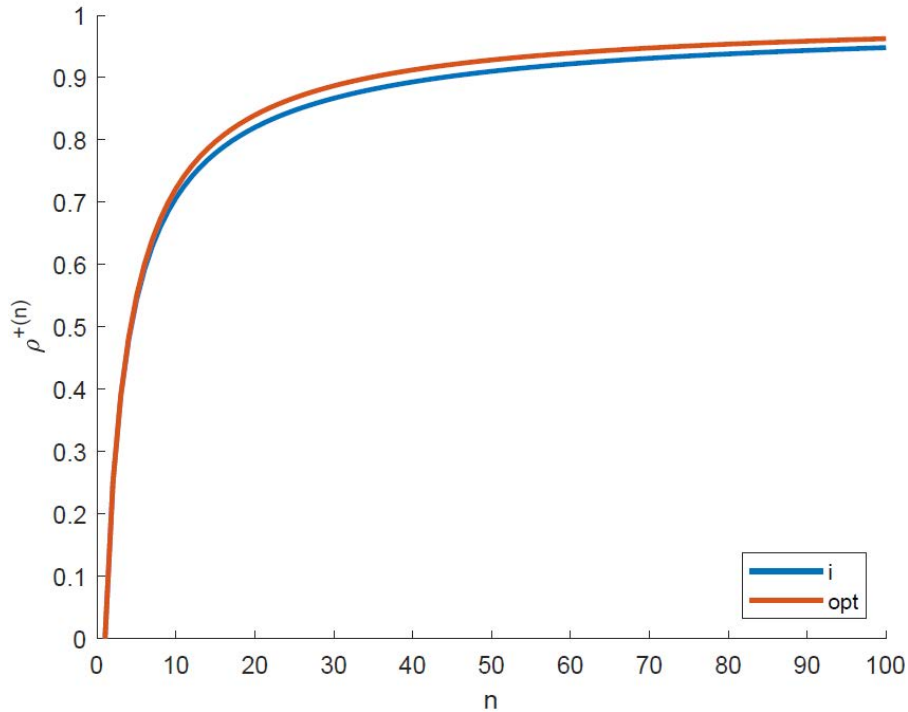
- Algorithm:
  - Start with  $n = 1$  and  $\mu_1 = 1$
  - Compute consecutively  $\mu_{n+1}$  from  $\mu_1, \dots, \mu_n$  resulting in

$$\mu_{n+1} = \frac{2}{1 - \rho_{X,X}^{+(n+1)}} \text{ and } \rho_{X,X}^{+(n+1)} = \rho_{X,X}^{+(n)} + 0.25 \left(1 - \rho_{X,X}^{+(n)}\right)^2$$

- Faster convergence towards 1 than rates  $i$
- Local conditions for optimality are observed



# Positive correlation

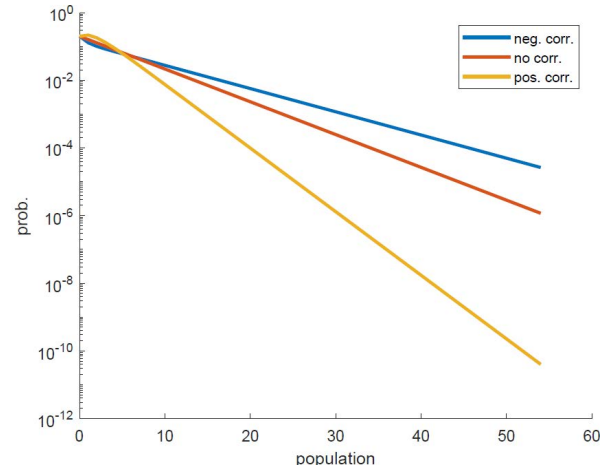
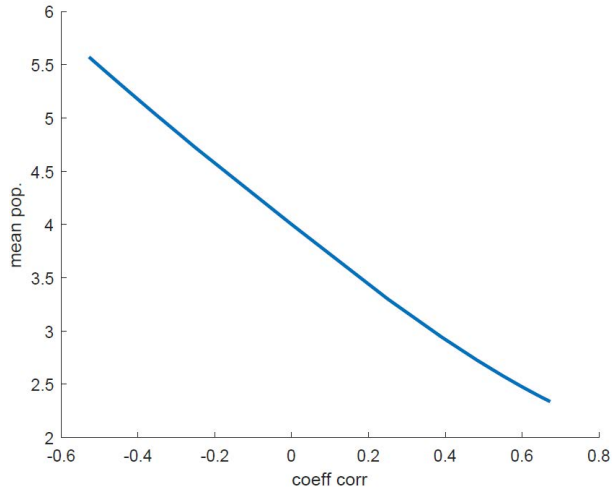


## Negative correlation

- Optimization problem with additional degree of freedom
- Consecutive computation of rates does not work
- Rates  $i$  observe local conditions for optimality
- Result is optimal for  $n = 2$  but for  $n = 3$  a better representation could be found

## An example

- APHD/APHD/1 (M/M/1) queue where inter-arrival and service times are correlated (matrix geometric solution with block size  $O(n_{arrival}n_{service}^2)$ )



## Conclusion

- An approach to describe correlated exponential distributions by acyclic phase type distributions
- Improved representation for positive correlation
- More detailed results can be found in <https://arxiv.org/pdf/2108.12223>
  - Building blocks for general PHDs
  - Extension to Markovian Arrival Processes