



## **Comparing Systems**

3

5

1

## **How to Compare Systems**

Perform the same number of replications for both systems and record the desired result sequences  $Y_1$  and  $Y_2$ , then compute confidence intervals.

#### Approach 1:

- Compute C.I. for each sequence separately
- If the two C.I. do not overlap, evaluate positions

Problem: difficult to interpret combined level of confidence

#### Approach 2:

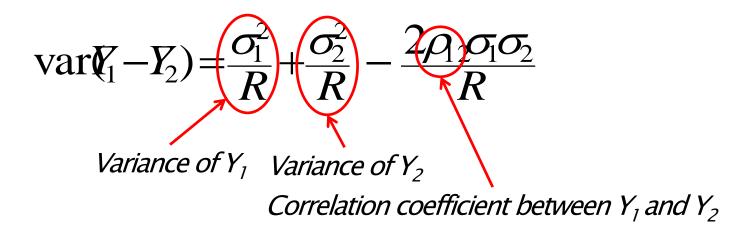
- Compute single C.I. for the difference sequence
- If C.I. does not include the origin, evaluate position

Y <sub>1</sub> -Y <sub>2</sub>	
-2	
-1	
1	

## Correlated Sampling - Why does it help?

#### Approach 2 is easier to compute and potentially more efficient

- C.I. width depends on data variance
- So, for A2 it depends on the variance of the differences:



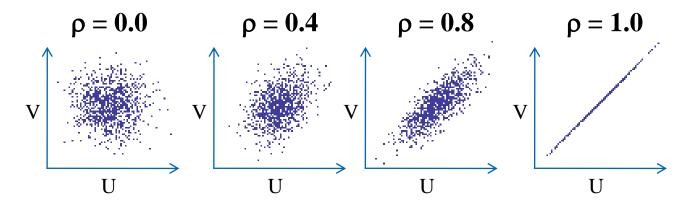
→ If results between both models correlate, variance and width of difference C.I. get smaller

Consequence: Fewer replications for desired level of accuracy

### **Review: Correlation**

## The correlation coefficient $\rho$ describes the statistical relationship between two data sequences U and V

•  $\rho > 0$  for a positive linear relationship between U and V



#### How do we create a positive correlation?

- Ensure that the models to be compared behave similarly in the same replication
- Do NOT manipulate the models or re-order the results!

<b>Y</b> <sub>1</sub>	Y <sub>2</sub>
100	105
345	314
57	64

## Correlated Sampling -Achieving a Positive Correlation

#### Idea:

Use the same random numbers in both models

#### More precisely:

- For each model, each replication uses different random numbers
- In each replication, both models use the same RNs

#### Then,

- Different replications will be independent (Required for C.I.)
- Every pair of results  $Y_{r1}$ ,  $Y_{r2}$  will be positively correlated

## **Correlated Sampling**

#### Algorithm for comparing two systems:

• Compute 
$$D_r = Y_{r1} - Y_{r2}$$

• Compute 
$$D = \frac{1}{R} \sum_{r=1}^{n} D_r$$

• Compute 
$$S^2 = \frac{1}{R-1} \sum_{r} (D_r - D)^2$$

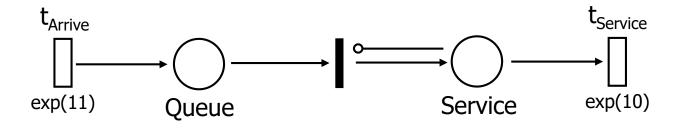
• Compute 
$$\sigma = \frac{S}{\sqrt{R}}$$

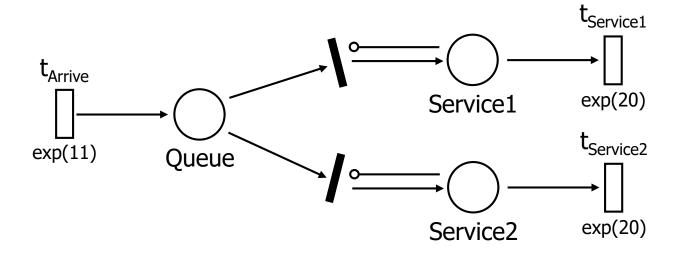
• Choose  $\alpha$  and compute the confidence interval



## **Example**

### Example: Compare one- and two-server models







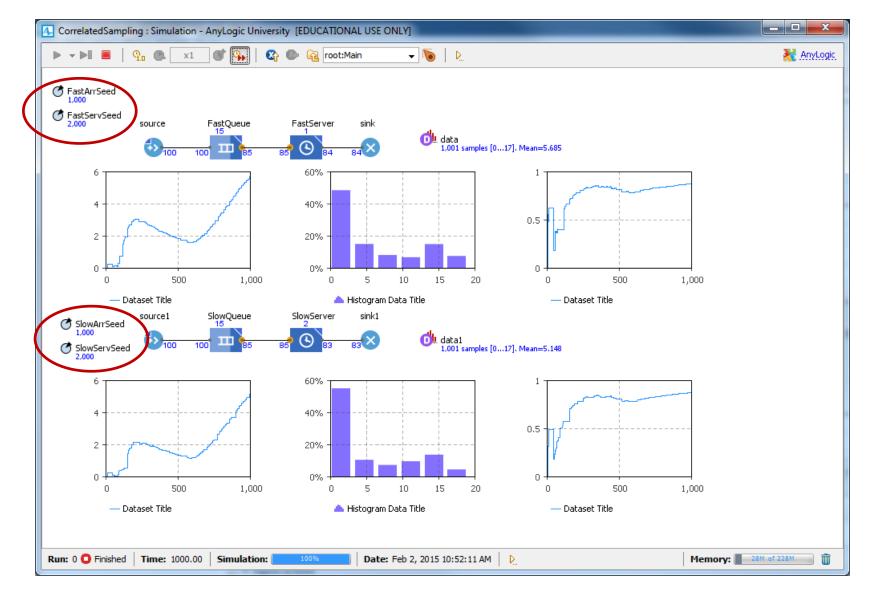
## Example

Do a simulation study to make the comparison

#### We will do the following experiment:

- Compute average queue length at T = 10,000
- Use R = 15 replications

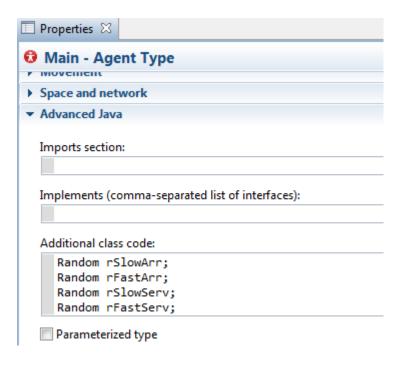
## **AnyLogic Model Using Correlated Sampling**





## Controlling Random Numbers in AnyLogic

Declare four separate random number streams
Initialize the random number streams in a function



```
initRandom - Function

Returns value

▶ Arguments

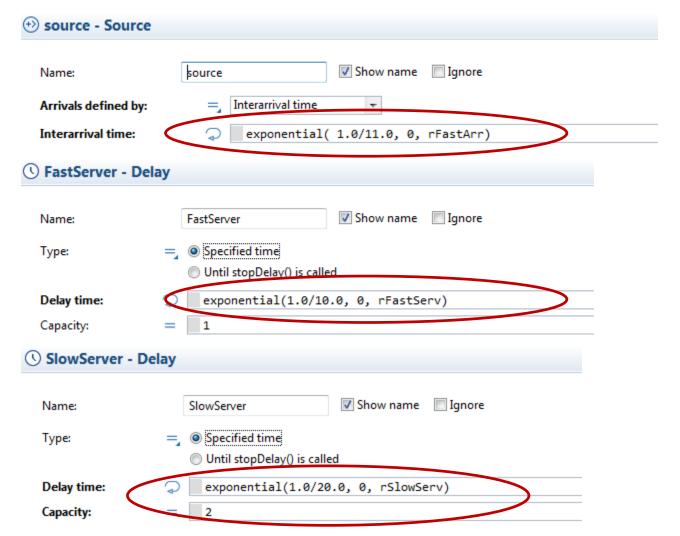
▼ Function body

rSlowArr = new Random(SlowArrSeed);
rFastArr = new Random(FastArrSeed);
rSlowServ = new Random(SlowServSeed);
rFastServ = new Random(FastServSeed);
```



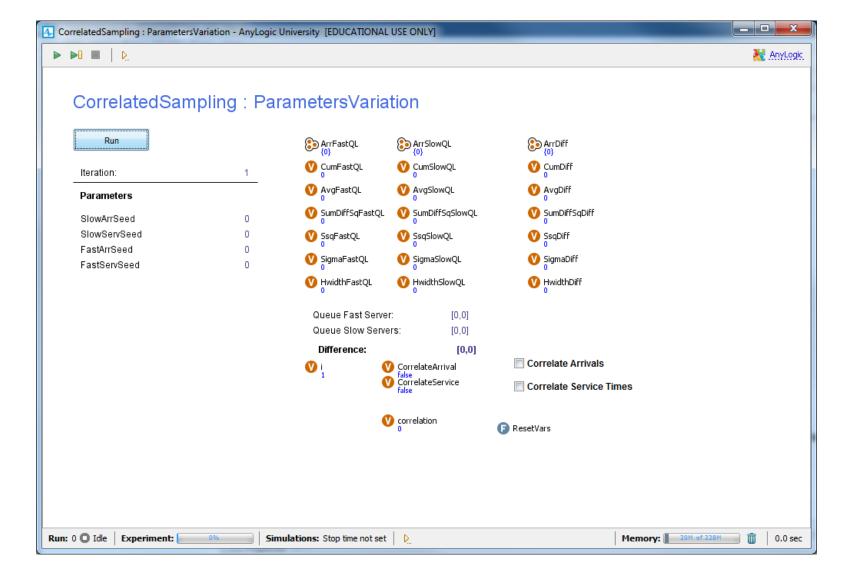
## Controlling Random Numbers in AnyLogic

Assign specific random number stream to each random variable





## **Experiment Setup**



## Controlling Random Numbers in AnyLogic

# Set seeds and initialize the random number streams from the experiment

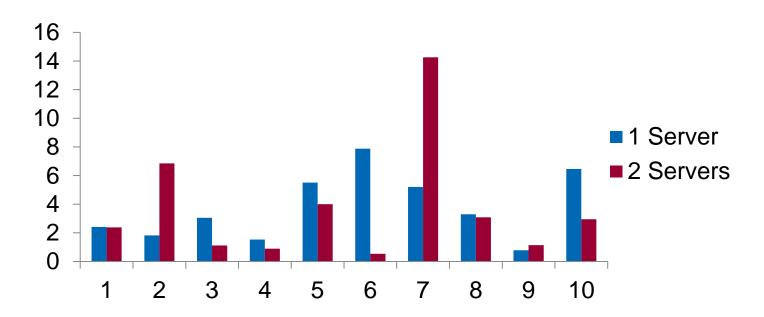
- Use the same seeds for correlated runs
- Use different seeds for independent runs

#### Before simulation run:

## **Experimental Results**

#### Individual replication results, unsynchronised

#### Unsynchronised

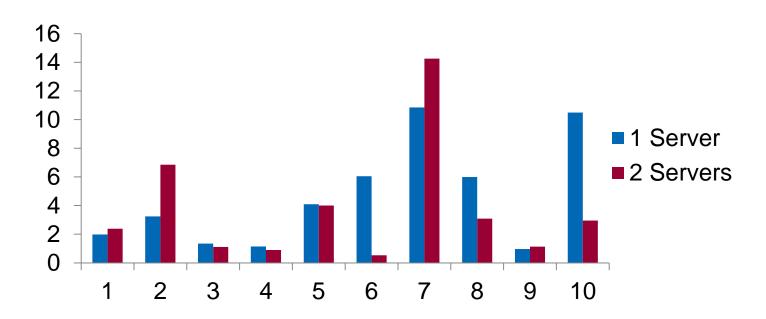




## **Experimental Results**

Individual replication results, only arrivals synchronised

#### **Arrivals Synchronised**

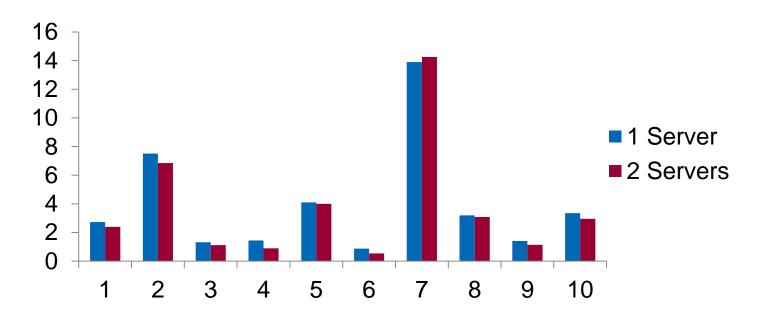




## **Experimental Results**

#### Individual replication results, fully synchronised

### **Fully Synchronised**





## **Results**

## For 100 Replications:

	Sample			C.I. ( $\alpha = 0.05$ )
RN	Corr.	Variance	Width	Interval
Independent	0.02	6.46	1.01	[-0.90, 0.11] -1 0 +1
Arrivals Synchronized	0.54	2.75	0.66	[-0.54, 0.12] -1 0 +1
Fully Synchronized	0.99	0.05	0.09	[-0.35, -0.26] -1 0 +1