

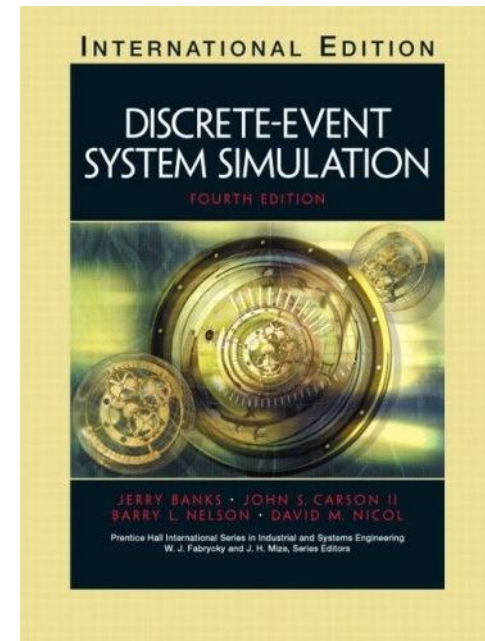
# Introduction to Simulation

## Output Analysis

# Background Reading

## Relevant sections of the book:

- 11.1
- 11.2
- 11.3
- 11.4 (parts)
- 11.5 (parts)



# A Question

Let  $X$  be the random variable

- "Result from throwing a die"



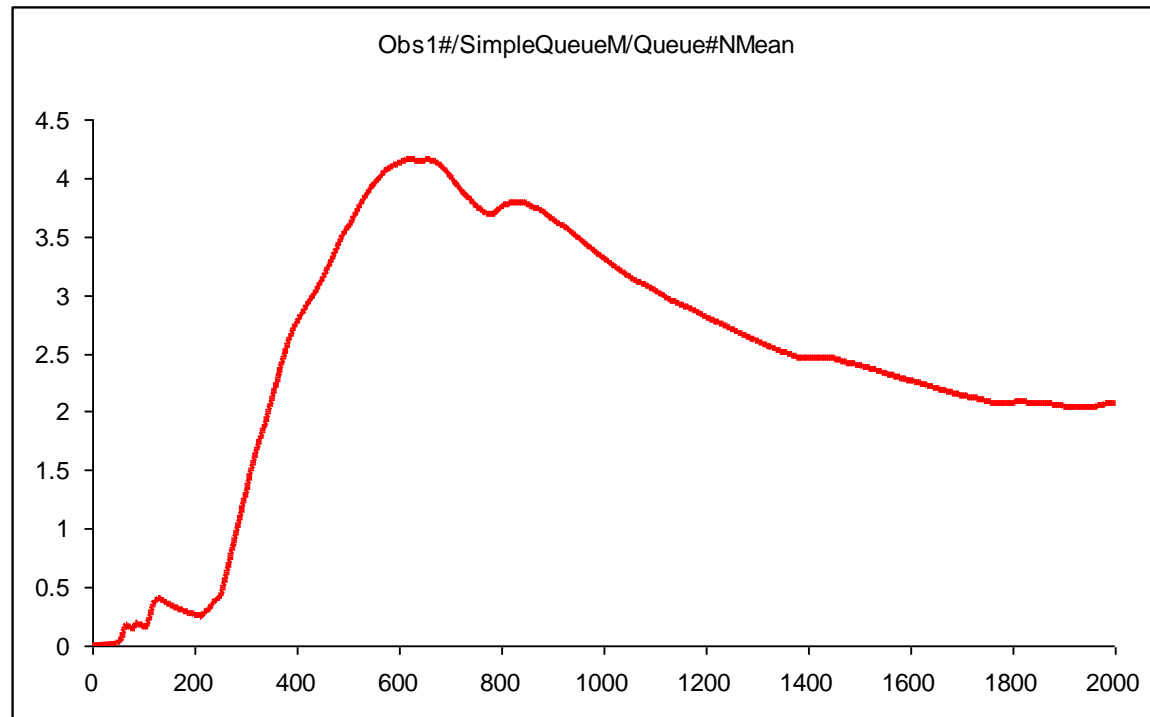
Question: What is the expected value  $E(X)$ ?

Would you throw the die just once and take the result as your answer?

# What Are We Doing Wrong?

We have been leaving something out in our simulations

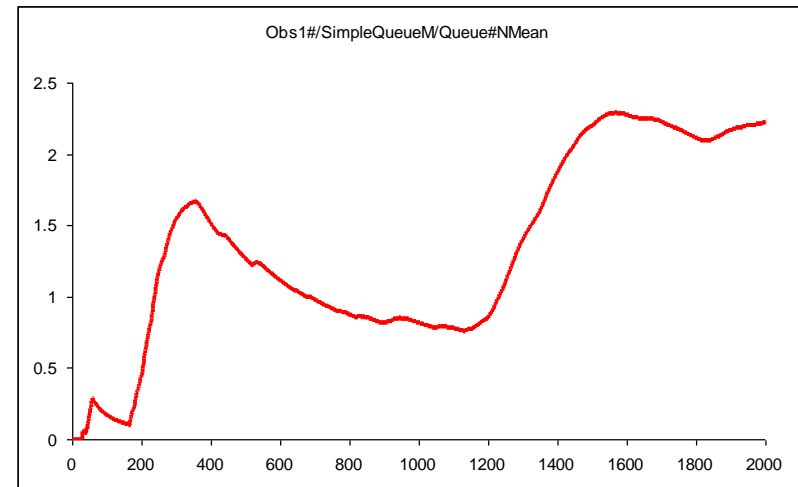
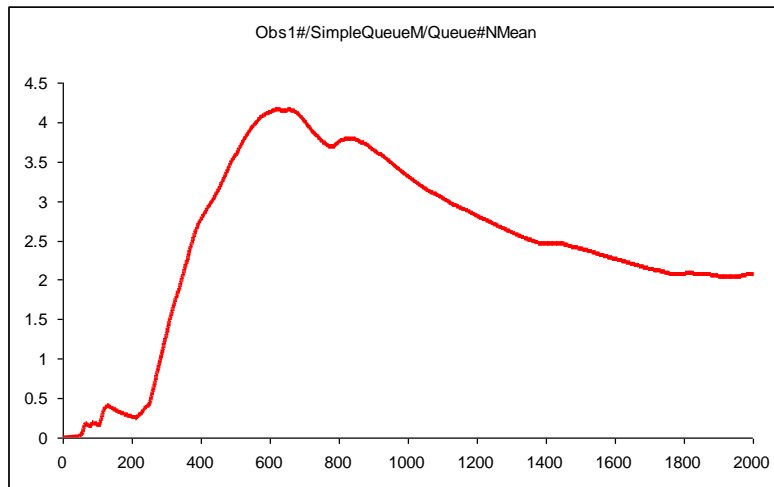
- We have made one observation only
- Example – the average length of the queue in the bank:



# Example

## Consider our bank example

- Observe the average queue length  $0 < T < 2000$
- Simulate with different sets of random numbers (with the same distribution!)



# Example

## Consider our bank example

- Observe the queue length at  $T = 2000$
- Repeat the simulation using different sets of random numbers

## Results:

- 2, 12, 3, 7, 0, 10, 2, 5

Which is the "right" answer?

# What Are We Doing Wrong?

## Real systems behave randomly

- They contain random variables

## Our simulation results are also random

- They depend on random numbers

## Running a simulation means taking one sample of a RV

- $\Rightarrow$  We need a more sophisticated approach!

# Mean & Sample Variance

Consider a random variable  $Y$

Take a set of observations  $Y_i \ i=1 \dots n$

The *Sample Mean* is defined as:  $\bar{Y} = \frac{1}{n} \sum Y_i$

The *Sample Variance* is defined as:  $S^2 = \frac{1}{n-1} \sum (Y_i - \bar{Y})^2$

Then  $S^2 \approx \text{var}(Y)$

(The sample variance approximates the actual variance in  $Y$ )



# Bias

Given a value  $\theta$  and an estimator for it  $\hat{\theta}$

In general, we may have  $E(\hat{\theta}) = \theta + b$   
(i.e. the estimator may be biased)

Bias means we have a systematic error

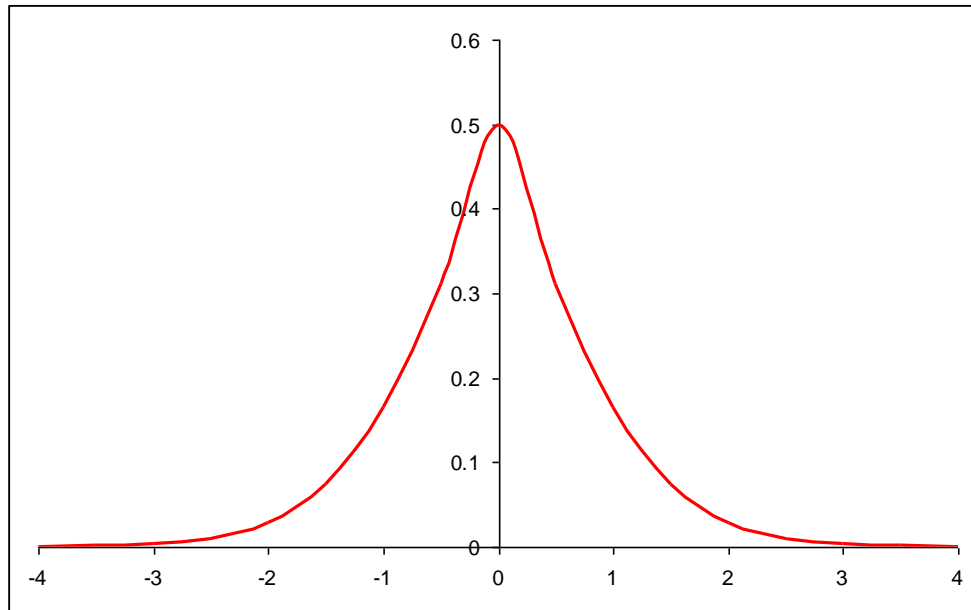
- Our estimate will be too large / too small

Many statistical methods assume (require) that  $b=0$

# Student's $t$ -Distribution

Introduced by W. Gosset (a.k.a. "Student")

- Has one parameter  $f$  ("degrees of freedom")
- Used for hypothesis testing
- Tables of values are available



W. Gosset  
1876-1937

# Student's $t$ -Distribution

## Given:

- A measure of the real system  $\theta$
- An estimator  $\hat{\theta}$  for  $\theta$
- An estimator  $\hat{\sigma}(\hat{\theta})$  for  $\sigma(\theta)$

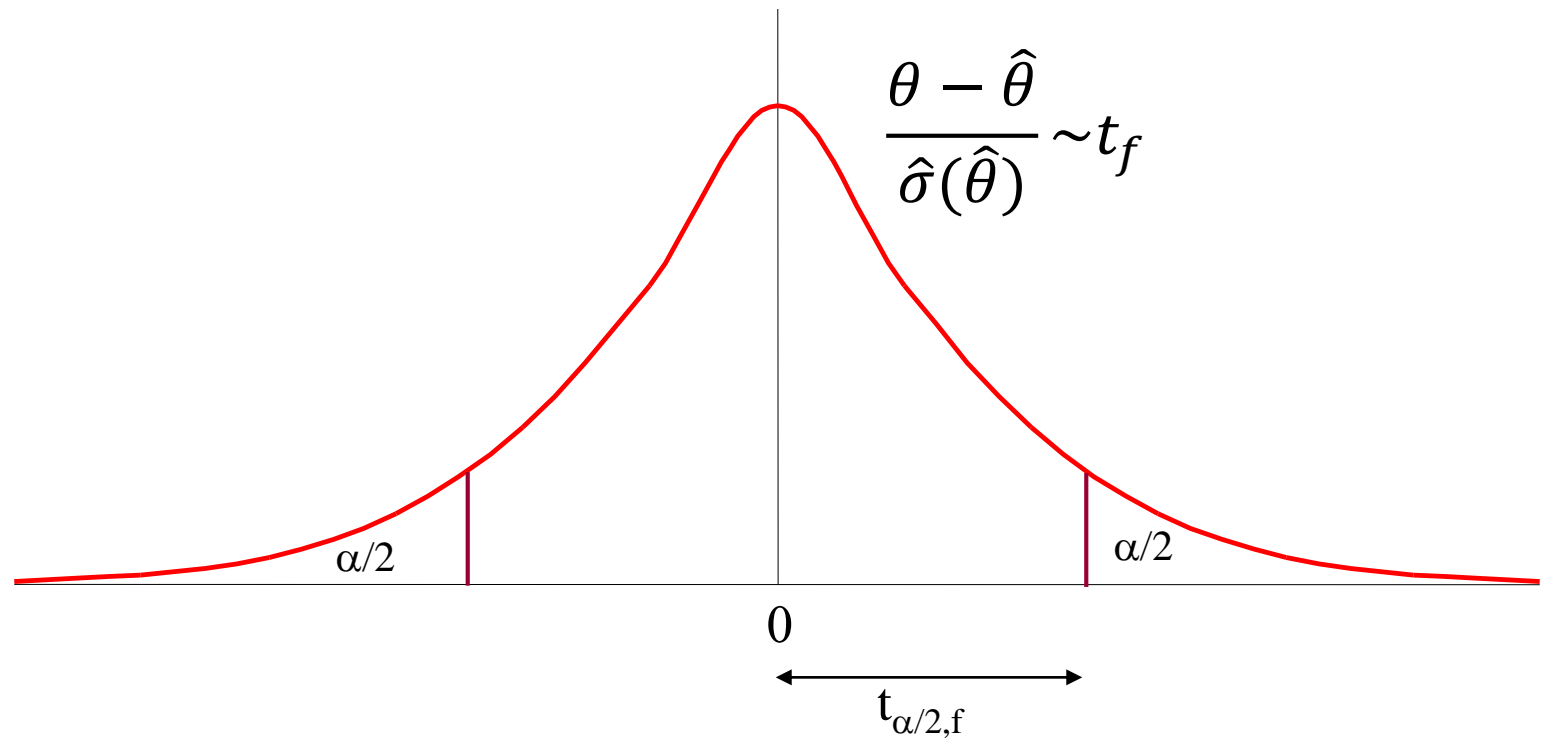
If  $\theta = E(\hat{\theta})$  then the value

$$t = \frac{\theta - \hat{\theta}}{\hat{\sigma}(\hat{\theta})}$$

is  $t$ -distributed with  $n - 1$  d.o.f.

# Confidence Intervals

What are we doing?



# Confidence Intervals

Choose a level of significance  $\alpha$

Rearrange the expression

$$0 - t_{\alpha/2,f} \leq \frac{\theta - \hat{\theta}}{\hat{\sigma}(\hat{\theta})} \leq 0 + t_{\alpha/2,f}$$

to obtain the confidence interval

$$\hat{\theta} - \hat{\sigma}(\hat{\theta}) \cdot t_{\alpha/2,f} \leq \theta \leq \hat{\theta} + \hat{\sigma}(\hat{\theta}) \cdot t_{\alpha/2,f}$$

# Confidence Intervals

What does the confidence interval mean?

$$\hat{\theta} - \hat{\sigma}(\hat{\theta}) \cdot t_{\alpha/2, f} \leq \theta \leq \hat{\theta} + \hat{\sigma}(\hat{\theta}) \cdot t_{\alpha/2, f}$$

The value of  $\theta$  lies in the c.i. with certainty  $1 - \alpha$

This is the preferred way to present the results of a simulation

Reminder:

- $\theta$  is the (theoretical) output of the simulation model
- $\hat{\theta}$  is the result of a (finite) simulation experiment

# Confidence Intervals

How to obtain  $\hat{\sigma}(\hat{\theta})$  ?

Answer: Use the approximation  $\frac{s}{\sqrt{n}} \approx \hat{\sigma}(\hat{\theta})$

# Confidence Intervals

Improving accuracy:

We have  $\frac{s}{\sqrt{n}} \approx \hat{\sigma}(\hat{\theta})$

If we wish to halve the width of the the c.i. ...

- ... we must use  $4n$  samples!

This can imply:

- High accuracy can be expensive in simulation!

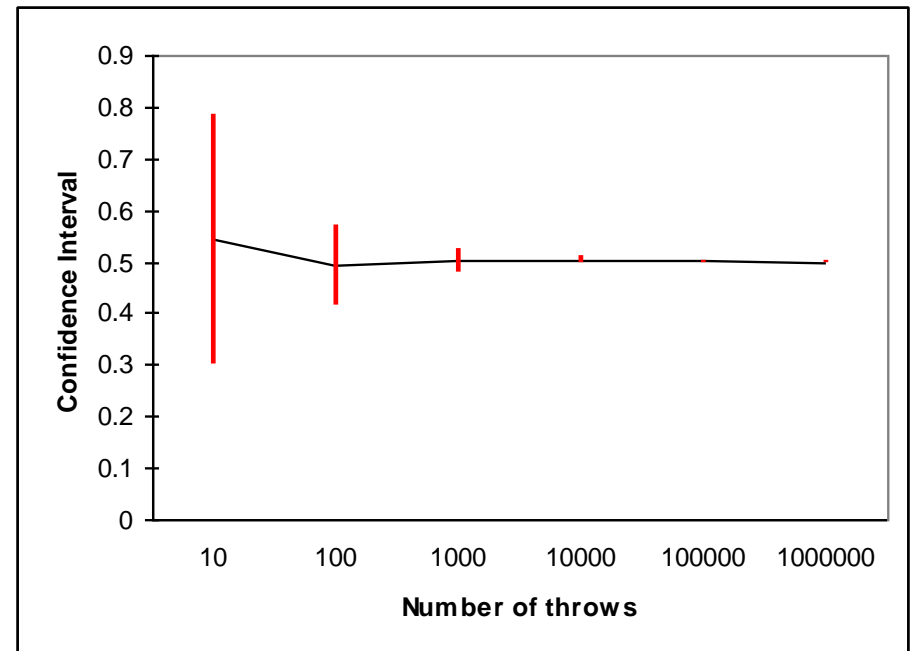


# Confidence Intervals

## Example: Throwing a coin (interpret the result as 0 or 1)

- $\theta$  is the (true, theoretical) expected value
- $\hat{\theta}$  is the result of a (finite) experiment

# Throws	Lower	Mean	Upper
10	0.302	0.545	0.788
100	0.416	0.493	0.570
1000	0.479	0.503	0.527
10000	0.496	0.503	0.511
100000	0.499	0.501	0.503
1000000	0.499	0.500	0.501



# Confidence Intervals

A confidence interval is only accurate if:

- $\hat{\theta}$  is an unbiased estimator of  $\theta$
- $\hat{\sigma}^2(\hat{\theta})$  is an unbiased estimator of  $\sigma^2(\hat{\theta})$

If the observations are not independent...

- ... then the estimator will be biased
- ... the confidence interval will be shifted

# Example

## Consider a simple queue

- One arrival stream
- One server

## Intervals:

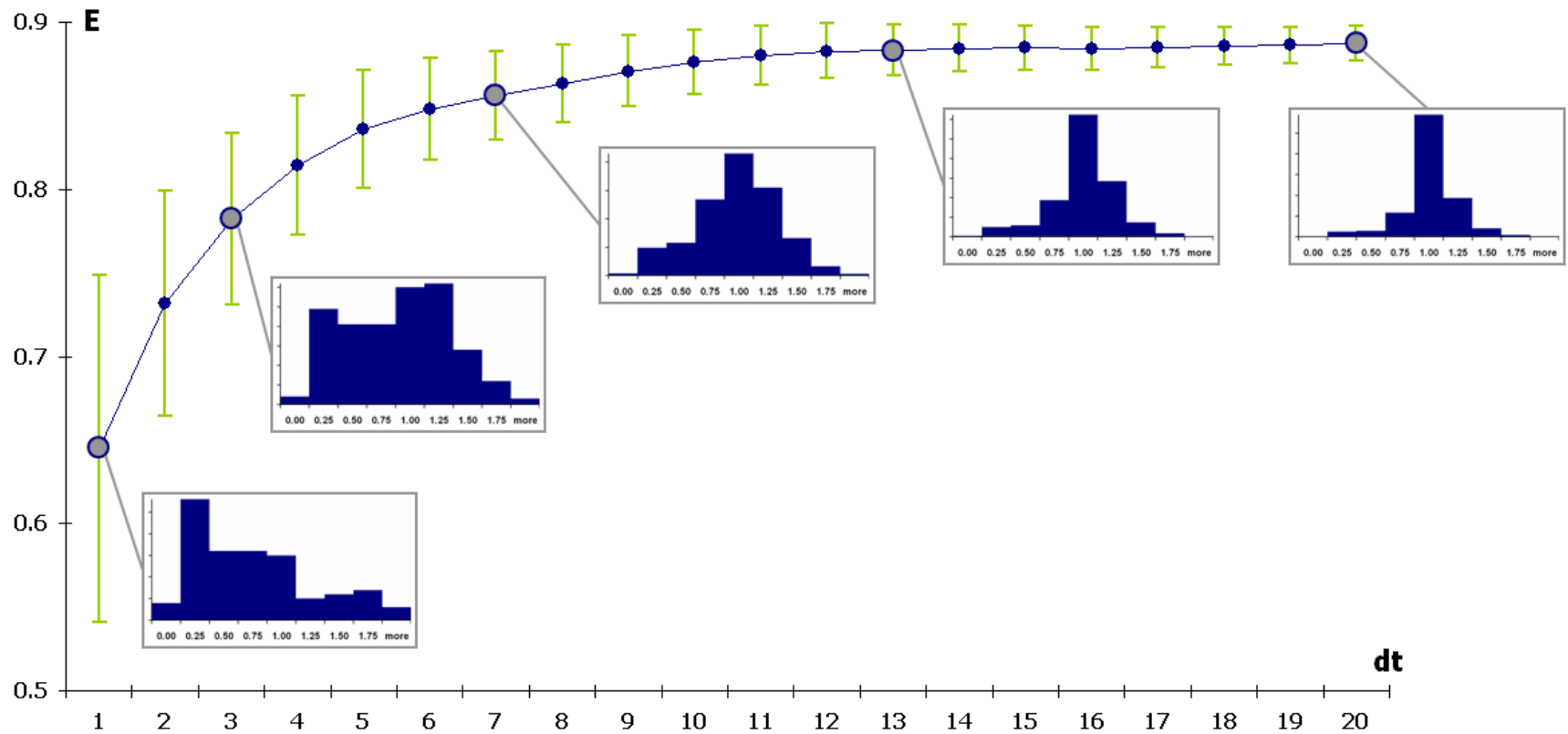
- Arrivals: Exponential distribution, mean = 13s
- Service: Normal distribution,  $\mu = 10s$ ,  $\sigma = 2s$

## How does the queue length behave over time?

## Simulation experiment:

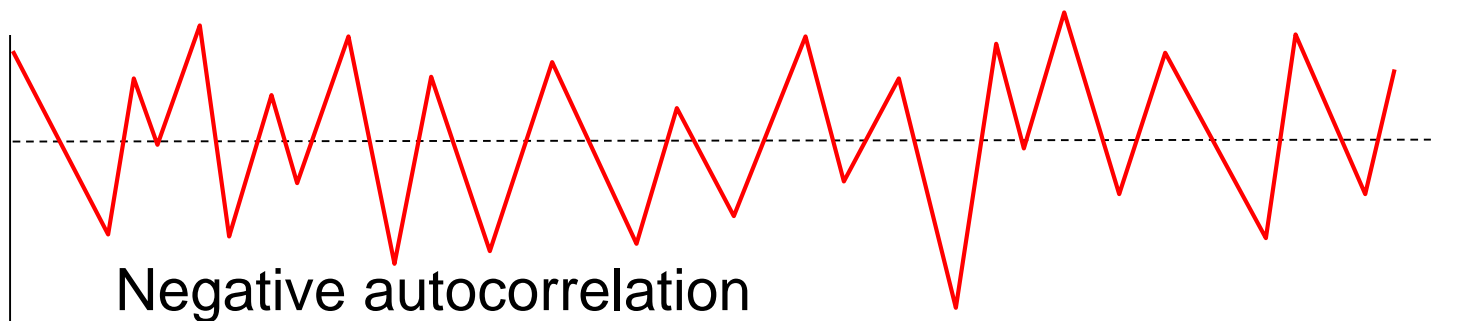
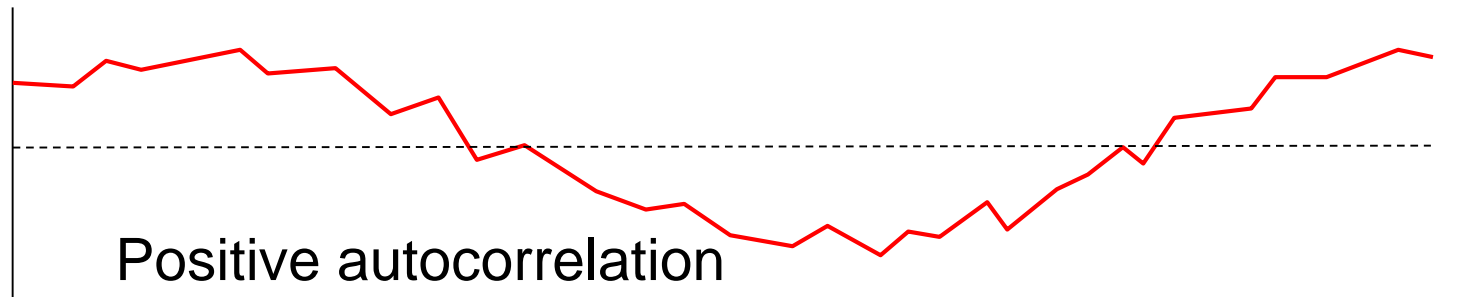
- 100 samples (replications),  $\alpha = 0.05$

# Example



# Bias due to Autocorrelation

Consider the  $Y_i$  from one simulation run (time series):



# Bias due to Autocorrelation

## Positive autocorrelation ...

- leads to under-estimation of  $S^2$
- leads to over-optimistic confidence intervals
- Example: Most queues

## Negative autocorrelation ...

- leads to over-estimation of  $S^2$
- leads to under-optimistic confidence intervals
- Example: Some inventory systems

⇒ Don't use time series for confidence intervals!

# Independent Replications

The method of independent replications:

- Run the simulation  $R$  times
- Use independent sets of random numbers for each run
- Make the observations  $Y_r$ ,  $r=1 \dots R$
- Compute  $\hat{\theta}$  and  $S^2$  from the  $Y_r$
- Compute a confidence interval from  $\hat{\theta}$  and  $S^2$

Independence ensures that  $\hat{\theta}$  is unbiased

# Terminating Simulations

A terminating simulation is one...

- which runs up to a specified time
- which has known initial conditions
- in which the initial conditions are important

Examples:

- Will the satellite survive for 5 years?
- How full is the bank 2 hours after opening time?



# Non-Terminating Simulations

A non-terminating simulation is one...

- which runs for an indefinite period
- in which the steady-state behaviour is of interest
- in which the initial conditions are not important

Examples:

- Any continuously running system
- Computer centre, traffic system, ...

# Non-Terminating Simulations

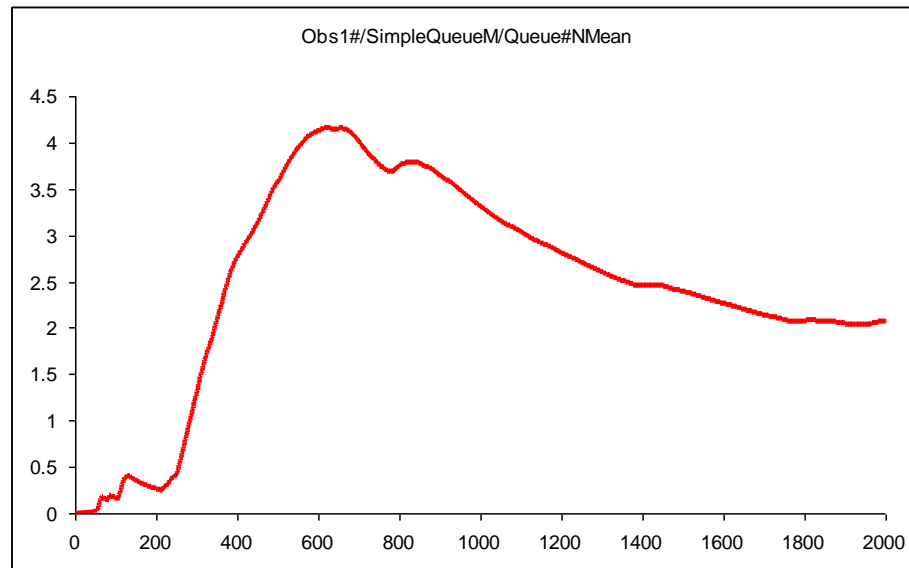
## Difficulties with non-terminating simulations:

- Initial bias
- How long to run the simulation?
- Trade off between replication and duration

# Initial Bias

## In a non-terminating simulation

- We are interested in the steady-state behaviour
- The values at the beginning will usually be untypical ("initial bias")
- Example: Queue in bank (starting empty):

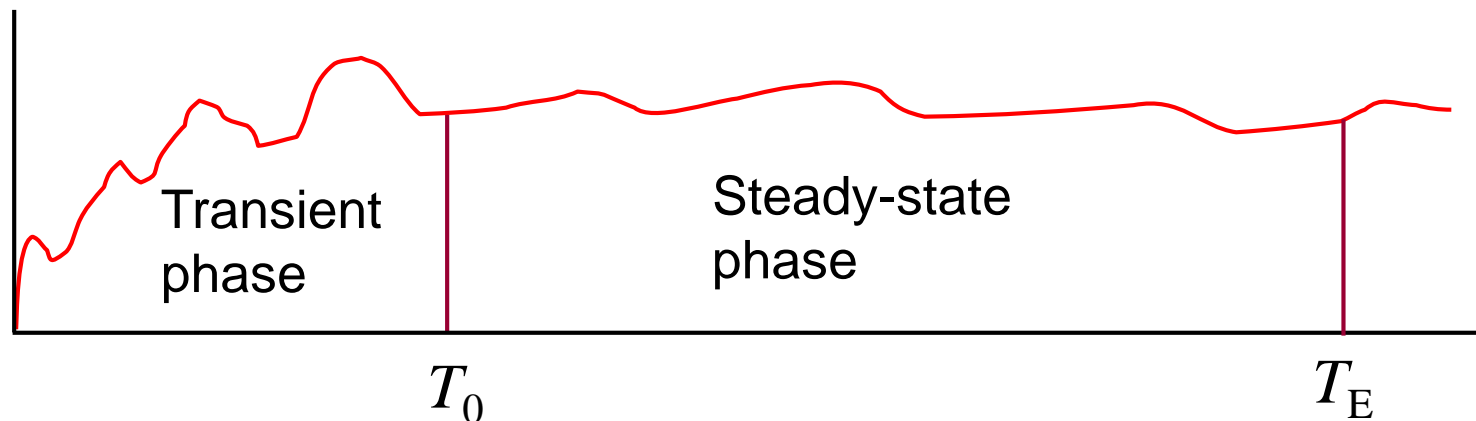


# Initial Bias

## Solution:

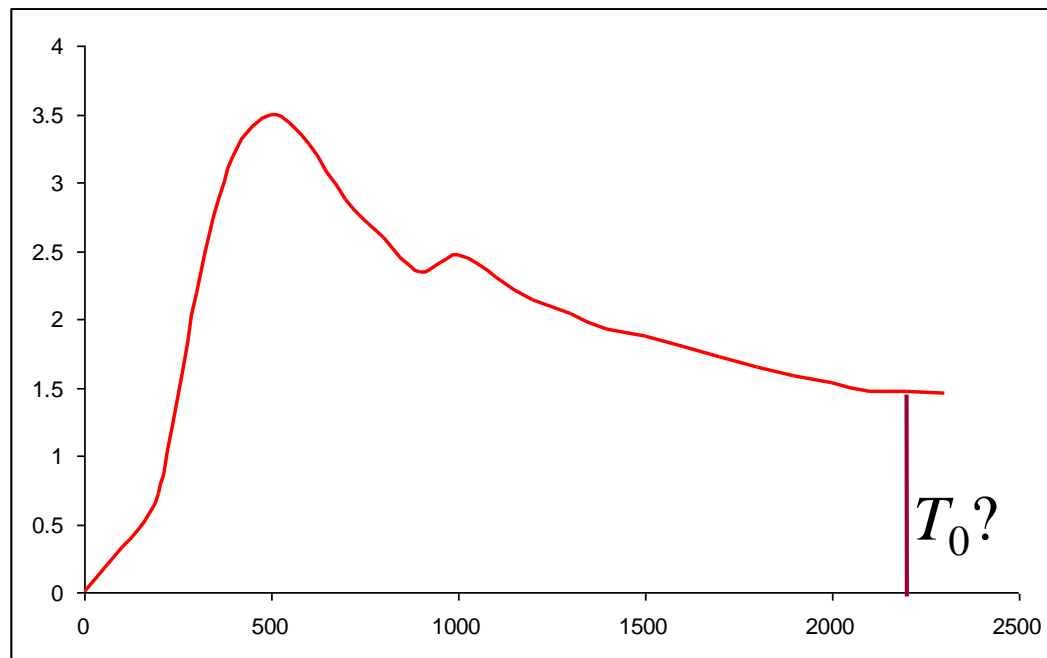
- Delete values from transient phase
- Choose  $T_0$  after transient phase is over
- Observe from  $T_0$  to  $T_E$

## How to choose $T_0$ (and $T_E$ )?



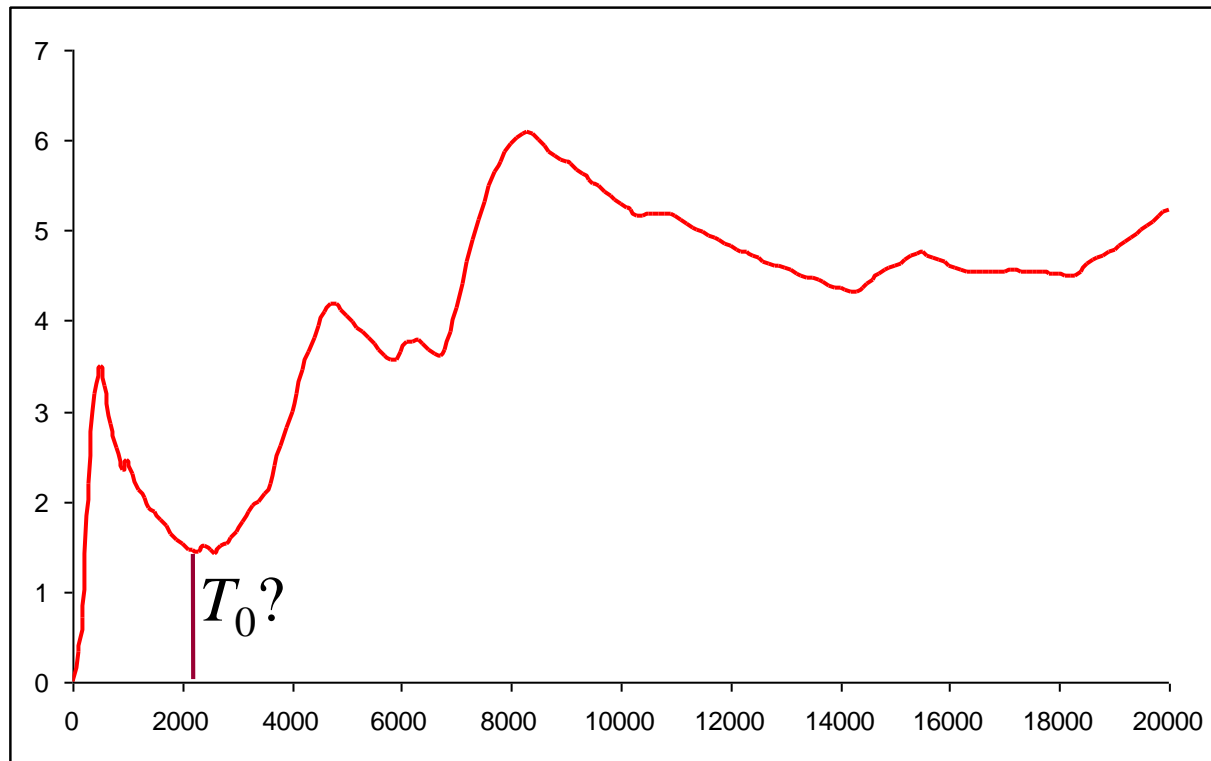
# Initial Bias

How long is the transient phase in a non-terminating simulation?



# Initial Bias

A difficult question!



# Ensemble Averages

Using just one run to find  $T_0$  is dangerous

- Compute ensemble averages

Perform independent replications to obtain

$$Y_{r,i} \quad i = 1 \dots n, \quad r = 1 \dots R$$

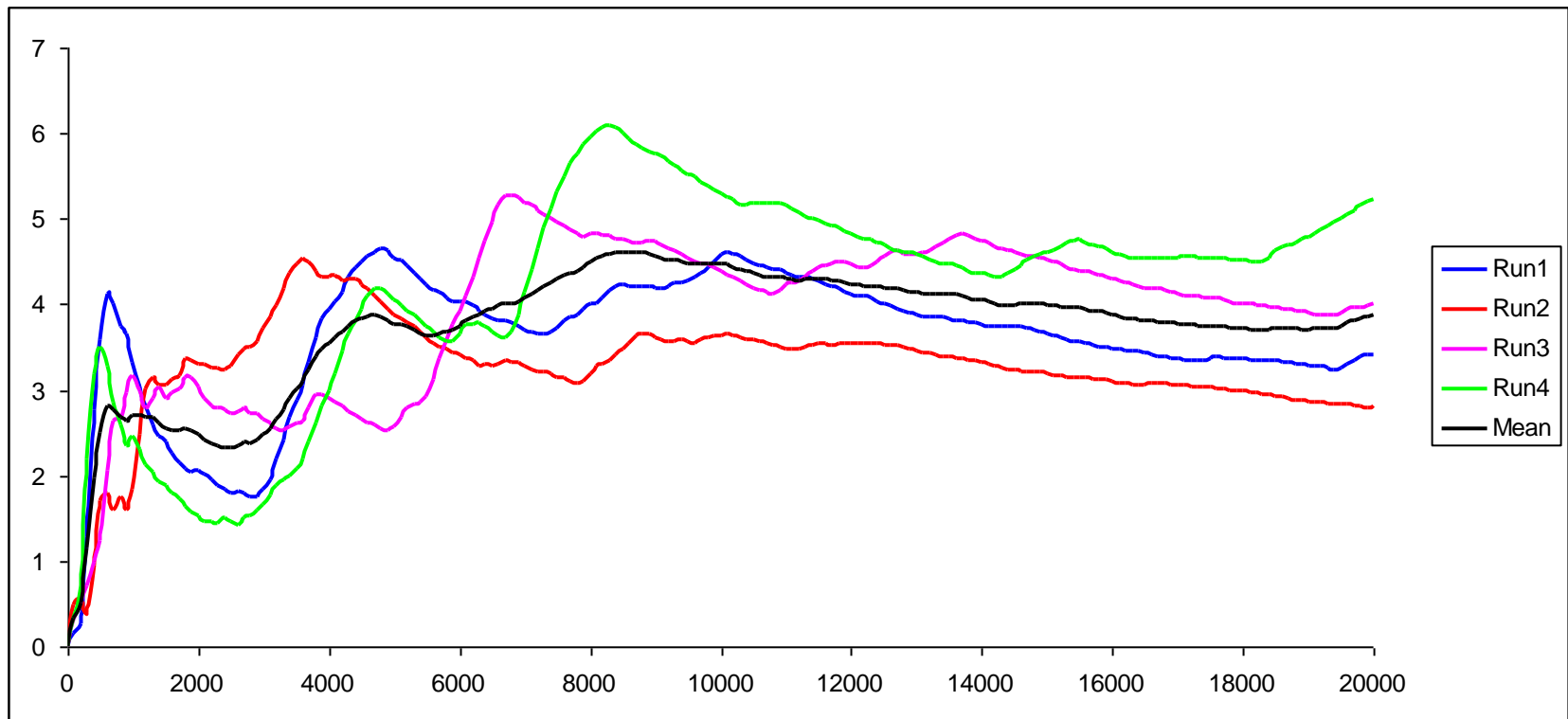
Compute average values across replications

$$Y_i = \frac{1}{R} \sum_{r=1}^R Y_{r,i} \quad i = 1 \dots n$$

Test the sequence  $Y_i$  for the end of the transient phase

# Ensemble Averages

Compute ensemble averages:





# Non-Terminating Simulations

Increasing  $R$  will make the c.i. narrower

It will not reduce the initial bias, i.e.

- ...we will get a better c.i. around  $\theta + b$ !

A computing time tradeoff is necessary between

- ...increasing  $T_E$  and increasing  $R$

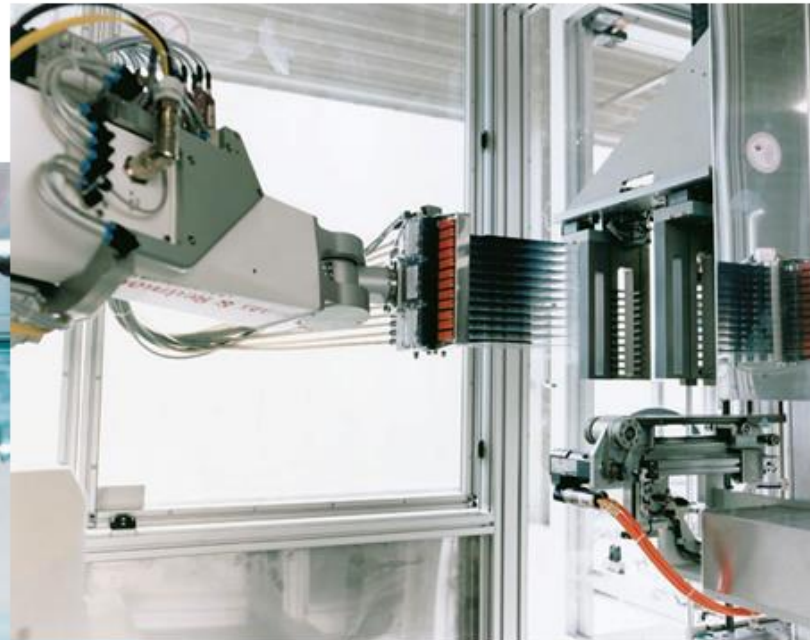
Solution:

- $R > 25$  is not useful

# Simulation Project

## Utilization of a machine ("Nassbank") at Q-Cells:

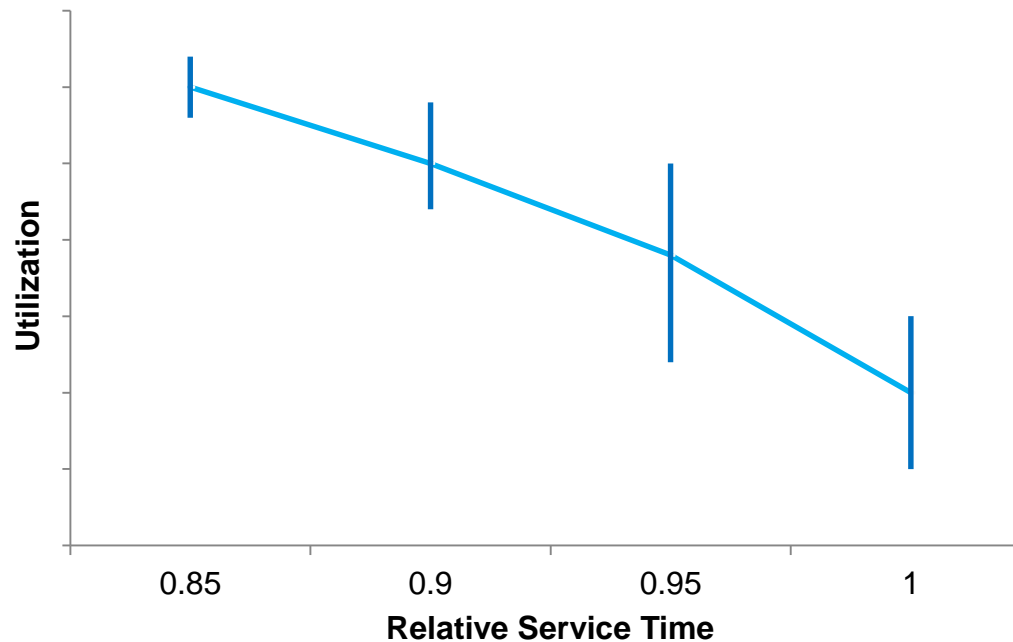
- Test dependency of utilization on service time reduction



# Simulation Project

## Simulation experiment:

- 25 replications
- 99% confidence interval



# Simulation Project

## Intersection of Gustav-Adolf-Straße and Bundesstraße 1

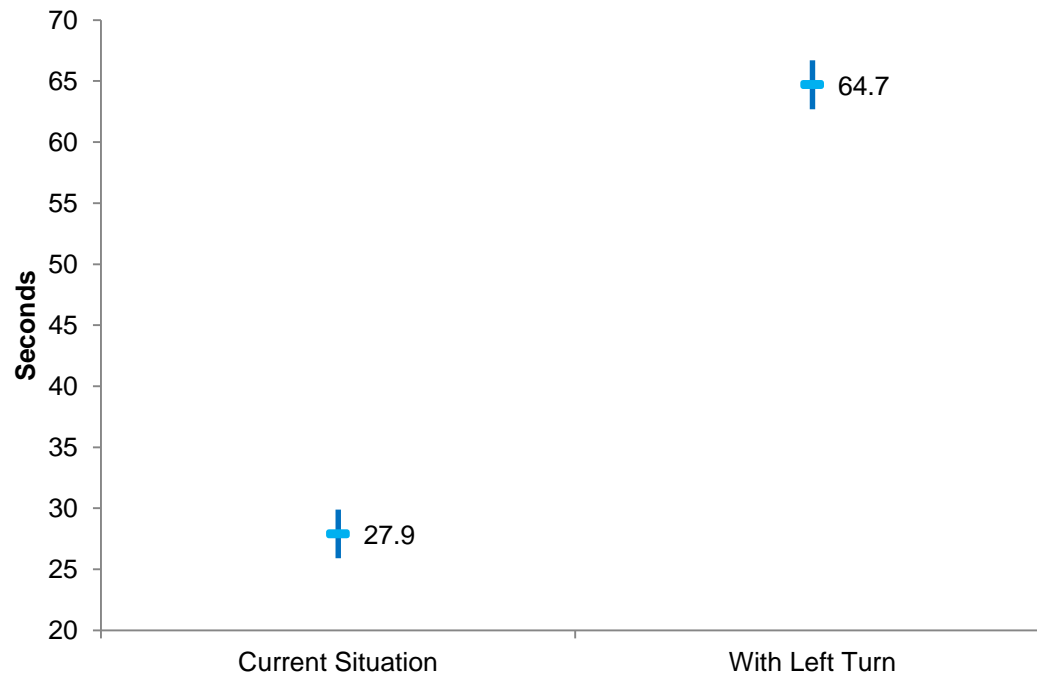
- Analyse the effect of allowing left turns to the north



# Simulation Project

## Simulation experiment for waiting time for westbound traffic:

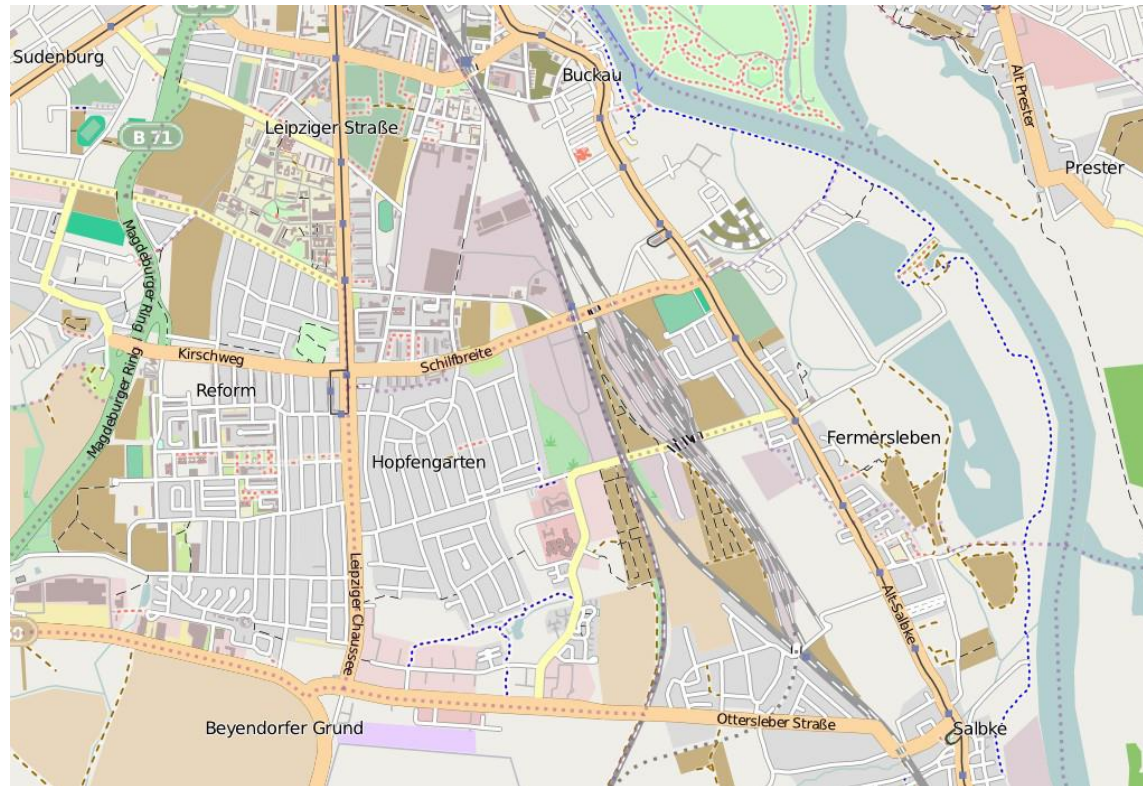
- 50 replications
- 90% confidence interval



# Simulation Project

## Magdeburg Buckau

- Analyze the effect of a planned street to disburden an existing one of through traffic when developing new industrial areas

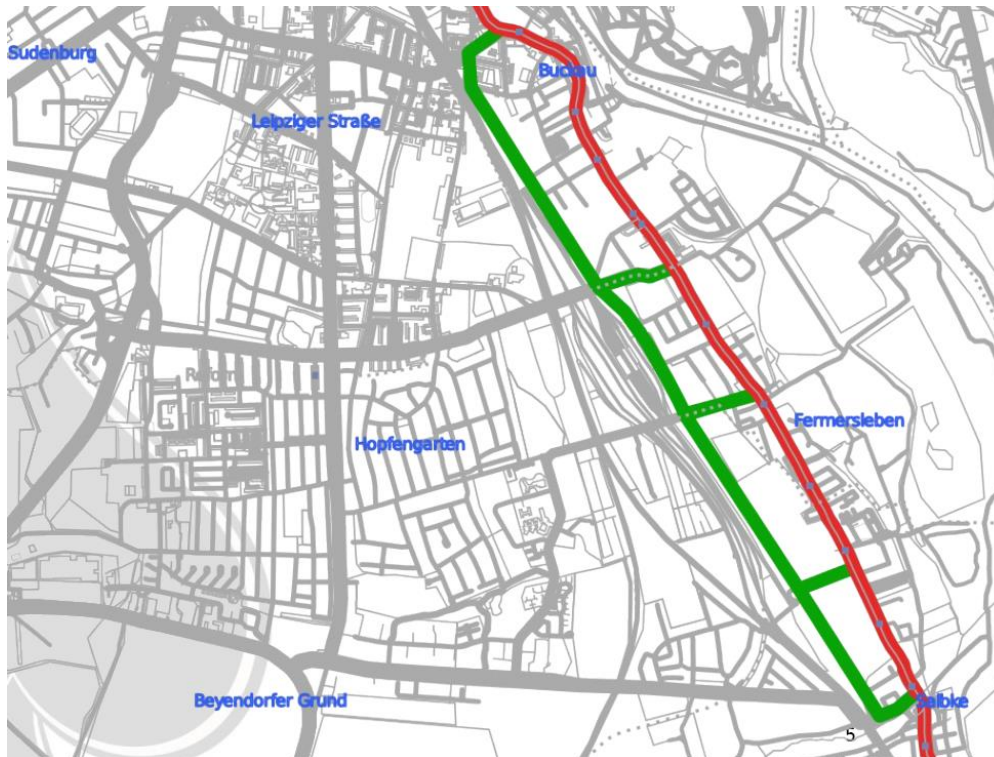




# Simulation Project

## Magdeburg Buckau

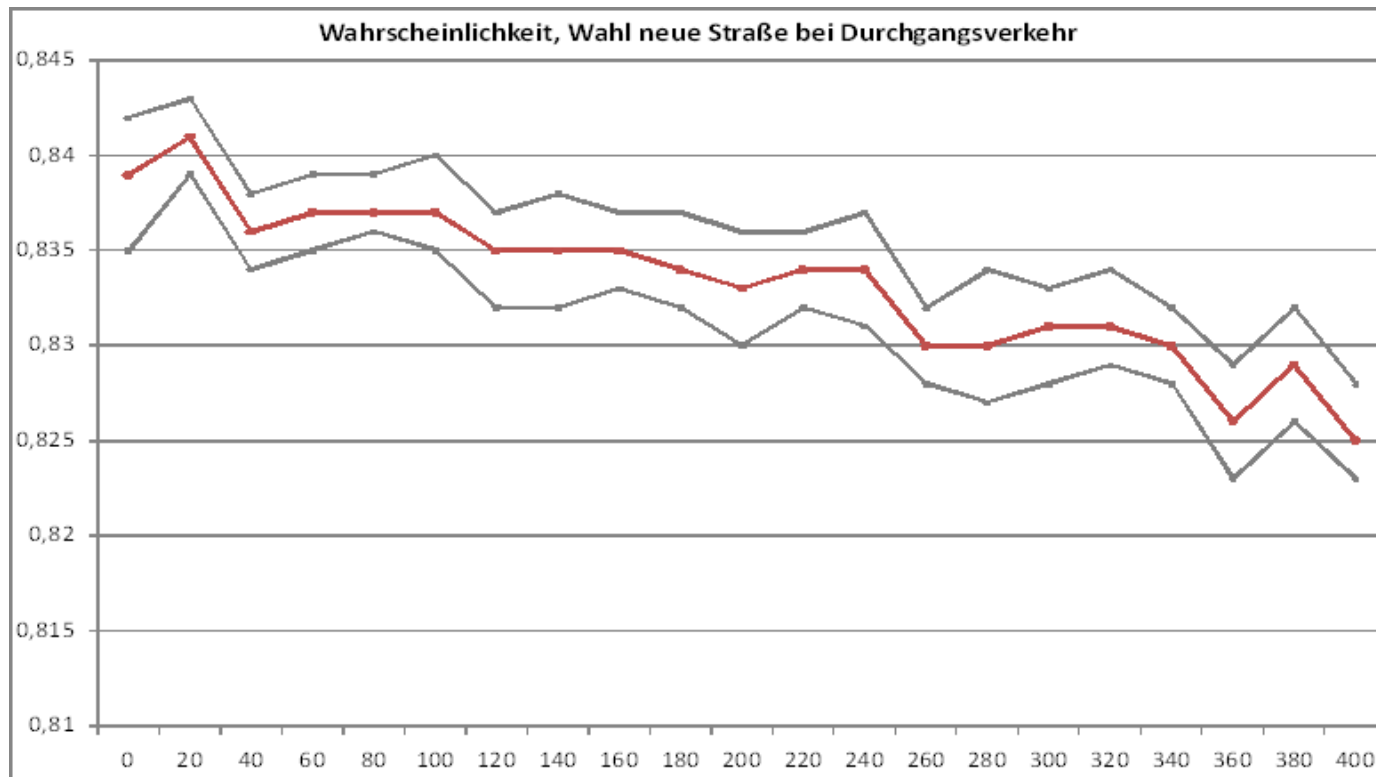
- Analyze the effect of a planned street to disburden an existing one of through traffic when developing new industrial areas



# Simulation Project

Probability of choosing the new street when varying the industry density adjacent to the new street:

- 10 replications, 90% confidence interval



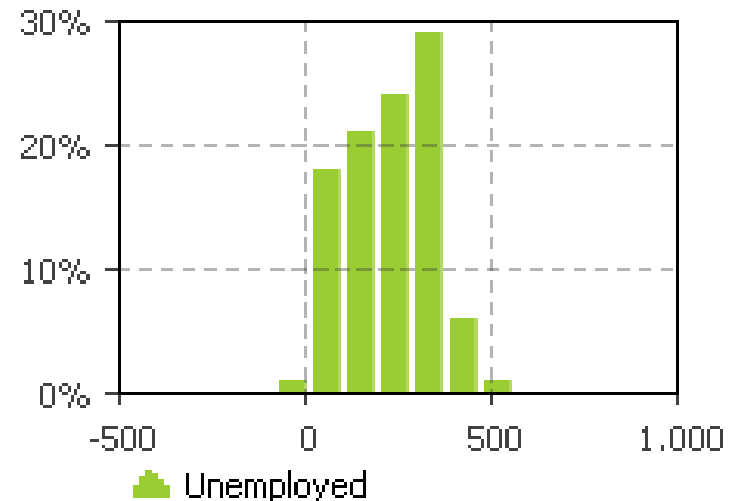
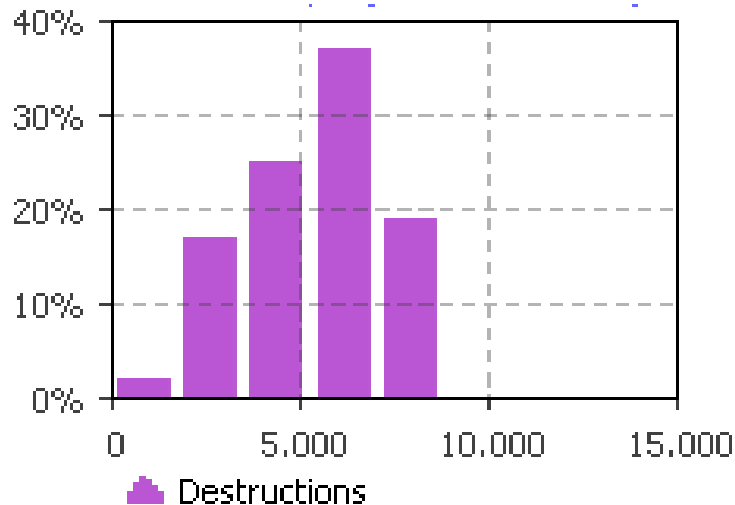


# The Sims – Almost Normal Family Life



You have to compute confidence intervals:

- For how long will the father be unemployed on average?
- How much money will be spent on damaged school property?

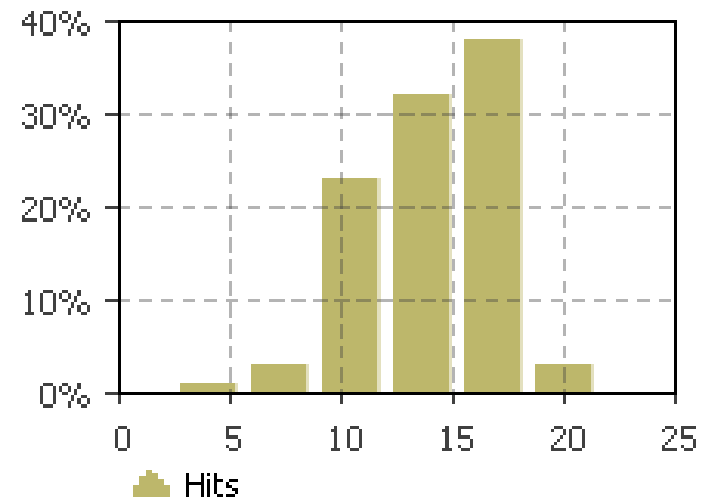
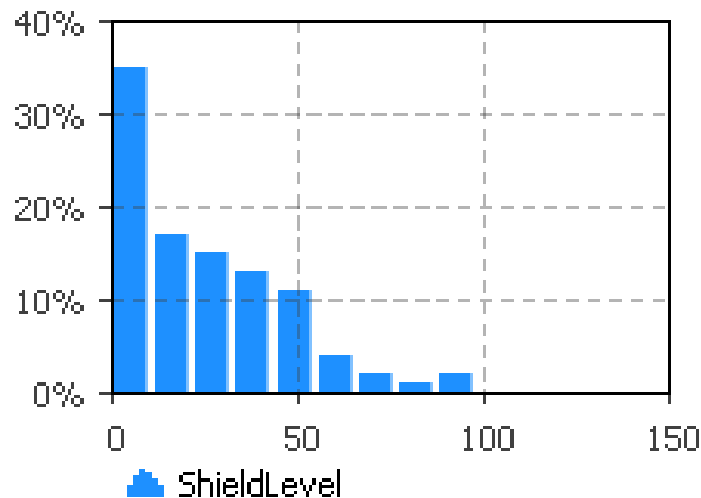


# Star Trek – USS Enterprise in Danger



You have to compute confidence intervals:

- What will the shield energy level be after 2 hours?
- How many antimatter particles will hit the shield?



# Learning Goals

## Learning questions:

- How is the sample variance of a set of random samples defined?
- What is the method of independent replications?
- How is a confidence interval computed?
- What does a confidence interval signify?
- How would you reduce the width of a confidence interval?
- What are terminating and non-terminating simulations?
- What is initial bias? How can it be avoided?
- What is an ensemble average?