

sixth EDITION

SIMULATION with Arena

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Fundamental Simulation Concepts

Chapter 2

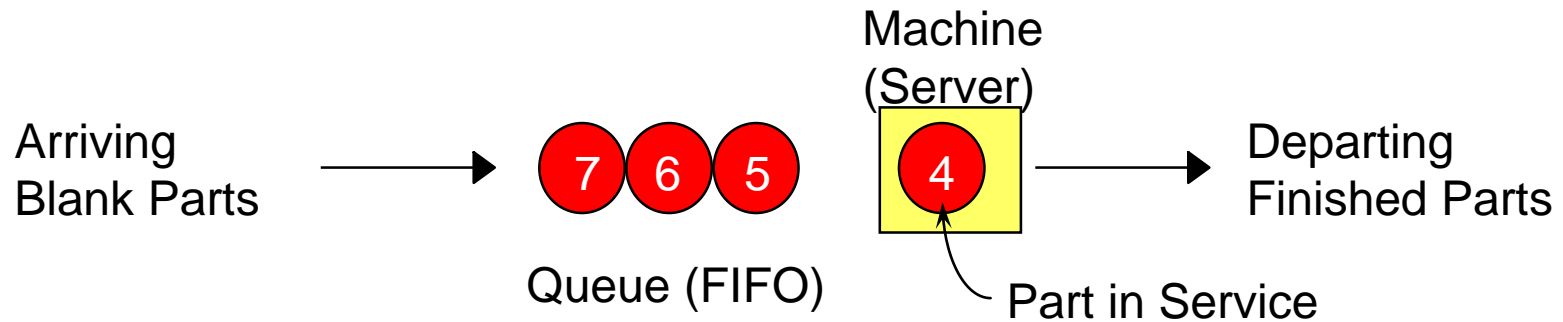
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What We'll Do ...

- **Underlying ideas, methods, and issues in simulation**
- **Software-independent (setting up for Arena)**
- **Example of a simple processing system**
 - Decompose problem
 - Terminology
 - Simulation by hand
 - Some basic statistical issues
- **Spreadsheet simulation**
 - Simple static, dynamic models
- **Overview of a simulation study**

The System:

A Simple Processing System



- **General intent:**
 - Estimate expected production
 - Waiting time in queue, queue length, proportion of time machine is busy
- **Time units**
 - Can use different units in different places ... must declare
 - Be careful to check units when specifying inputs
 - Declare *base time units* for internal calculations, outputs
 - Be reasonable (interpretation, roundoff error)

Model Specifics

- Initially (time 0) empty and idle
- Base time units: minutes
- Input data (assume given for now ...), in minutes:

Part Number	Arrival Time	Interarrival Time	Service Time
1	0.00	1.73	2.90
2	1.73	1.35	1.76
3	3.08	0.71	3.39
4	3.79	0.62	4.52
5	4.41	14.28	4.46
6	18.69	0.70	4.36
7	19.39	15.52	2.07
8	34.91	3.15	3.36
9	38.06	1.76	2.37
10	39.82	1.00	5.38
11	40.82	.	.
.	.	.	.
.	.	.	.

- Stop when 20 minutes of (simulated) time have passed

Goals of Study:

Output Performance Measures

- **Total production** of parts over run (P)
- **Average waiting time** of parts in queue:

$$\frac{\sum_{i=1}^N WQ_i}{N}$$

N = no. of parts completing queue wait

WQ_i = waiting time in queue of i th part

Know: $WQ_1 = 0$ (why?)

$N \geq 1$ (why?)

- **Maximum waiting time** of parts in queue:

$$\max_{i=1, \dots, N} WQ_i$$

Goals of Study:

Output Performance Measures (cont'd.)

- **Time-average number of parts in queue:**

$$\frac{\int_0^{20} Q(t) dt}{20}$$

$Q(t)$ = number of parts in queue
at time t

- **Maximum number of parts in queue:** $\max_{0 \leq t \leq 20} Q(t)$
- **Average and maximum total time in system** of parts (a.k.a. **cycle time**):

$$\frac{\sum_{i=1}^P TS_i}{P},$$

$$\max_{i=1, \dots, P} TS_i$$

TS_i = time in system of part i

Goals of Study:

Output Performance Measures (cont'd.)

- **Utilization** of machine (proportion of time busy)

$$\frac{\int_0^{20} B(t) dt}{20}, \quad B(t) = \begin{cases} 1 & \text{if machine is busy at time } t \\ 0 & \text{if machine is idle at time } t \end{cases}$$

- Many others possible (information overload?)

Analysis Options

- **Educated guessing**

- Average interarrival time = 4.08 minutes
- Average service time = 3.46 minutes
- So (on average) parts are being processed faster than they arrive
 - System has a chance of operating in a stable way in long run, i.e., might not “explode”
 - If all interarrivals and service times were exactly at their mean, there would never be a queue
 - But data clearly exhibit variability, so a queue could form
- If we'd had average interarrival < average service time, and this persisted, then queue would explode
- Truth — between these extremes
- Guessing has its limits ...

Analysis Options (cont'd.)

- **Queueing theory**

- Requires additional assumptions about model
- Popular, simple model: *M/M/1 queue*
 - Interarrival times ~ exponential
 - Service times ~ exponential, indep. of interarrivals
 - Must have $E(\text{service}) < E(\text{interarrival})$
 - Steady-state (long-run, forever)
 - Exact analytic results; e.g., average waiting time in queue is

$$\frac{\mu_S^2}{\mu_A - \mu_S}, \quad \begin{array}{l} \mu_A = E(\text{interarrival time}) \\ \mu_S = E(\text{service time}) \end{array}$$

- Problems: validity, estimating means, time frame
- Often useful as first-cut approximation

Mechanistic Simulation

- **Individual operations (arrivals, service times) will occur exactly as in reality**
- **Movements, changes occur at right “times,” in right order**
- **Different pieces interact**
- **Install “observers” to get output performance measures**
- **Concrete, “brute-force” analysis approach**
- **Nothing mysterious or subtle**
 - But a lot of details, bookkeeping
 - Simulation software keeps track of things for you

Pieces of a Simulation Model

- **Entities**

- “Players” that move around, change status, affect and are affected by other entities
- *Dynamic objects* — get created, move around, leave (maybe)
- Usually represent “real” things
 - Our model: entities are parts
- Can have “fake” entities for modeling “tricks”
 - Breakdown demon, break angel
 - Though Arena has built-in ways to model these examples directly
- Usually have multiple *realizations* floating around
- Can have different types of entities concurrently
- Usually, identifying types of entities is first thing to do in building model

Pieces of a Simulation Model (cont'd.)

- **Attributes**

- Characteristic of all entities: describe, differentiate
- All entities have same attribute “slots” but different values for different entities, for example:
 - Time of arrival
 - Due date
 - Priority
 - Color
- Attribute value tied to a specific entity
- Like “local” (to entities) variables
- Some automatic in Arena, some you define

Pieces of a Simulation Model (cont'd.)

- **(Global) *Variables***
 - Reflects a characteristic of whole model, not of specific entities
 - Used for many different kinds of things
 - Travel time between all station pairs
 - Number of parts in system
 - Simulation clock (built-in Arena variable)
 - Name, value of which there's only one copy for whole model
 - Not tied to entities
 - Entities can access, change variables
 - Writing on wall (rewriteable)
 - Some built-in by Arena, you can define others

Pieces of a Simulation Model (cont'd.)

- **Resources**

- What entities compete for
 - People
 - Equipment
 - Space
- Entity *seizes* a resource, uses it, *releases* it
- Think of a *resource being assigned to an entity*, rather than an entity “belonging to” a resource
- “A” resource can have several *units* of capacity
 - Seats at a table in a restaurant
 - Identical ticketing agents at an airline counter
- Number of units of resource can be changed during simulation

Pieces of a Simulation Model (cont'd.)

- **Queues**

- Place for entities to wait when they can't move on (maybe since resource they want to seize is not available)
- Have names, often tied to a corresponding resource
- Can have a finite capacity to model limited space — have to model what to do if an entity shows up to a queue that's already full
- Usually watch length of a queue, waiting time in it

Pieces of a Simulation Model (cont'd.)

- ***Statistical accumulators***

- Variables that “watch” what’s happening
- Depend on output performance measures desired
- “Passive” in model — don’t participate, just watch
- Many are automatic in Arena, but some you may have to set up and maintain during simulation
- At end of simulation, used to compute final output performance measures

Pieces of a Simulation Model (cont'd.)

- **Statistical accumulators for simple processing system**
 - Number of parts produced so far
 - Total of waiting times spent in queue so far
 - No. of parts that have gone through queue
 - Max time in queue we've seen so far
 - Total of times spent in system
 - Max time in system we've seen so far
 - Area so far under queue-length curve $Q(t)$
 - Max of $Q(t)$ so far
 - Area so far under server-busy curve $B(t)$

Simulation Dynamics: Event-Scheduling “World View”

- Identify characteristic *events*
- Decide on *logic* for each type of event to:
 - Effect *state changes* for each event type
 - Observe statistics
 - Update times of future events (maybe of this type, other types)
- Keep a simulation *clock*, future *event calendar*
- *Jump* from one event to the next, process, observe statistics, update event calendar
- Must specify an appropriate *stopping rule*
- Usually done with general-purpose programming language (C++, Java, Matlab, FORTRAN, etc.)

Events for the Simple Processing System

- **Arrival** of a new part to system
 - Update time-persistent statistical accumulators (from last event to now)
 - Area under $Q(t)$
 - Max of $Q(t)$
 - Area under $B(t)$
 - “Mark” arriving part with current time (use later)
 - If machine is idle:
 - Start processing (schedule departure), Make machine busy, Tally waiting time in queue (0)
 - Else (machine is busy):
 - Put part at end of queue, increase queue-length variable
 - Schedule next arrival event

Events for the Simple Processing System (cont'd.)

- **Departure** (when a service is completed)
 - Increment number-produced stat accumulator
 - Compute & tally time in system (now – time of arrival)
 - Update time-persistent statistics (as in arrival event)
 - If queue is non-empty:
 - Take first part out of queue, compute & tally its waiting time in queue, begin service (schedule departure event)
 - Else (queue is empty):
 - Make machine idle (Note: there will be no departure event scheduled on future events calendar, which is as desired)

Events for the Simple Processing System (cont'd.)

- ***The End***
 - Update time-persistent statistics (to end of simulation)
 - Compute final output performance measures using current (= final) values of statistical accumulators
- **After each event, event calendar's top record is removed to see what time it is, what to do**
- **Also must initialize everything**

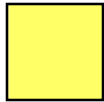
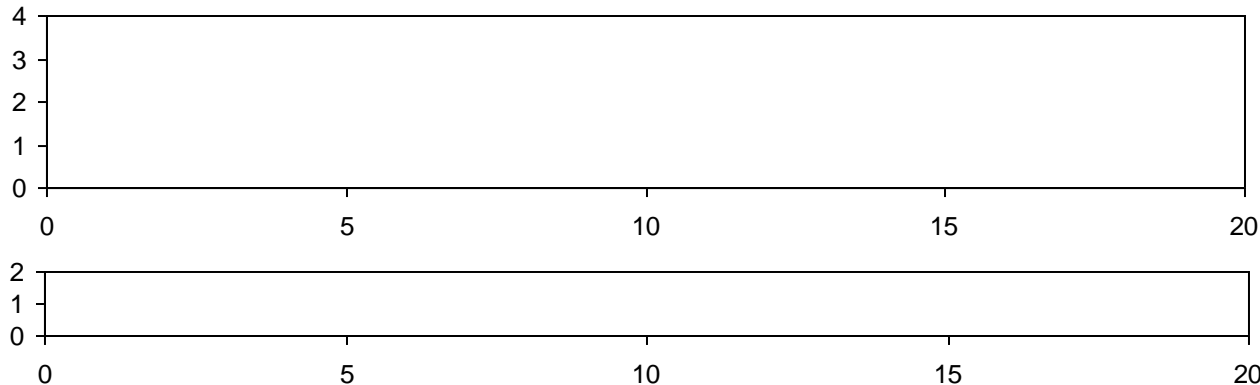
Some Additional Specifics for the Simple Processing System

- **Simulation clock variable (internal in Arena)**
- **Event calendar: list of event *records*:**
 - [Entity No., Event Time, Event Type]
 - Keep *ranked* in increasing order on Event Time
 - Next event always in top record
 - Initially, schedule first Arrival, The End (Dep.?)
- **State variables: describe current status**
 - Server status $B(t) = 1$ for busy, 0 for idle
 - Number of customers in queue $Q(t)$
 - Times of arrival of each customer now in queue (a list of random length)

Simulation by Hand

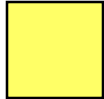
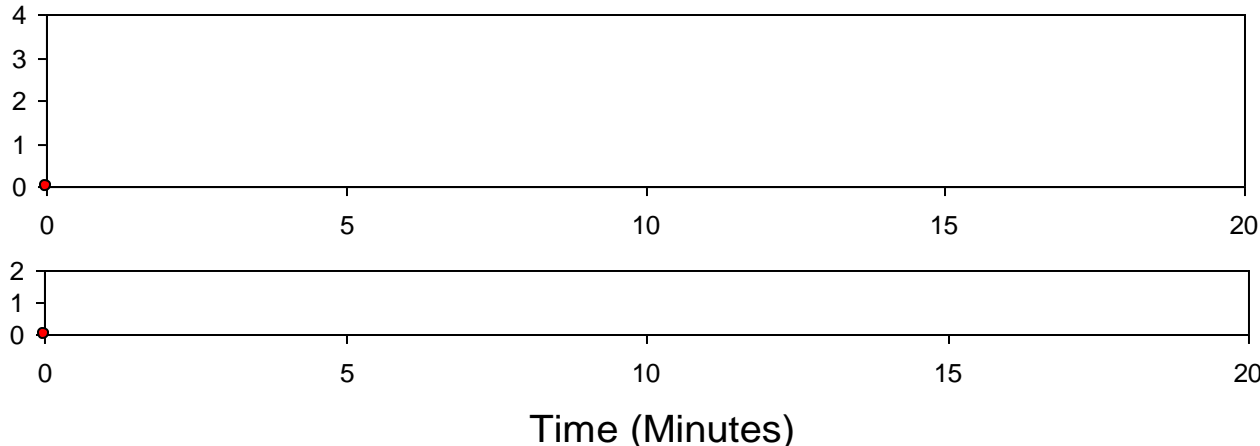
- **Manually track state variables, statistical accumulators**
- **Use “given” interarrival, service times**
- **Keep track of event calendar**
- **“Lurch” clock from one event to next**
- **Will omit times in system, “max” computations here (see text for complete details)**

Simulation by Hand: Setup

System <div></div>	Clock	$B(t)$	$Q(t)$	Arrival times of custs. in queue	Event calendar
Number of completed waiting times in queue	Total of waiting times in queue		Area under $Q(t)$		Area under $B(t)$
<div><div>$Q(t)$ graph</div><div>$B(t)$ graph</div></div>	<div></div>				
Interarrival times	1.73, 1.35, 0.71, 0.62, 14.28, 0.70, 15.52, 3.15, 1.76, 1.00, ...				
Service times	2.90, 1.76, 3.39, 4.52, 4.46, 4.36, 2.07, 3.36, 2.37, 5.38, ...				


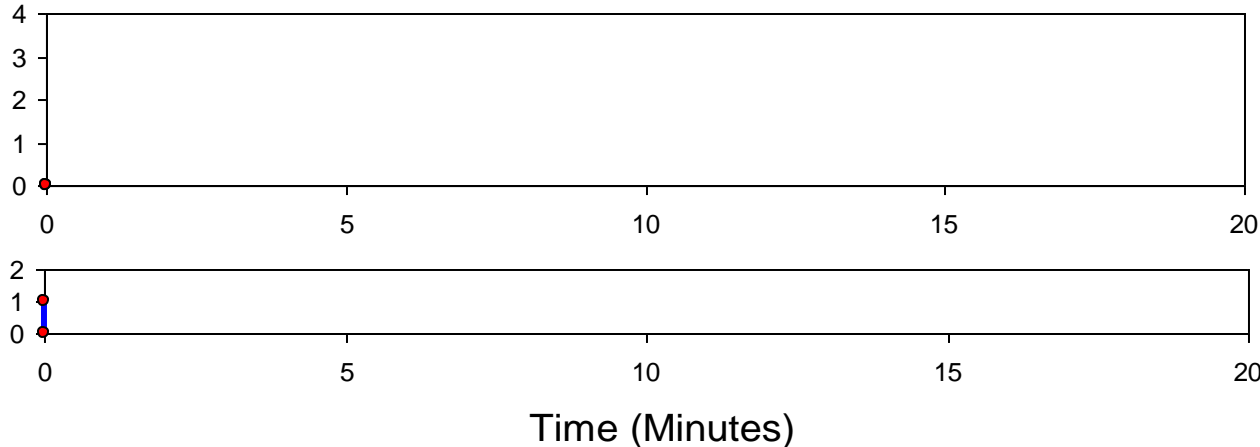
Simulation by Hand:

$t = 0.00$, Initialize

System <div></div>	Clock 0.00	$B(t)$ 0	$Q(t)$ 0	Arrival times of custs. in queue <empty>	Event calendar [1, 0.00, Arr] [-, 20.00, End]
Number of completed waiting times in queue 0	Total of waiting times in queue 0.00			Area under $Q(t)$ 0.00	Area under $B(t)$ 0.00
<div>$Q(t)$ graph $B(t)$ graph</div>	<div></div>				
Interarrival times	1.73, 1.35, 0.71, 0.62, 14.28, 0.70, 15.52, 3.15, 1.76, 1.00, ...				
Service times	2.90, 1.76, 3.39, 4.52, 4.46, 4.36, 2.07, 3.36, 2.37, 5.38, ...				

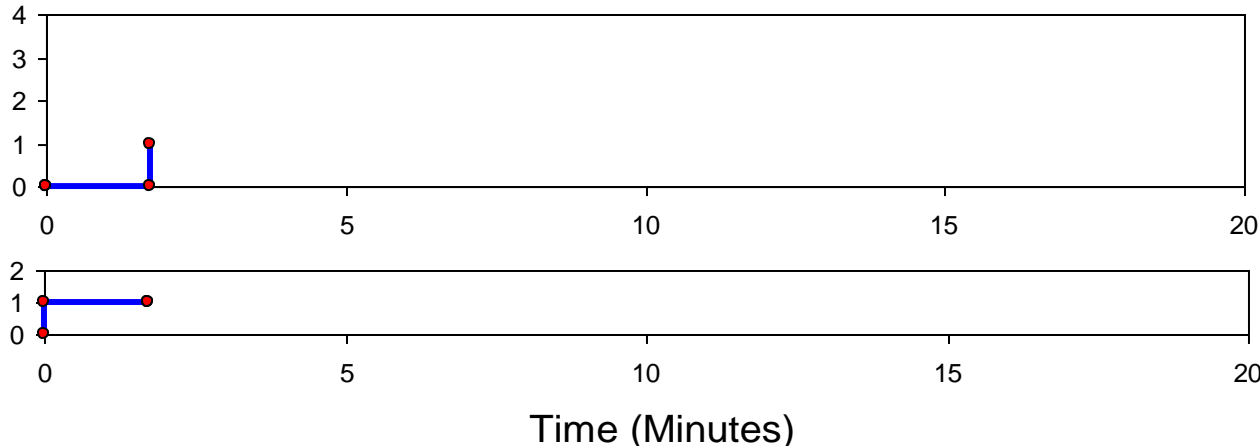
Simulation by Hand:

$t = 0.00$, Arrival of Part 1

System <div></div>	Clock 0.00	$B(t)$ 1	$Q(t)$ 0	Arrival times of custs. in queue <empty>	Event calendar [2, 1.73, Arr] [1, 2.90, Dep] [-, 20.00, End]
Number of completed waiting times in queue 1	Total of waiting times in queue 0.00		Area under $Q(t)$ 0.00		Area under $B(t)$ 0.00
<div>$Q(t)$ graph $B(t)$ graph</div>	<div></div>				
Interarrival times	1.73 , 1.35, 0.71, 0.62, 14.28, 0.70, 15.52, 3.15, 1.76, 1.00, ...				
Service times	2.90 , 1.76, 3.39, 4.52, 4.46, 4.36, 2.07, 3.36, 2.37, 5.38, ...				

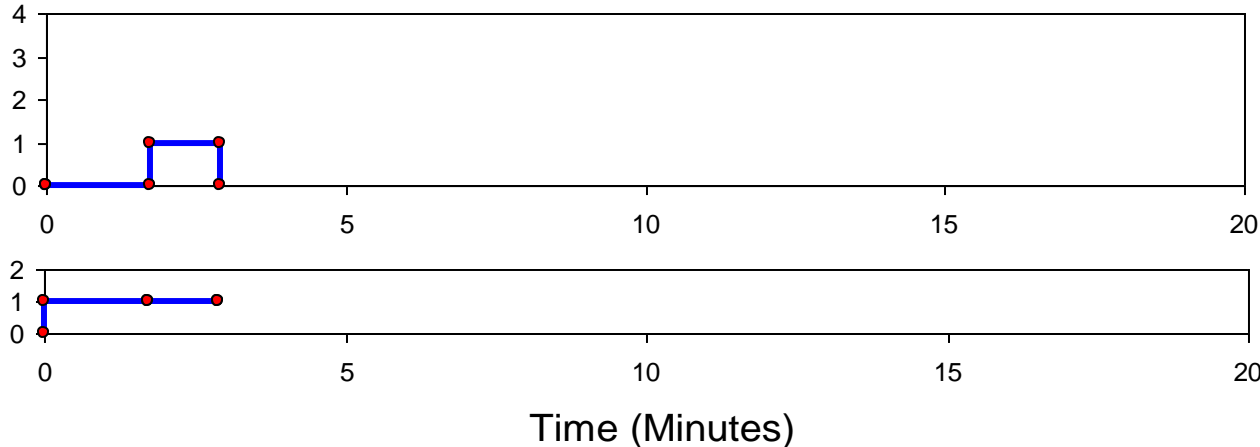
Simulation by Hand:

$t = 1.73$, Arrival of Part 2

System <div><div>2</div><div>1</div></div>	Clock 1.73	$B(t)$ 1	$Q(t)$ 1	Arrival times of custs. in queue (1.73)	Event calendar [1, 2.90, Dep] [3, 3.08, Arr] [-, 20.00, End]
Number of completed waiting times in queue 1	Total of waiting times in queue 0.00		Area under $Q(t)$ 0.00		Area under $B(t)$ 1.73
<div><div>$Q(t)$ graph</div><div>$B(t)$ graph</div></div>	<div></div> <div>Time (Minutes)</div>				
Interarrival times	1.73, 1.35 , 0.71, 0.62, 14.28, 0.70, 15.52, 3.15, 1.76, 1.00, ...				
Service times	2.90 , 1.76, 3.39, 4.52, 4.46, 4.36, 2.07, 3.36, 2.37, 5.38, ...				

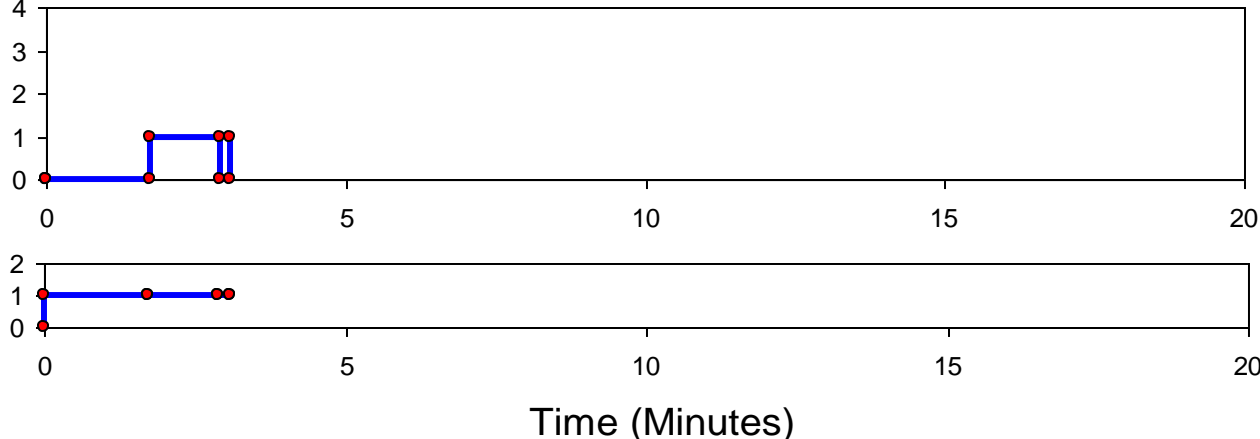
Simulation by Hand:

$t = 2.90$, Departure of Part 1

System <div><div></div><div>2</div></div>	Clock 2.90	$B(t)$ 1	$Q(t)$ 0	Arrival times of custs. in queue <empty>	Event calendar [3, 3.08, Arr] [2, 4.66, Dep] [-, 20.00, End]
Number of completed waiting times in queue 2	Total of waiting times in queue 1.17		Area under $Q(t)$ 1.17		Area under $B(t)$ 2.90
<div><div>$Q(t)$ graph</div><div>$B(t)$ graph</div></div>	<div></div> <div>Time (Minutes)</div>				
Interarrival times	1.73, 1.35 , 0.71, 0.62, 14.28, 0.70, 15.52, 3.15, 1.76, 1.00, ...				
Service times	2.90, 1.76 , 3.39, 4.52, 4.46, 4.36, 2.07, 3.36, 2.37, 5.38, ...				

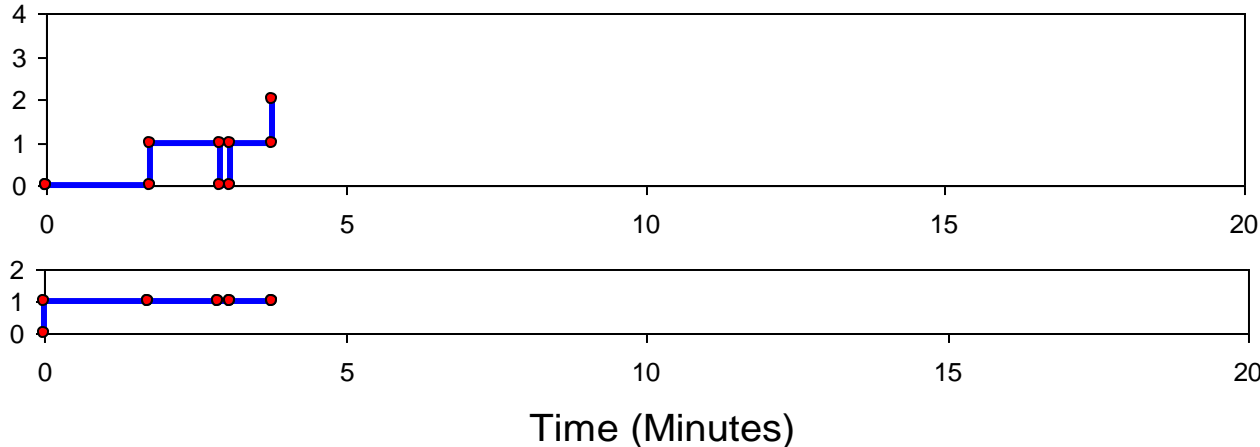
Simulation by Hand:

$t = 3.08$, Arrival of Part 3

System <div><div>3</div><div>2</div></div>	Clock 3.08	$B(t)$ 1	$Q(t)$ 1	Arrival times of custs. in queue (3.08)	Event calendar [4, 3.79, Arr] [2, 4.66, Dep] [-, 20.00, End]
Number of completed waiting times in queue 2	Total of waiting times in queue 1.17		Area under $Q(t)$ 1.17		Area under $B(t)$ 3.08
<div><div>$Q(t)$ graph</div><div>$B(t)$ graph</div></div>	<div><p style="text-align: center;">Time (Minutes)</p></div>				
Interarrival times	1.73, 1.35, 0.71 , 0.62, 14.28, 0.70, 15.52, 3.15, 1.76, 1.00, ...				
Service times	2.90, 1.76 , 3.39, 4.52, 4.46, 4.36, 2.07, 3.36, 2.37, 5.38, ...				

Simulation by Hand:

$t = 3.79$, Arrival of Part 4

System <div><div>4</div><div>3</div><div>2</div></div>	Clock 3.79	$B(t)$ 1	$Q(t)$ 2	Arrival times of custs. in queue (3.79, 3.08)	Event calendar [5, 4.41, Arr] [2, 4.66, Dep] [-, 20.00, End]
Number of completed waiting times in queue 2	Total of waiting times in queue 1.17		Area under $Q(t)$ 1.88		Area under $B(t)$ 3.79
<div><div>$Q(t)$ graph</div><div>$B(t)$ graph</div></div>	<div></div> <div>Time (Minutes)</div>				
Interarrival times	1.73, 1.35, 0.71, 0.82 , 14.28, 0.70, 15.52, 3.15, 1.76, 1.00, ...				
Service times	2.90, 1.76 , 3.39, 4.52, 4.46, 4.36, 2.07, 3.36, 2.37, 5.38, ...				

Simulation by Hand:

$t = 4.41$, Arrival of Part 5

System <div><div>5</div><div>4</div><div>3</div><div>2</div></div>	Clock 4.41	$B(t)$ 1	$Q(t)$ 3	Arrival times of custs. in queue (4.41, 3.79, 3.08)	Event calendar [2, 4.66, Dep] [6, 18.69, Arr] [–, 20.00, End]
Number of completed waiting times in queue 2	Total of waiting times in queue 1.17		Area under $Q(t)$ 3.12		Area under $B(t)$ 4.41
<div><div>$Q(t)$ graph</div><div>$B(t)$ graph</div></div>	<div></div>				
Interarrival times	1.73, 1.35, 0.71, 0.62, 14.28 , 0.70, 15.52, 3.15, 1.76, 1.00, ...				
Service times	2.90, 1.76 , 3.39, 4.52, 4.46, 4.36, 2.07, 3.36, 2.37, 5.38, ...				

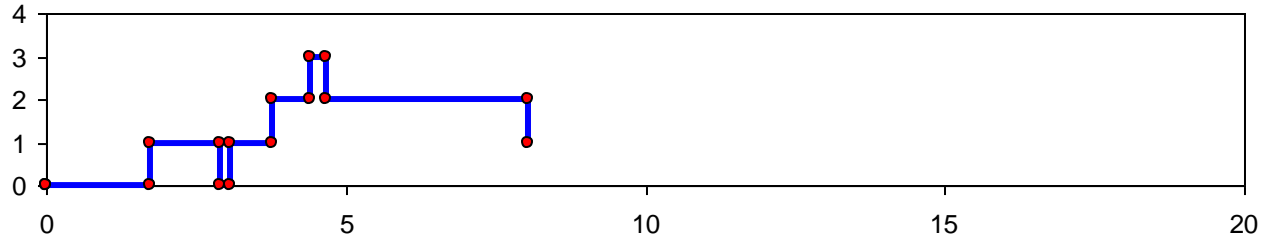
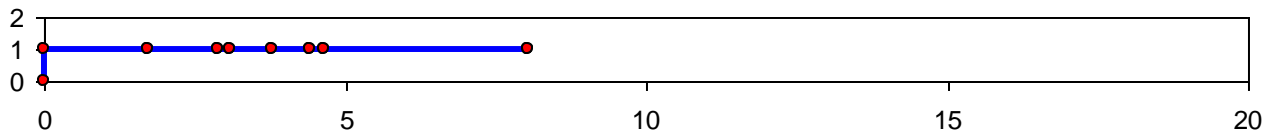
Simulation by Hand:

$t = 4.66$, Departure of Part 2

System <div><div>5</div><div>4</div><div>3</div></div>	Clock 4.66	$B(t)$ 1	$Q(t)$ 2	Arrival times of custs. in queue (4.41, 3.79)	Event calendar [3, 8.05, Dep] [6, 18.69, Arr] [–, 20.00, End]
Number of completed waiting times in queue 3	Total of waiting times in queue 2.75		Area under $Q(t)$ 3.87		Area under $B(t)$ 4.66
<div><div>$Q(t)$ graph</div><div>$B(t)$ graph</div></div>					
Interarrival times	1.73, 1.35, 0.71, 0.62, 14.28 , 0.70, 15.52, 3.15, 1.76, 1.00, ...				
Service times	2.90, 1.76, 3.39 , 4.52, 4.46, 4.36, 2.07, 3.36, 2.37, 5.38, ...				

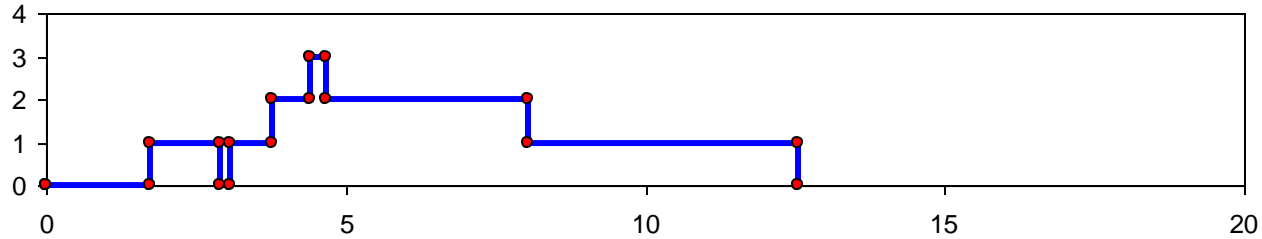
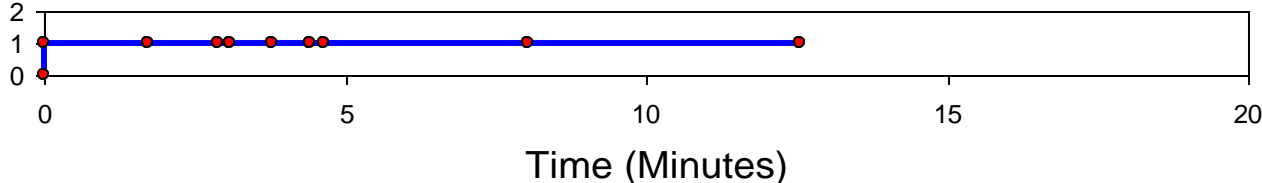
Simulation by Hand:

$t = 8.05$, Departure of Part 3

System <div><div>5</div><div>4</div></div>	Clock 8.05	$B(t)$ 1	$Q(t)$ 1	Arrival times of custs. in queue (4.41)	Event calendar [4, 12.57, Dep] [6, 18.69, Arr] [–, 20.00, End]
Number of completed waiting times in queue 4	Total of waiting times in queue 7.01		Area under $Q(t)$ 10.65		Area under $B(t)$ 8.05
$Q(t)$ graph					
$B(t)$ graph					
Interarrival times	1.73, 1.35, 0.71, 0.62, 14.28 , 0.70, 15.52, 3.15, 1.76, 1.00, ...				
Service times	2.90, 1.76, 3.39, 4.52 , 4.46, 4.36, 2.07, 3.36, 2.37, 5.38, ...				

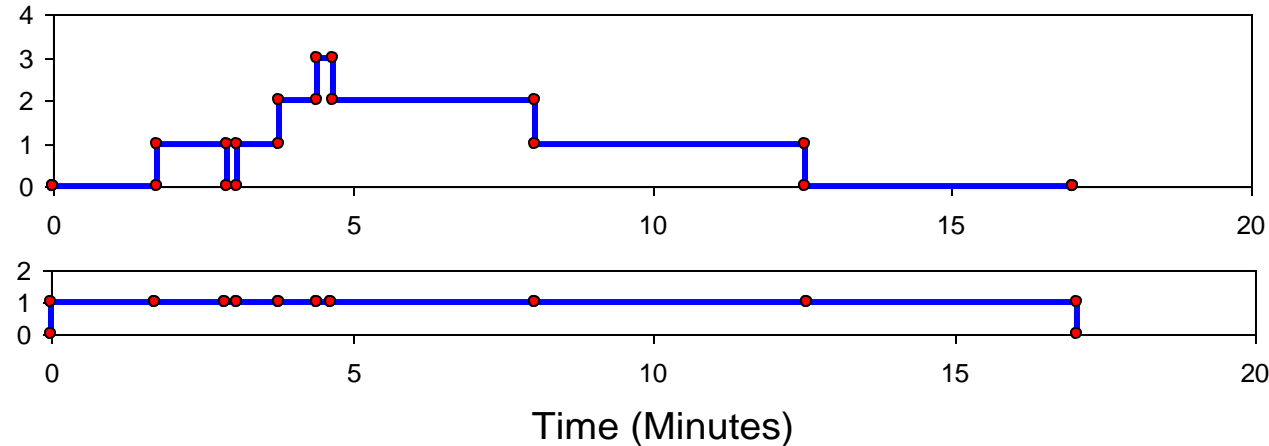
Simulation by Hand:

$t = 12.57$, Departure of Part 4

System <div><div></div><div>5</div></div>	Clock 12.57	$B(t)$ 1	$Q(t)$ 0	Arrival times of custs. in queue ()	Event calendar [5, 17.03, Dep] [6, 18.69, Arr] [–, 20.00, End]
Number of completed waiting times in queue 5	Total of waiting times in queue 15.17		Area under $Q(t)$ 15.17		Area under $B(t)$ 12.57
$Q(t)$ graph					
$B(t)$ graph					
Interarrival times	1.73, 1.35, 0.71, 0.62, 14.28 , 0.70, 15.52, 3.15, 1.76, 1.00, ...				
Service times	2.90, 1.76, 3.39, 4.52, 4.46 , 4.36, 2.07, 3.36, 2.37, 5.38, ...				

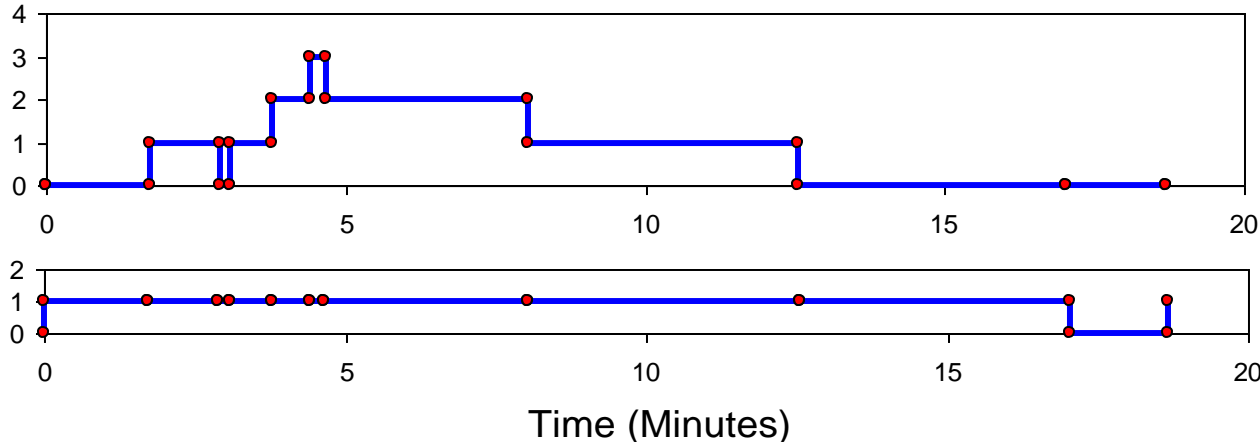
Simulation by Hand:

$t = 17.03$, Departure of Part 5

System <div><div></div></div>	Clock 17.03	$B(t)$ 0	$Q(t)$ 0	Arrival times of custs. in queue ()	Event calendar [6, 18.69, Arr] [−, 20.00, End]
Number of completed waiting times in queue 5	Total of waiting times in queue 15.17		Area under $Q(t)$ 15.17		Area under $B(t)$ 17.03
<div><div>$Q(t)$ graph</div><div>$B(t)$ graph</div></div>	<div></div>				
Interarrival times	1.73, 1.35, 0.71, 0.62, 14.28 , 0.70, 15.52, 3.15, 1.76, 1.00, ...				
Service times	2.90, 1.76, 3.39, 4.52, 4.46 , 4.36, 2.07, 3.36, 2.37, 5.38, ...				

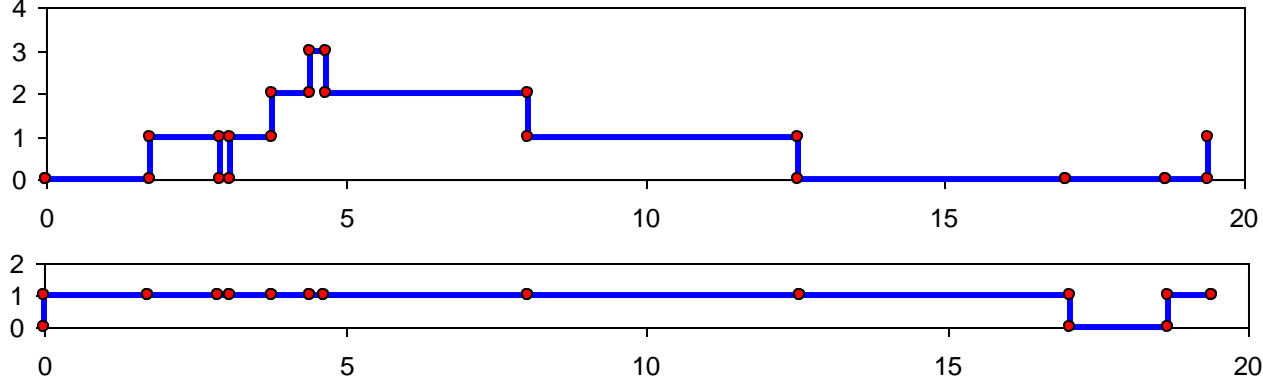
Simulation by Hand:

$t = 18.69$, Arrival of Part 6

System <div><div></div><div>6</div></div>	Clock 18.69	$B(t)$ 1	$Q(t)$ 0	Arrival times of custs. in queue ()	Event calendar [7, 19.39, Arr] [-, 20.00, End] [6, 23.05, Dep]
Number of completed waiting times in queue 6	Total of waiting times in queue 15.17		Area under $Q(t)$ 15.17		Area under $B(t)$ 17.03
<div><div>$Q(t)$ graph</div><div>$B(t)$ graph</div></div>	<div></div> <div>Time (Minutes)</div>				
Interarrival times	1.73, 1.35, 0.71, 0.62, 14.28, 0.70, 15.52, 3.15, 1.76, 1.00, ...				
Service times	2.90, 1.76, 3.39, 4.52, 4.46, 4.36, 2.07, 3.36, 2.37, 5.38, ...				

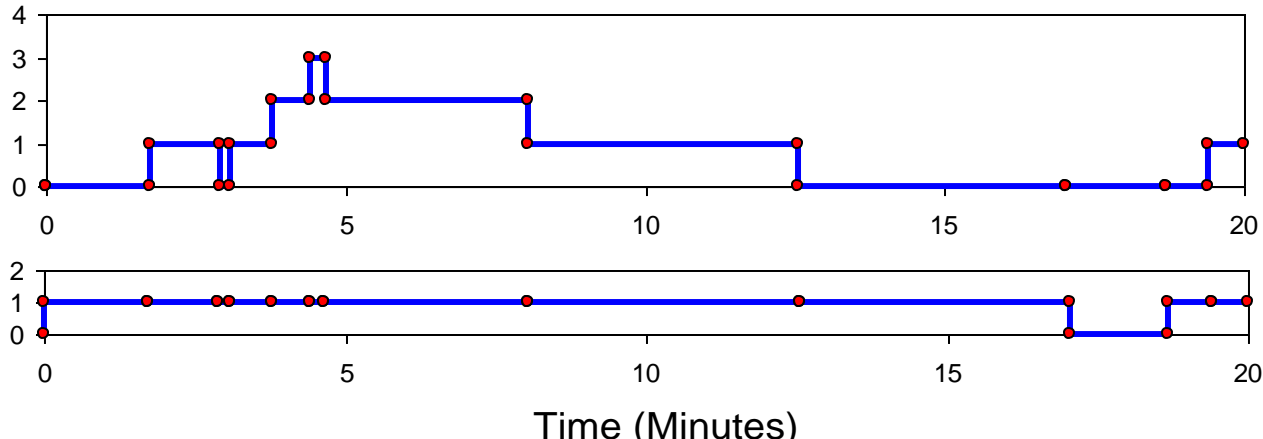
Simulation by Hand:

$t = 19.39$, Arrival of Part 7

System <div><div>7</div><div>6</div></div>	Clock 19.39	$B(t)$ 1	$Q(t)$ 1	Arrival times of custs. in queue (19.39)	Event calendar [-, 20.00, End] [6, 23.05, Dep] [8, 34.91, Arr]
Number of completed waiting times in queue 6	Total of waiting times in queue 15.17		Area under $Q(t)$ 15.17		Area under $B(t)$ 17.73
<div><div>$Q(t)$ graph</div><div>$B(t)$ graph</div></div>					
Interarrival times	1.73 , 1.35 , 0.71 , 0.82 , 14.28 , 0.70 , 15.52 , 3.15, 1.76, 1.00, ...				
Service times	2.90 , 1.76 , 3.39 , 4.52 , 4.46 , 4.36 , 2.07, 3.36, 2.37, 5.38, ...				

Simulation by Hand:

$t = 20.00$, The End

System <div><div>7</div><div>6</div></div>	Clock 20.00	$B(t)$ 1	$Q(t)$ 1	Arrival times of custs. in queue (19.39)	Event calendar [6, 23.05, Dep] [8, 34.91, Arr]
Number of completed waiting times in queue 6	Total of waiting times in queue 15.17		Area under $Q(t)$ 15.78		Area under $B(t)$ 18.34
<div><div>$Q(t)$ graph</div><div>$B(t)$ graph</div></div>	 <p>Time (Minutes)</p>				
Interarrival times	1.73 , 1.35 , 0.71 , 0.82 , 14.28 , 0.70 , 15.52 , 3.15, 1.76, 1.00, ...				
Service times	2.90 , 1.76 , 3.39 , 4.52 , 4.46 , 4.36 , 2.07, 3.36, 2.37, 5.38, ...				

Simulation by Hand: Finishing Up

- **Average waiting time in queue:**

$$\frac{\text{Total of times in queue}}{\text{No. of times in queue}} = \frac{15.17}{6} = 2.53 \text{ minutes per part}$$

- **Time-average number in queue:**

$$\frac{\text{Area under } Q(t) \text{ curve}}{\text{Final clock value}} = \frac{15.78}{20} = 0.79 \text{ part}$$

- **Utilization of drill press:**

$$\frac{\text{Area under } B(t) \text{ curve}}{\text{Final clock value}} = \frac{18.34}{20} = 0.92 \text{ (dimensionless)}$$

Complete Record of the Hand Simulation

Just-Finished Event			Variables		Attributes		Statistical Accumulators									Event Calendar		
Entity No.	Time t	Event Type	$Q(t)$	$B(t)$	Arrival Times: (In Queue) In Service		P	N	ΣWQ	WQ^*	ΣTS	TS^*	$\int Q$	Q^*	$\int B$	[Entity No., Time, Type]		
-	0.00	Init	0	0	() -		0	0	0.00	0.00	0.00	0.00	0.00	0	0.00	[1, 0.00, Arr] [-, 20.00, End]		
1	0.00	Arr	0	1	() 0.00		0	1	0.00	0.00	0.00	0.00	0.00	0	0.00	[2, 1.73, Arr] [1, 2.90, Dep] [-, 20.00, End]		
2	1.73	Arr	1	1	(1.73) 0.00		0	1	0.00	0.00	0.00	0.00	0.00	1	1.73	[1, 2.90, Dep] [3, 3.08, Arr] [-, 20.00, End]		
1	2.90	Dep	0	1	() 1.73		1	2	1.17	1.17	2.90	2.90	1.17	1	2.90	[3, 3.08, Arr] [2, 4.66, Dep] [-, 20.00, End]		
3	3.08	Arr	1	1	(3.08) 1.73		1	2	1.17	1.17	2.90	2.90	1.17	1	3.08	[4, 3.79, Arr] [2, 4.66, Dep] [-, 20.00, End]		
4	3.79	Arr	2	1	(3.79, 3.08) 1.73		1	2	1.17	1.17	2.90	2.90	1.88	2	3.79	[5, 4.41, Arr] [2, 4.66, Dep] [-, 20.00, End]		
5	4.41	Arr	3	1	(4.41, 3.79, 3.08) 1.73		1	2	1.17	1.17	2.90	2.90	3.12	3	4.41	[2, 4.66, Dep] [6, 18.69, Arr] [-, 20.00, End]		
2	4.66	Dep	2	1	(4.41, 3.79) 3.08		2	3	2.75	1.58	5.83	2.93	3.87	3	4.66	[3, 8.05, Dep] [6, 18.69, Arr] [-, 20.00, End]		
3	8.05	Dep	1	1	(4.41) 3.79		3	4	7.01	4.26	10.80	4.97	10.65	3	8.05	[4, 12.57, Dep] [6, 18.69, Arr] [-, 20.00, End]		
4	12.57	Dep	0	1	() 4.41		4	5	15.17	8.16	19.58	8.78	15.17	3	12.57	[5, 17.03, Dep] [6, 18.69, Arr] [-, 20.00, End]		
5	17.03	Dep	0	0	() -		5	5	15.17	8.16	32.20	12.62	15.17	3	17.03	[6, 18.69, Arr] [-, 20.00, End]		
6	18.69	Arr	0	1	() 18.69		5	6	15.17	8.16	32.20	12.62	15.17	3	17.03	[7, 19.39, Arr] [-, 20.00, End] [6, 23.05, Dep]		
7	19.39	Arr	1	1	(19.39) 18.69		5	6	15.17	8.16	32.20	12.62	15.17	3	17.73	[-, 20.00, End] [6, 23.05, Dep] [8, 34.91, Arr]		
-	20.00	End	1	1	(19.39) 18.69		5	6	15.17	8.16	32.20	12.62	15.78	3	18.34	[6, 23.05, Dep] [8, 34.91, Arr]		

Event-Scheduling Logic via Programming

- **Clearly well suited to standard programming language (C, C++, Java, etc.)**
- **Often use “utility” libraries for:**
 - List processing
 - Random-number generation
 - Random-variate generation
 - Statistics collection
 - Event-list and clock management
 - Summary and output
- **Main program ties it together, executes events in order**

Simulation Dynamics:

Process-Interaction World View

- Identify characteristic *entities* in system
- Multiple copies of entities co-exist, interact, compete
- “Code” is non-procedural
- Tell a “story” about what happens to a “typical” entity
- May have many types of entities, “fake” entities for things like machine breakdowns
- Usually requires special simulation software
 - Underneath, still executed as event-scheduling
- View normally taken by Arena
 - Arena translates your model description into a program in SIMAN simulation language for execution

Randomness in Simulation

- Above was just one “replication” — a sample of size one (not worth much)
- Made a total of five *replications* (IID):

Performance Measure	Replication					Sample		95%
	1	2	3	4	5	Avg.	Std. Dev.	Half Width
Total production	5	3	6	2	3	3.80	1.64	2.04
Average waiting time in queue	2.53	1.19	1.03	1.62	0.00	1.27	0.92	1.14
Maximum waiting time in queue	8.16	3.56	2.97	3.24	0.00	3.59*	2.93*	3.63*
Average total time in system	6.44	5.10	4.16	6.71	4.26	5.33	1.19	1.48
Maximum total time in system	12.62	6.63	6.27	7.71	4.96	7.64*	2.95*	3.67*
Time-average number of parts in queue	0.79	0.18	0.36	0.16	0.05	0.31	0.29	0.36
Maximum number of parts in queue	3	1	2	1	1	1.60*	0.89*	1.11*
Drill-press utilization	0.92	0.59	0.90	0.51	0.70	0.72	0.18	0.23

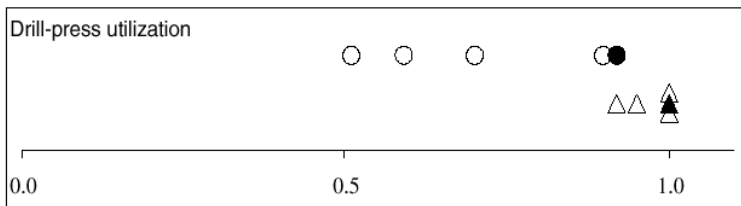
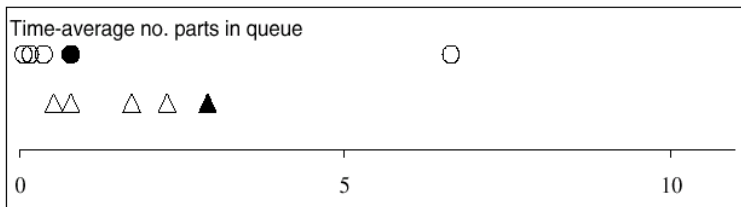
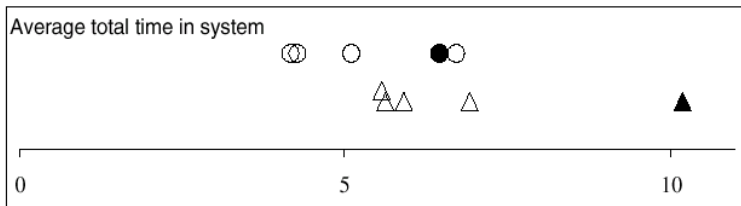
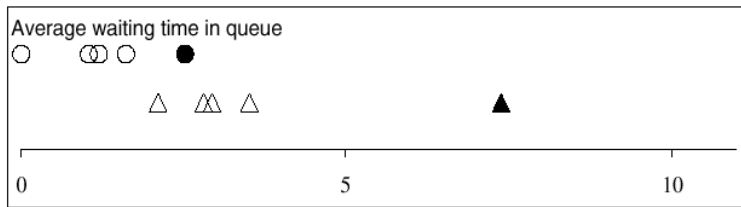
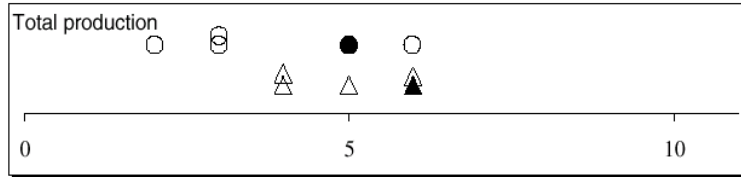
Substantial variability across replications

- Confidence intervals for expected values:
 - In general, $\bar{X} \pm t_{n-1, 1-\alpha/2} s / \sqrt{n}$ (normality assumption?)
 - For expected total production, $3.80 \pm (2.776)(1.64 / \sqrt{5})$
 3.80 ± 2.04 *Precision?*

Comparing Alternatives

- Usually, simulation is used for more than just a single model “configuration”
- Often want to compare alternatives, select or search for best (via some criterion)
- Simple processing system: What would happen if arrival rate doubled?
 - Cut interarrival times in half
 - Rerun model for double-time arrivals
 - Make five replications

Results: Original vs. Double-Time Arrivals



- Original – circles
- Double-time – triangles
- Replication 1 – filled in
- Replications 2-5 – hollow
- Note variability
- Danger of making decisions based on one (first) replication
- Hard to see if there are really differences
- Need: Statistical analysis of simulation output data

Simulating with Spreadsheets: Introduction

- **Popular, ubiquitous tool**
- **Can use for simple simulation models**
 - Typically, only static models
 - Risk analysis, financial/investment scenarios
 - Only (very) simplest of dynamic models
- **Two examples**
 - Newsvendor problem (static)
 - Waiting times in single-server queue (dynamic)
 - Special recursion valid only in this case

Simulating with Spreadsheets:

Newsvendor Problem – Setup

- **Rupert sells daily newspapers on street**
 - Rupert buys for $c = \$0.55$ each, sells for $r = \$1.00$ each
- **Each morning, Rupert buys q copies**
 - q is a fixed number, same every day
- **Demand during a day: $D = \max(\lfloor X \rfloor, 0)$**
 - $X \sim \text{normal}(\mu = 135.7, \sigma = 27.1)$, from historical data
 - $\lfloor X \rfloor$ rounds X to nearest integer
- **If $D \leq q$, satisfy all demand, and $q - D \geq 0$ left over, sell for scrap at $s = \$0.03$ each**
- **If $D > q$, sells out (sells all q copies), no scrap**
 - But missed out on $D - q > 0$ sales
- **What should q be?**

Simulating with Spreadsheets:

News vendor Problem – Formulation

- **Choose q to maximize expected profit per day**
 - q too small – sell out, miss \$0.45 profit per paper
 - q too big – have left over, scrap at a loss of \$0.52 per paper
- **Classic operations-research problem**
 - Many versions, variants, extensions, applications
 - Much research on exact solution in certain cases
 - But easy to simulate, even in a spreadsheet
- **Profit in a day, as a function of q :**
$$W(q) = \underbrace{r \min(D, q)}_{\text{Sales revenue}} + \underbrace{s \max(q - D, 0)}_{\text{Scrap revenue}} - \underbrace{cq}_{\text{Cost}}$$
 - $W(q)$ is a random variable – profit varies from day to day
- **Maximize $E(W(q))$ over nonnegative integers q**

Simulating with Spreadsheets:

Newsvendor Problem – Simulation

- **Set trial value of q , generate demand D , compute profit for that day**
 - Then repeat this for many days independently, average to estimate $E(W(q))$
 - Also get confidence interval, estimate of $P(\text{loss})$, histogram of $W(q)$
 - Try for a range of values of q
- **Need to generate demand $D = \max(\lfloor X \rfloor, 0)$**
 - So need to generate $X \sim \text{normal}(\mu = 135.7, \sigma = 27.1)$
 - (Much) ahead – Sec. 12.2, generating random *variates*
 - In this case, generate $X = \Phi_{\mu, \sigma}^{-1}(U)$
 - U is a random number distributed uniformly on $[0, 1]$ (Sec. 12.1)
 - $\Phi_{\mu, \sigma}$ is cumulative distribution function of normal (μ, σ) distribution

Simulating with Spreadsheets:

News vendor Problem – Excel

- File Newsvendor.xls
- Input parameters in cells B4 – B8 (blue)
- Trial values for q in row 2 (pink)
- Day number (1, 2, ..., 30) in column D
- Demands in column E for each day:

All files in book: www.mhhe.com/kelton,
Student Edition, BookExamples.zip

$$= \text{MAX}(\text{ROUND}(\text{NORMINV}(\text{RAND}(), \$B\$7, \$B\$8), 0), 0)$$

Rounding
function

Φ^{-1}

$U(0, 1)$
random number

μ

σ

$X \sim \text{normal}(\mu, \sigma)$

RAND() is “volatile”
so regenerates on
any edit, or F9 key

\$ pins down following
column or row when
copying formula

Round to
nearest
integer

MAX 2nd
argument

Simulating with Spreadsheets:

Newsvendor Problem – Excel (cont'd.)

- **For each q :**
 - “Sold” column: number of papers sold that day
 - “Scrap” column: number of papers scrapped that day
 - “Profit” column: profit (+, –, 0) that day
 - Placement of “\$” in formulas to facilitate copying
- **At bottom of “Profit” columns (green):**
 - Average profit over 30 days
 - Half-width of 95% confidence interval on $E(W(q))$
 - Value 2.045 is upper 0.975 critical point of t distribution with 29 d.f.
 - Plot confidence intervals as “I-beams” on left edge
 - Estimate of $P(W(q) < 0)$
 - Uses **COUNTIF** function
- **Histograms of $W(q)$ at bottom**
 - Vertical red line at 0, separates profits, losses

Simulating with Spreadsheets:

Newsvendor Problem – Results

- **Fine point – used same daily demands (column E) for each day, across all trial values of q**
 - Would have been valid to generate them independently
 - Why is it better to use same demands for all q ?
- **Results**
 - Best q is about 140, maybe a little less
 - Randomness in all results (tap F9 key)
 - All demands, profits, graphics change
 - Confidence-interval, histogram plots change
 - Reminder that these are random outputs, random plots
 - Higher $q \Rightarrow$ more variability in profit
 - Histograms at bottom are wider for larger q
 - Higher chance of both large profits, but higher chance of loss, too
 - Risk/return tradeoff can be quantified – risk taker vs. risk-averse

Simulating with Spreadsheets: Single-Server Queue – Setup

- **Like hand simulation, but:**
 - Interarrival times \sim exponential with mean $1/\lambda = 1.6$ min.
 - Service times \sim uniform on $[a, b] = [0.27, 2.29]$ min.
 - Stop when 50th waiting time in queue is observed
 - i.e., when 50th customer *begins* service, not exits system
- **Watch waiting times in queue $WQ_1, WQ_2, \dots, WQ_{50}$**
 - Important – not watching anything else, unlike before
- **S_i = service time of customer i ,**
 A_i = interarrival time between custs. $i - 1$ and i
- ***Lindley's recursion* (1952):** Initialize $WQ_1 = 0$,
 $WQ_i = \max (WQ_{i-1} + S_{i-1} - A_i, 0), i = 2, 3, \dots$

Simulating with Spreadsheets:

Single-Server Queue – Simulation

- **Need to generate random variates:** let $U \sim U[0, 1]$
 - Exponential (mean $1/\lambda$): $A_i = -(1/\lambda) \ln(1 - U)$
 - Uniform on $[a, b]$: $S_i = a + (b - a) U$
- **File MU1.xls**

*All files in book: www.mhhe.com/kelton,
Student Edition, BookExamples.zip*
- **Input parameters in cells B4 – B6 (blue)**
 - Some theoretical outputs in cells B8 – B10
- **Customer number ($i = 1, 2, \dots, 50$) in column D**
- **Five IID replications (three columns for each)**
 - IA = interarrival times, S = service times
 - WQ = waiting times in queue (plot, thin curves)
 - First one initialized to 0, remainder use Lindley's recursion
Curves rise from 0, variation increases toward right
 - Creates positive autocorrelation down WQ columns
Curves have less abrupt jumps than if WQ_i 's were independent

Simulating with Spreadsheets: Single-Server Queue – Results

- **Column averages (green)**
 - Average interarrival, service times close to expectations
 - Average WQ_i within each replication
 - Not too far from steady-state expectation
 - Considerable variation
 - Many are below it (why?)
- **Cross-replication (by customer) averages (green)**
 - Column T, thick line in plot to dampen noise
- **Why no sample variance, histograms of WQ_i 's?**
 - Could have computed both, as in newsvendor; two issues:
 - Nonstationarity – what is a “typical” WQ_i here?
 - Autocorrelation – biases variance estimate, may bias histogram if run is not “long enough”

Simulating with Spreadsheets:

Recap

- **Popular for static models**
 - Add-ins – @RISK, Crystal Ball
- **Inadequate tool for dynamic simulations if there's any complexity**
 - Extremely easy to simulate single-server queue in Arena – Chapter 3 main example
 - Can build very complex dynamic models with Arena – most of rest of book

Overview of a Simulation Study

- Understand system
- Be clear about goals
- Formulate model representation
- Translate into modeling software
- Verify “program”
- Validate model
- Design experiments
- Make runs
- Analyze, get insight, document results

More: Chapter 13