

Mechanics of Simulation

Richard de Neufville
Professor of Engineering Systems
MIT
Institute of Data, Systems, and Society

Organization

- **Presentation is**
- **Based on text**
- **Plus some Sensitivity Analysis**

What is Simulation?

- Replicates outcomes of uncertain process (often called “Monte Carlo” simulation)
- It provides a way to describe what may occur over range of possible scenarios
- Can use variety of distributions:
 - _ regular or
 - _ irregular
 - continuous or not

Use of Simulation: not widespread New in General Practice

- Modern software makes simulation practical
 - 1000' s of repetitions in seconds
- Model often simple, e.g., spreadsheet of profits
 - Simple example: Excel Add-in
 - Market versions: Crystal Ball®; @Risk®
 - ...
 - See Antamina case (other slide deck)

Requirements for Simulation

- **Distributions for Key parameters**
 - **May be observed, estimated, assumed, or guessed**
- **Examples:**
 - **Observed: Rainfall, river flows over years**
 - **Estimated: Technical Cost Models**
 - **Assumed: Market data (historical prices)**
 - **Guessed: Judgment (ore quantity, quality)**

Recommended Process

Steps:

1. **Produce a standard valuation model**
2. **Do Sensitivity Analysis**
 - **One variable at a time**
 - **Probabilistically**
3. **Model System Performance**
 - **No Flexibility**
 - **With alternative flexibilities**

Step 1: Produce Valuation Model

Parking Garage Case

- **Garage in area where population expands**
- **Inspired by Bluewater Mall in SE England**

- **Actual demand is necessarily uncertain**

- **Question (for this step): How do we value?**

Parking Garage Case details

- **Demand**
 - **At start is for 750 spaces**
 - **Over next 10 years is expected to rise exponentially by another 750 spaces**
 - **After year 10 may be 250 more spaces**
 - **could be +/- 50% off the projections**
 - **Annual volatility for growth is 10%**
- **Average annual revenue/space = \$10,000**
- **The discount rate is taken to be 12%**

Parking Garage details (Cont)

- **Costs**
 - annual operating costs (staff, cleaning, etc.) = \$2,000 /year/space available
 - spaces used often < spaces available
 - Annual lease of the land = \$3.6 Million
 - construction cost = \$16,000/space + 10% for each level above the ground level

- **Site can accommodate 200 cars per level**

Valuation Model

SYSTEM PARAMETERS					
Capacity per level	200	cars			
Number of levels	6	levels	[DESIGN PARAMETERS]		
PERFORMANCE CALCULATION					
Year	0	1	2	...	15
Demand		750	893	...	1,634
Capacity		1,200	1,200	...	1,200
Revenue		7.5	8.9	...	12.0
Operating costs	0.0	3.6	3.6	...	3.6
Land leasing and fixed costs	3.3	3.3	3.3	...	3.3
Cashflow	-3.3	0.6	2.0	...	5.1
Discounted cashflow	-3.3	0.5	1.7	...	1.2
Present value of cashflow	26.7				
Capacity cost for up to two levels	6.8				
Capacity costs for levels above 2	17.4				
Net present value	2.5				

Figure D.1
Flexibility in
Engineering
Design
de Neufville
and Scholtes
MIT Press
2011

Step 2: Sensitivity Analysis

- One dimensional

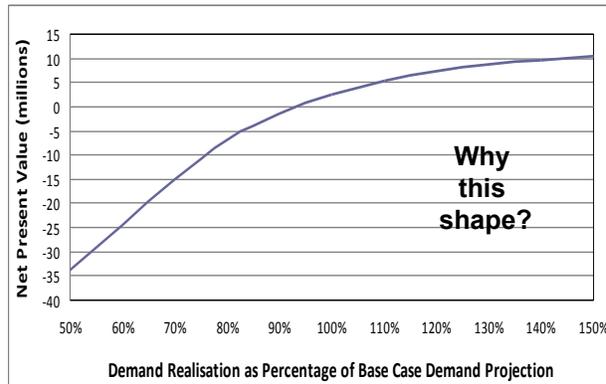


Figure D.2
Flexibility in
Engineering
Design
de Neufville
and Scholtes
MIT Press
2011

Intro to Simulation
Massachusetts Institute of Technology

Richard de Neufville ©
Slide 11 of 25

Tornado Diagram

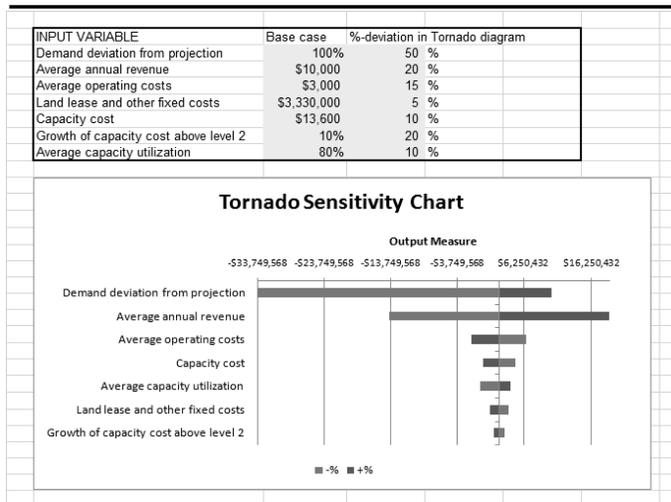


Figure D.3
Flexibility in
Engineering
Design
de Neufville
and Scholtes
MIT Press
2011

Intro to Simulation
Massachusetts Institute of Technology

Richard de Neufville ©
Slide 12 of 25

Interpreting Sensitivity Analysis

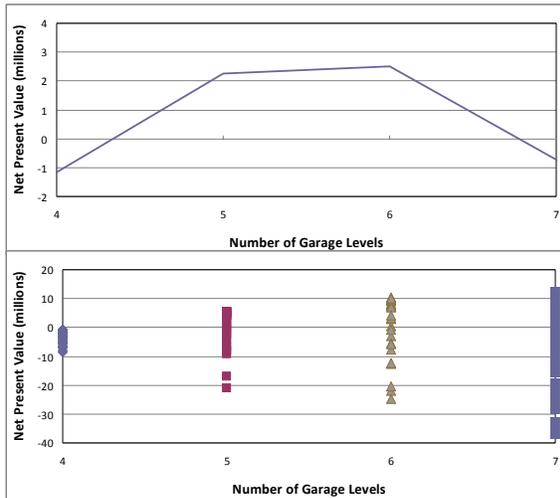
- Everything is uncertain – what should we pay attention to?
- Sensitivity analysis helps to answer this question
- In this case, the level of demand seems to affect performance (here, the value of investment) most.
- Such observations drive rest of analysis

Probabilistic Sensitivity Analysis

Trial	Demand deviation	Average annual revenue	Average operating costs	Land lease and other fixed costs	Capacity cost	Growth of cap cost above 2 levels	Average capacity utilization	NPV
1	134%	\$11,016	\$3,140	\$3,495,435	\$12,399	9.88%	79.69%	\$17,625,663
2	141%	\$11,852	\$3,260	\$3,304,375	\$13,795	9.84%	79.06%	\$23,221,598
3	86%	\$10,193	\$3,312	\$3,430,938	\$13,068	10.43%	79.50%	-\$4,718,694
4	83%	\$9,055	\$3,283	\$3,354,325	\$14,036	9.90%	76.09%	-\$15,456,972
5	83%	\$10,114	\$3,047	\$3,316,798	\$12,685	11.52%	87.40%	-\$3,934,426
6	62%	\$8,111	\$3,283	\$3,206,261	\$13,100	11.02%	84.27%	-\$36,456,858
7	76%	\$10,515	\$2,647	\$3,386,862	\$14,260	11.81%	83.77%	-\$6,799,900
8	68%	\$8,343	\$3,440	\$3,269,874	\$14,830	9.71%	78.82%	-\$32,839,236
9	114%	\$10,272	\$3,291	\$3,410,315	\$14,852	10.11%	85.44%	\$5,120,553
10	68%	\$10,151	\$3,413	\$3,437,877	\$14,110	10.78%	85.28%	-\$24,320,813
11	68%	\$9,664	\$3,153	\$3,163,520	\$13,841	11.51%	82.62%	-\$21,314,749
12	129%	\$11,189	\$2,636	\$3,430,040	\$14,713	10.59%	75.51%	\$16,118,317
13	132%	\$11,832	\$2,695	\$3,179,077	\$14,469	11.09%	84.10%	\$30,373,102
14	114%	\$10,250	\$2,586	\$3,189,684	\$14,311	9.34%	75.61%	\$9,916,978
15	105%	\$9,754	\$2,641	\$3,218,399	\$14,147	9.14%	72.32%	\$2,261,573
16	63%	\$9,595	\$3,089	\$3,240,407	\$13,222	9.92%	77.66%	-\$22,983,479
17	147%	\$10,881	\$2,702	\$3,374,710	\$13,140	10.25%	79.15%	\$20,546,268
18	91%	\$8,783	\$3,275	\$3,222,284	\$12,731	9.97%	86.09%	-\$10,350,864
19	67%	\$9,320	\$3,182	\$3,477,948	\$14,446	9.57%	81.29%	-\$26,886,841
20	67%	\$8,961	\$3,302	\$3,206,613	\$14,770	9.87%	75.37%	-\$26,433,140

Figure D.4 – variables moving together, not one at a time

Choice of best design in Context of Uncertainty



Figures D.8 and D.9
Flexibility in Engineering Design
Top is Expected NPV,
Bottom Graph shows range of outcomes.

Notice the difference in scale between the two graphs!
In this case, the uncertainties may dominate the choice.

Intro to Simulation
Massachusetts Institute of Technology

Richard de Neufville ©
Slide 15 of 25

Step 3: Introduce dynamic scenarios

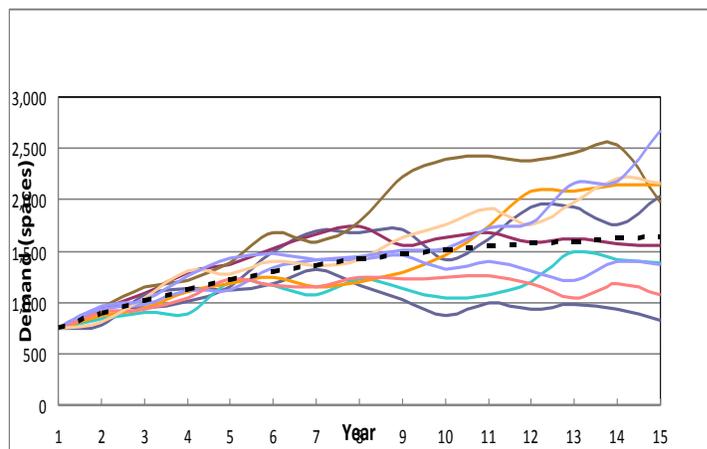


Figure D.6
Flexibility in Engineering Design
de Neufville and Scholtes
MIT Press
2011

Intro to Simulation
Massachusetts Institute of Technology

Richard de Neufville ©
Slide 16 of 25

Simulation: no flexibility

- Generate a dynamic scenario and apply it to model, period by period
- Calculate the performance in each period and over life of project (such as NPV)
- Report for many scenarios (1000 or more) representing distribution of uncertainty
- Calculate mean (such as ENPV) of results, also distribution

Shape of output distributions

- Traditional (NPV) analysis is:
“numbers in, numbers out”
- Uncertainty analysis is:
“shapes in, shapes out”
- “Target curve” is useful representation of
“shape out.” It is its cumulative distribution.

Target Curve

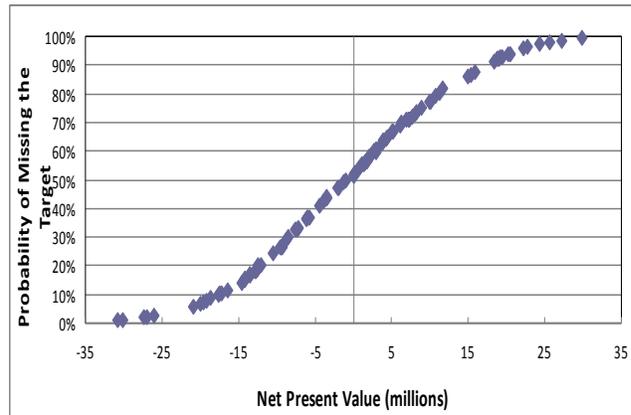


Figure D.14
Flexibility in
Engineering
Design
de Neufville
and Scholtes
MIT Press
2011

Creation of Target Curve

- At end of Simulation, we have many trials
- What is probability of each?
- They are equal !!
- How do we get pdf?
- “Binning” the outcomes by ranges (= bins)
 - Percent samples in bin = $P(\text{outcome in bin})$
- Note: We look at all, 1000s, of simulations!
 - But binning process easily automated

Simulation: with flexibility

- **First question is: what kind of flexibility should we consider?**
- **Answer flows from the uncertainties that impact performance most**
- **In garage case, demand for spaces is most significant, so what kind of flexibility?**
- **For this case, we want flexibility as to size, or capacity of project – the number of levels to garage**

Rules for exercising flexibility (such as option to close mine – Chile)

- **When should we exercise flexibility?**
- **In simulation, this time cannot be calculated**
- **Why?**
- **Because number of possible future paths, states are too large to be searched**
- **Procedure: set up a priori conditions for when to exercise flexibility. These are “decision rules” for exercising flexibility**

Example of “Decision Rules”

Consider a possible rule for Parking Garage

- “Expand if, over 2 years, observed demand is greater than capacity ”
- Why would this make sense?
=> Because, want some assurance that growth is ‘permanent’
- How could this be improved?
=> Change rule toward end of life? No addition in last 5 years?

Simulation WITH flexibility

- As for inflexible case, BUT
- At each period, check if decision rule for that period tells you to exercise flexibility – if yes, do so and take advantage of result
- Note Carefully: **CAN EXERCISE MANY FLEXIBILITIES** in same analysis! This is not possible in conventional financial analyses of options

Value of Flexibility by Simulation

- 1. Get distribution of consequences for plan or design without flexibility
=> NPV pdf, EV(NPV); Target Curve**
- 2. Repeat above, considering availability of flexibility, and exercise at desired times
=> EV(NPV); Target Curve**
- 3. Flexibility Value = difference of EV(NPV)s
Compare Target Curves to see source of value**

MIT OpenCourseWare
<https://ocw.mit.edu/>

IDS.333 Risk and Decision Analysis
Fall 2021

For information about citing these materials or our Terms of Use, visit: <https://ocw.mit.edu/terms>.