## ABM for Management Decision Support [Spring 2013]

Lecture 2

#### Agent-Based Modelling and Simulation Case Studies

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#### Content Lecture 2

- Case Study 1
  - Department Store Customer Service
- Case Study 2
  - Office Energy Consumption
- Case Study 3
  - Forced Innovation Adoption



#### **Recap: Simulation Paradigms**

- System Dynamics (continuous, deterministic)
  - Aggregate view; differential equations
    - Modelling methods: Causal loop diagrams; stock and flow diagrams
- Discrete Event (discrete, stochastic)
  - Process oriented (top down); one thread of control; passive objects
    - Modelling methods: Process flow diagrams; activity cycle diagrams
- Agent Based (discrete, stochastic)
  - Individual centric (bottom up); each agent has its own thread of control; active objects
    - Modelling methods: UML (class diagrams + state chart diagrams) + Equations
- Multi-Method (linked or integrated)





# Case Study 1

#### (For more details see Siebers and Aickelin 2011)

A queuing system





#### Case Study 1: Context

- Case study sector
  - Retail (department store operations)
- Developing some tools for understanding the impact of management practices on company performance
  - Operational management practices are well researched
  - People management practices are often neglected
- Problem:
  - How can we model proactive customer service behaviour?





- Two case studies at two different locations
  - Two departments (A&TV and WW) at two department stores
- Knowledge gathering
  - Informal participant observations
  - Staff interviews
  - Informational sources internal to the case study organisation

















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- Software: AnyLogic v5 (later translated into v6)
  - Multi-method simulation software (SD, DES, ABS, DS)
  - State charts + Java code

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- Knowledge representation
  - Frequency distributions for determining state change delays

Situation	Min.	Mode	Max.
Leave browse state after	1	7	15
Leave help state after	3	15	30
Leave pay queue (no patience) after	5	12	20



- Probability distributions to represent decisions made

Event	Probability of event
Someone makes a purchase after browsing	0.37
Someone requires help	0.38
Someone makes a purchase after getting help	0.56

boolean x=(Math.random()<0.37)?true:false;</pre>





• Implementation of customer types

Customor typo	Likelihood to								
Customer type	buy	wait	ask for help	ask for refund					
Shopping enthusiast	high	moderate	moderate	low					
Solution demander	high	low	low	low					
Service seeker	moderate	high	high	low					
Disinterested shopper	low	low	low	high					
Internet shopper	low	high	high	low					

for (each threshold to be corrected) do {
 if (OT < 0.5) limit = OT/2 else limit = (1-OT)/2
 if (likelihood = 0) CT = OT - limit
 if (likelihood = 1) CT = OT
 if (likelihood = 2) CT = OT + limit
}
where: OT = original threshold
 CT = corrected threshold
 likelihood: 0 = low, 1 = moderate, 2 = high</pre>





- Implementation of staff proactiveness
  - Non-cashier staff opening and closing tills proactively depending on demand and staff availability
  - Expert staff helping out as normal staff
- Other noteworthy features of the model
  - Realistic footfall and opening hours
  - Staff pool (static)
  - Customer pool (dynamic)
  - Customer evolution through internal stimulation (triggered by memory of ones own previous shopping experience)
  - Customer evolution through external stimulation (word of mouth)





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# Case Study 1: Implementation

- Performance measures
  - Service performance measures
    - Service experience
  - Utilisation performance measures
    - Staff utilisation
    - Staff busy times in different roles
  - Level of proactivity
    - Frequency and duration of role swaps
  - Monetary performance measures (productivity and profitability)
    - Overall staff cost per day
    - Sales turnover
    - Sales per employee







\*1 = number of people queueing for this service

\*3 = considering accumulated history [number]

\*6 = experience per visit [satisfaction growth]

\*4 = considering accumulated history [satisfaction growth]

\*2 = % of those leaving the queue

\*5 = experience per visit [number]

#### Department: Audio & TV (A&TV) Sunday: Shop open for 8 hours

red: cashier green: normal staff member blue: expert staff member magenta: section manager yellow: department manager cyan: advisor lighter colours: free darker colours: serving very dark colours: supporting (expert advice)

#### 

		real	planned		years	weeks	days	hours	minut	es	Current o	ustomer populat	ion:			8000		
Average arrival rate per h	our:	73	(73)	Runtime:	0	21	0	5	52									
Customers in store: 21	7			Overall custo	mers:			86255	100 %				Transac	tions:		29101		
- browsing: 9				- leave happ	/ (transa	ction or re	efund):	29101	34 %	*1	*2		Av. Tra	nsaction	[£]:	149.7		
- seeking help: 0		3	6	- leave not v	/aiting fo	r normal h	nelp:	2464	3 %	19921	12 %		Sales [£	];		4,356,4	20	
- queuing for help: 0		Č.		- leave not v	/aiting fo	r expert ł	nelp:	826	1 %	1907	43 %		Missed	[£]:		8,551,9	12	
- standard:		0		- leave not v	/aiting to	) pay:		10855	13 %	39001	28 %							
- expert:	1	0		- leave witho	ut findin	g anythin	g:	42982	50 %									
- refund author.:	1	0		- leave unhaj	opy (no i	refund):		0	0 %	Custom	ers left:		86228		477406			
- getting help: 7													*3	100 %	*4	*5	100 %	*6
- standard:		7		Till queue ler	ngth: me	an: 3.78;	max: 13	7.0		- satisfie	ed (> 0):		61697	72 %	518960	35188	41 %	10156
- expert:	i	0		Normal help	queue le	ngth: me	an: 1.25	5; max: 14	4.0	- don't	know (= 0):		10574	12 %		40652	47 %	
- refund author.:	1	o		Expert help	queue le	ngth: mea	an: 0.08	); max: 4.	0	- not sa	tisfied (< 0):		13957	16 %	-41554	10388	12 %	-26726
- wait at till: 8										Overall	refunds:		0	100 %				
- to pay:	1	8		Overall Satisf	action Le	evel Index	C.	477406		- refund	ls accepted:		0	0%				
- for refund:	i	0		- shopping:				477406		- refund	ls denied:		0	0%	*1	*2		
- served at till: 3			8	- refund:				0		- leave i	not waiting for r	efund decision:	0	0%	0	0%		
- to pay:		3								- leave i	not waiting for a	author. decision:	0	0%	0	0%		
- for refund:		0								Overall	decisions by casl	nier:	0					
			Importa	nt parameters:						Overall	decisions by aut	horised person:	0	22	242	25		
Finite population:			- Replica	ation number:				З				1 served 255					11 served	10
- shopping enthusiasts:		400	- Empov	verment level o	f cashier	for refund	ds:	0.7				2 served 435				10	12 served	10
- solution demanders:		3200	- Probab	obability that refund is granted by cashier:		0.8			8	3 served 265				1	13 served	10		
<ul> <li>service seekers:</li> </ul>		3200	- Probab	pility that refund	is grante	ed by auti	horiser:	0.7				4 served 164				10	14 served	10
- disinterested shoppers	s: •	400	- Probab	pility that staff s	ay with	customer	:	0				5 served 74				1	15 served	10
<ul> <li>internet shoppers:</li> </ul>		800	- Points	required to bec	ome an	expert:		100000		1		6 served 47				35	16 served	10
intNumBroactiveOpport	unity		- Word	of mouth adopt	ion fracti	on:		0.5				7 served 25				1	17 served	10
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## **Case Study 1: Experimentation**

#### • A&TV: 2 cashiers, 4 normal staff, 4 expert staff

Overall customers:	41235	100 %				Transac	tions:		12057		
- leave happy (transaction or refund):	12057	29 %	*1	*2		Av. Tra	nsaction	[£]:	149.7		
- leave not waiting for normal help:	930	2 %	8839	11 %		Sales [£	]:	1,804,933		33	
- leave not waiting for expert help:	134	0%	583	23 %		Missed (	[£]:		4,367,9	47	
- leave not waiting to pay:	7468	18 %	19128	39 %	-						
- leave without finding anything:	20646	50 %									
			Custom	ers left:		41235		122742			
						*3	100 %	*4	*5	100 %	*6
Till queue length: mean: 4.23; max: 19.0			- satisfied (> 0):			24972	61 %	144905	15682	38 %	48215
Normal help queue length: mean: 1.09; max: 13.0		3.0	- don't know (= 0):			8085	20 %		19670	48 %	
			- not sa	tisfied (< 0):		8178	20 %	-22163	5883	14 %	-13796

\*1 = number of people queueing for this service

\*2 = % of those leaving the queue

\*3 = considering accumulated history [number]

- \*4 = considering accumulated history [satisfaction growth]
- \*5 = experience per visit [number]
- \*6 = experience per visit [satisfaction growth]





## **Case Study 1: Experimentation**

#### • A&TV: 3 cashiers, 6 normal staff, 1 expert staff

Overall customers:	40960	100 %			Trar	sactior	ns:		16800		
- leave happy (transaction or refund):	16800	41 %	*1	*2	Av.	Transa	iction	[£]:	149.7		
- leave not waiting for normal help:	1724	4 %	10958	16 %	Sale	; [£]:	2,514,960		60		
- leave not waiting for expert help:	761	2 %	1085	70 %	Miss	ed [£]:	:		3,616,7	52	
- leave not waiting to pay:	1687	4 %	15605	11 %							
- leave without finding anything:	19988	49 %									
			Custom	ers left:	409	60		136411			
					*3	1	00 %	*4	*5	100 %	*6
Till queue length: mean: 2.15; max: 17.0			- satisfied (> 0):		279	79 61	8 %	152775	18512	45 %	50894
Normal help queue length: mean: 1.56	; max: 14	4.0	- don't	know (= 0):	757	9 1	9%		18924	46 %	
			- not sa	tisfied (< 0):	540	2 13	3 %	-16364	3524	9%	-11610

- \*1 = number of people queueing for this service
- \*2 = % of those leaving the queue
- \*3 = considering accumulated history [number]
- \*4 = considering accumulated history [satisfaction growth]
- \*5 = experience per visit [number]
- \*6 = experience per visit [satisfaction growth]





## Case Study 2

(For more details see Zhang et al 2010)

A non-queuing system





#### Case Study 2: Context

- Office building energy consumption
  - We focus on modelling electricity consumption
  - Organisational dilemma
    - Need to meet the energy needs of staff



- Need to minimise its energy consumption through effective organisational energy management policies/regulations
- Objective
  - Test the effectiveness of different electricity management strategies, and solve practical office electricity consumption problems





- Electricity consumption
  - Base electricity consumption: security devices, information displays, computer servers, shared printers and ventilation systems.
  - Flexible electricity consumption: lights and office computers
- Current electricity management technologies
  - Each room is equipped with light sensors
  - Each floor is equipped with half-hourly metering system
- Strategic questions to be answered
  - Automated vs. manual lighting management
  - Local vs. global energy consumption information





- The mathematical model
  - Ctotal = Cbase + Cflexible
    - where Cflexible =  $\beta 1^*Cf1 + \beta 2^*Cf2 + ... + \beta n^*Cfn$
    - and Cf1 ...Cfn = maximum electricity consumption of each flexible appliance
    - and  $\beta 1 \dots \beta n$  = parameters reflecting the behaviour of the electricity user
      - $-\beta$  close to 0 = electricity user switches flexible appliances always off
      - $-\beta$  close to 1 = electricity user leaves flexible appliances always on
  - Ctotal = Cbase + ( $\beta$ 1\*Cf1+  $\beta$ 2\*Cf2+ ... +  $\beta$ n\*Cfn)





- Knowledge gathering
  - Consultations with the school's director of operations and the university estate office
  - Survey amongst the school's 200 PhD students and staff on electricity use behaviour (response rate 71.5%)
- User stereotypes
  - Working hour habits
    - Early birds, timetable compliers, flexible workers
  - Energy saving awareness
    - Environment champion; energy saver; regular user; big user





Conceptual model





- Energy user agent
  - Proactive





The University of Nottingham



- Computer agent
  - passive



- Light agent
  - passive



- Office agent
  - passive







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	Trigger type: Timeout 💌 Mode: Cyclic 💌	Controls
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# **Case Study 2: Experimentation**

- Validation
  - Comparing simulation and empirical results







# **Case Study 2: Experimentation**

- Scenario #1
  - Comparing automated and manual operation (low user interaction)



Automated operation: Base scenario (simulation) Manual operation: High user interaction (simulation)





## Case Study 3

(For more details see Zhang et al 2012)

A non-queuing system





# Case Study 3: Context

- Modelling the effects of user learning on forced innovation diffusion
  - Technology adoption theories assume that users' acceptance of an innovative technology is on a voluntary basis
  - Sometimes the adoption decisions are made by a few authoritative individuals and implemented enforcedly
- Our Focus is on:



- Classical consumer learning theories
- Residential energy consumer in the City of Leeds
- Interventions that local authorities can take to manage smart metering deployments





• Residential Energy Consumer (REC) template







• Residential Energy Consumer (REC) archetypes

	Attributes									
Archetype	Property energy efficiency	Greenness of behaviour	Duration of daytime occupancy							
1: Pioneer Greens	High	High	Short							
2: Follower Greens	Low	High	Short							
3: Concerned Greens	Low	High	Long							
4: Home-Stayers	High	High	Long							
5: Unconscientious Wasters	High	Low	Short							
6: Regular Wasters	Low	Low	Short							
7: Daytime Wasters	High	Low	Long							
8: Disengaged Wasters	Low	Low	Long							





• Residential Energy Consumer (REC) agent





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**ABM-MDS Short Course** 



## Case Study 3: Experimentation

• Validation: Simulated Load Curve vs. Real Load Curve



• Experiment: Inexperienced vs. Experiences REC





# **Further Reading**

- Siebers and Aickelin (2008) Introduction to Multi-Agent Simulation
- Macal and North (2007) Agent-Based Modeling and Simulation: Desktop ABMS
- Bonabeau (2002) Agent-Based Modeling: Methods and Techniques for Simulating Human Systems



#### **Questions or Comments**





#### References

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