





Simulating livestock systems

- Study the animal behaviour
- Simulation model building
- Model verification and validation
- RSM Regression Optimization
- Work with the model, meetings, discussions
- Conclusions, reporting, documentation

Topic of this course

- ► Simulation optimization– needs and theory
 - The paradox
- The solution
- ► Case studies
- Milking robots
- Aquaculture
- Ictical implementation Animal behavior sensors EU projects





Advantages of (simulation) models Study the model instead of the real system... • usually much easier, faster, cheaper, safer Can try wide-ranging ideas with the model Simulating living animals: • Allows the isolation of a single parameter, • Any number of repetitions, 100% repeatability

The real power of Simulation

- Studying complex systems where analytical methods fail
- Allows system variability in modeling

Advantages of modelling

Same technology; for both fish and cow

• Quantify the animal behaviour

Build simulation model





The Bad News

The main limitation of simulation lies in its heuristic character:

- ▶ simulation responses are observed only
- for the selected input combinations, i.e.,
- ► there is <u>no proof of the optimality</u> of the solution.

The solution (1)

- •:
- ► Given a validated model
- Design experiments DOE
 Response surface methodology (RSM)
- Fitting polynomial function via Multiple Regression
- ► Optimization
- ► Simulation runs: fine tuning grid around the optimum
- Meetings in the farm, analyze, get insight, document results

The solution (2)

- ► the first step is to select the combination
- of parameters that is to be simulated in
- simulation experiments.
- ► In the simulation literature, this phase is
- called "design of experiment," or DOE

(Banks, 1998).





Example, Sde Eliho aquaculture farm

- <u>"... 448,000</u> possible input combinations. Using DOE, the number of simulation runs needed was set to <u>3,880</u>".
- Halachmi et al.,(2005) Aquaculture Engineering 32: 443–464

D-Optimal Designs Matlab implementation settings = cordexch(2,9,'q'),h=plot(settings(:,1),settings(:,2),'.'); set(gca,'Xtick',[-1 0 1]); set(gca,'Ytick',[-1 0 1]); set(h,'Markersize',20);





First Order Strategy

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Quadratic Models can only take certain forms					
Maximum	Minimum	Saddle Point	Stationary Ridge		
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Run	Facto origina	ors in al units	Facto codec	ors in Lunits	Response
	Time (min.)	Temp. (°C)			Yield (gms)
	<i>x</i> ₁	<i>x</i> ₂	x_1	x_2	У
1	70	127.5		200	54.3
2	80	127.5	+	-	60.3
3	70	132.5	-	+	64.6
4	80	132.5	+	+	68.0
5	75	130.0	0	0	60.3
6	75	130.0	0	0	64.3
7	75	130.0	0	0	62.3



Must Use Regression to find the Predictive Equation

The Quadratic Model in Original units $\hat{y} = -3977 + 17.86 * time + 45.00 * temp$ -0.0975 * time * temp - 0.0215 * time² - 0.1247 * temp²

RSM – Example II – 4th step – goal function

The yearly profit in the model is given by:	
$P = p \times T - f \times F \times iBW - f_{c} \times T \times f_{f} - o \times T - E - L - tr \times T - i - w$	(8)
and	
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RSM – Example II – step 5 -	
constraints	

max. P subjected to the following constraints:	(9)
Stocking density in phase $1 < 55 \text{ kg m}^{-3}$ Stocking density in phase $2 < 65 \text{ kg m}^{-3}$ 3500 < back bize < 7000 35 g < initial body weight < 80 g $2.4 gday < weight gain in phase 1 < 2.9 gday10 < number of small or large tanks < 1417 \text{ m}^3 < \text{small tank volume} < 34 \text{ m}^334 \text{ m}^3 (\text{ arge tank volume} < 40 \text{ m}^3)$	(Constraint 1) (Constraint 2) (Constraints 3 and 4) (Constraints 5 and 6) (Constraints 7 and 8) (Constraints 1 and 12) (Constraints 11 and 12)
11 (μ9) 11 (μ9) 10 (μ9) 12 (μ9) 13 (μ9) 14	

The solution (3)

- Equation 2 is a convex function, and consequently Kuhn-Tucker conditions are necessary and sufficient for global optimality.
- ► Matlab solves this LP problem by a projection method, which is a variation of the well-known Simplex method (Coleman et al., 1999).

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Optimal	Existing	g <u>Parameter</u>	
Solution	Situation		
160 ton	70 ton	Yearly turnove	r
494K \$	-120K \$	Yearly Profit/los	s
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Recommendation (valid to Sde Eliahu's RAS) - To split one large tank into two small tanks - Size grading frequency - once per 4-5 days - Size grading criteria - 450 g - Batch size 3500-3800 fish, depend on the size grading criteria Dr. Ilan Halachmi; halachmi@volcani.agri.gov.il 30

The interesting reader

Halachmi, I. (2007) Biomass management in recirculating aquaculture systems using queuing networks. *Aquaculture*, 262(514-520.

Halachmi, I. (2006) Systems engineering for ornamental fish production in a recirculating aquaculture system *Aquaculture*, 259(1-4), 300-314.

Halachmi, I., Simon, Y., Guetta, R. & Hallerman, E. M. (2005) A novel computer simulation model for design and management of recirculating aquaculture systems. *Aquacultural Engineering*, 32(3-4), 443–464.

















D	Decision Variables				
x(1)	Time Interval(days)	Time between batches			
x(2)	Batch size\Number of fish per year	Batch size: number of fish in batch. A batch includes fish at the size of 0- 2' (weight -0.1) - input for the <u>simulation</u> . Number of fish per year: (365/time interval) *Batch size. Is the input for the <u>regression</u> , in order to reduce dependence between factors. The <u>range</u> is for the <u>number of fish</u> <u>per year</u> .			
x(3)	Growth Rate	Factor. Value of 1 represents the current growth rate.			
x(13)	Sales Factor	Factor. Value of 1 represents the current growth rate.			
x(4)	Next Stage4	The percentage of fish at stage 4 that continues to the next growi stage (5). (1-next stage4) is the percentage that is going to be sale. under stage 4 are not for sale.			
x(5)	(5) Next Stage5 The percentage of fish at stage 5 that continues to the next of stage (6).				
x(6)	Next Stage6 The percentage of fish at stage 6 that continues to the next growin stage (7).				
x(7)	 Next Stage 7 The percentage of fish at stage 7 that continues to the next grow stage (8). 				
x(8)	Next Stage8	The percentage of fish at stage 8 that continues to the next growing stage (9).			
x(9)	Next Stage9	The percentage of fish at stage 9 that continues to the next growing			





Results				
	AII	Met filter constraints		
# of scenarios	25,530	1417		
R ²	0.652	0.852		
Explanatory Variables	8	11		
G′(X*)	30.26 M	20.8 M		
Simulation (X*)	9.8 M	19.8		
The gap: (RSM- Simulation)	21 M NIS	1 M NIS		



Cows examples Robotic milking A mega dairy in India Automatic Lameness detection Body condition scoring



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Designing the optimal robotic milking barn (RMB	2
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Milking Robots

 Halachmi I., Metz J.H.M., Maltz E., Dijkhuizen A.A., and Speelman L. (2000). Designing the optimal robotic barn, Part 1: quantifying facility usage, . Journal of Agricultural Engineering Research, 2000. 76: p. 37-49.

 Halachmi I. (2000). Designing the optimal robotic barn, Part 2: Behaviourbased simulation. Journal of Agricultural Engineering Research, 2000. 77(1): p. 67-79.

8. Halachmi et al. 2002. **Optimal** Facility Allocation in a Robotic Milking Barn The Transactions of the ASAE 45(5): 1539-1546,

5 Halachmi I., Adan I.J.B.F., van der Wal J., Heesterbeek J.A.P., and van Beek P. (2000). The design of robotic dairy barns using **closed queuing networks**. European Journal of Operational Research 124(3): p. 437-446

7 Halachmi I., Dzidic A., Metz J.H.M., Speelman L., Dijkhuizen A.A., Kleijnen J.P.C. (2001). Validation of simulation model for robotic milking barn design: case study. European Journal of Operational Research, 134(3): p. 677-688.

13 Halachmi I. (2004). Designing the Automatic Milking Farm in a Hot Climate. Journal of Dairy Science, 2004. 87(3): p. 764-775. 8 Application in the farms



I'm off to India to live like a Goddess...



Project Aim

Design a mega dairy

- ▶ 10,000 cows in milking
- ► Three rotary milking parlors
- ► Two veterinary hospitals
- ► One animal-feed center
- ► Cow-manure handling
- & biogas production
- Cow cooling centers
- Calves, heifers, replacement
- Workers' traffic and facilities













Design tool 4. Optimization			
A deterministic design problem:	A probabilistic design problem:		
Minimizes: F(µ _y (X))	Minimizes: $F(\mu_y(X), \sigma_y(X))$		
subject to: $gi(\mu_y(X)) \leq 0$	$\label{eq:subject} \text{subject to:} \qquad \qquad \text{gi}(\mu_y(X), \ \sigma_y(X)) \leq 0$		
$X_L \leq \mu_X \leq X_U.$	$X_L {+} n \sigma X \leq \mu X \leq X_U {-} n \sigma X$		
$\mu_{y} - n\sigma_{y} \ge \text{Lower specification limit}$ $\mu_{y} + n\sigma_{y} \le \text{Upper specification limit}$ $\mathbf{n} = 6$			



- A mega dairy subsystems Seven simulation models were built: • Milking parlor cow flow (model 1) • In-parlor treatment cow flow (model 2) • Cow traffic to the milking parlor and cooling sheds (model 3) • Junction flow near the milking parlor (model 4) • Manure scraping (flow?) (model 5) • Feed-distribution flow (model 6) • Worker traffic flow (model 7)
- A mega dairy as a one single system
- In this project, the model comprises
- Optimization maximizing capacity of each facility
- Queuing network links all the facilities into one single system
- ► Reliability Quality over Time
- Robust (6 sigma) design











