

# Data analysis and validation of call center models

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based on joint work with Sihan Ding, Siqiao Li,  
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Workshop on  
Data Driven Operations Management  
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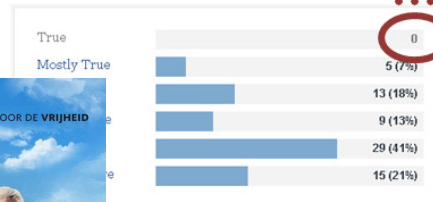
# What is data-driven OM?

- Politics is becoming fact-free...

- in the US...

**POLITIFACT**  
WINNER OF THE PULITZER PRIZE

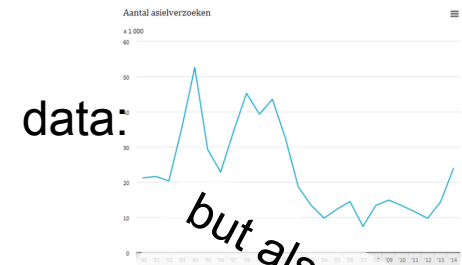
Donald Trump's file



- and in Holland



“tsunami of refugees”



- maths is to some extent fact-free...

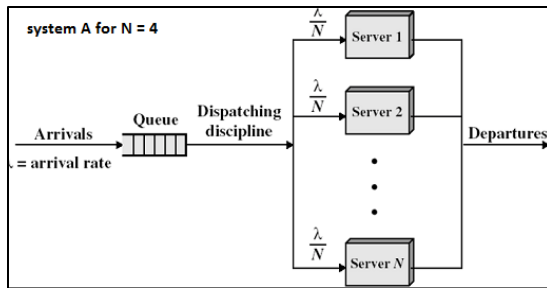
- but applied science has always used facts = data

- what's new? availability of data? *big data*?

- *integration* of statistics in decision making

but also scientists  
lie with data!

# Why use models in call centers?



model: SL predictions



realized SL



X - 4w: planning

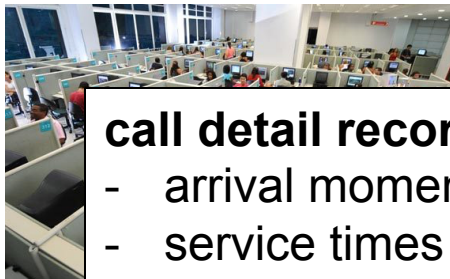
time →

eXecution

- 10M's of agents are scheduled based on SL predictions
- Question: is prediction close to realization?
- No, because parameters change (especially FC)
  - Therefore: flexibility in WF + real-time rescheduling
  - Also: validation of model is impossible
- Consequence: nobody knows if model is “correct”

# Research question

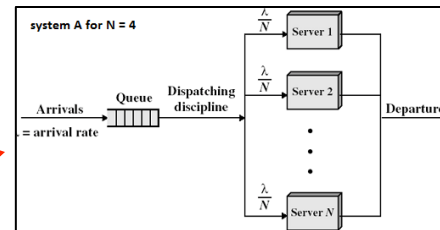
- Is prediction biased?
  - is it correct given parameter values?
- Method = compare afterwards



## call detail records

- arrival moments
- service times
- waiting times

compute  
parameters



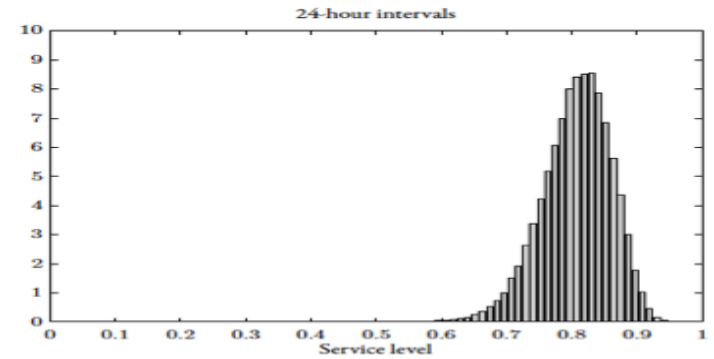
model-based SL

realized SL

- Data = 1 yr in multi-skill cc (VANAD/city of R'dam)
- Solution method = simulation



# Objective



- Data gives daily SL  $x_t^{\text{act}}$ 
  - No i.i.d. replications
- Model outcome = estimated daily SL = r.v.  $X_t^{\text{sim}}$
- Goal: Estimate “model error”  $w \sum_t | EX_t^{\text{sim}} - EX_t^{\text{act}} |$

$$w \sum_t | EX_t^{\text{sim}} - EX_t^{\text{act}} | \leq w \sum_t | EX_t^{\text{sim}} - x_t^{\text{act}} | + w \sum_t | EX_t^{\text{act}} - x_t^{\text{act}} |$$

simulation

$$w \sum_t | EX_t^{\text{act}} - x_t^{\text{act}} | \approx w \sum_t | EX_t^{\text{sim}} - x_t^{\text{sim}} |$$

simulation

# Models

- More than 1 model
- Options concern:
  - model for arrivals
  - handling time distribution
  - handling times agent-dependent
  - shrinkage = breaks
  - solution method (Erlang/SIPP/intra-day-sim)
  - ...

# Arrival process

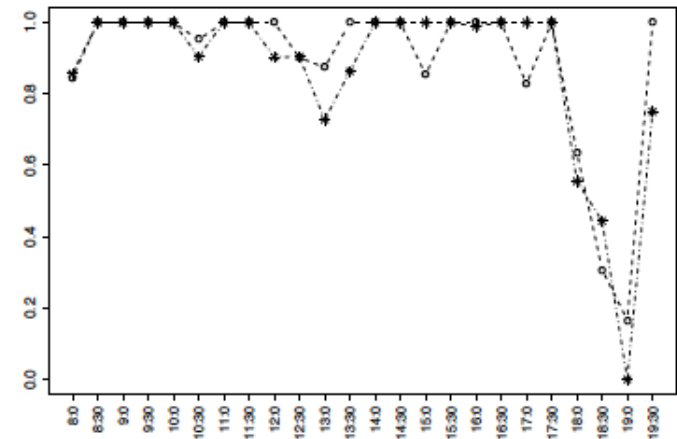


## Options:

- Use real rate – unknown
- Use forecast – bad, not enough data
- Use actuals – cheating

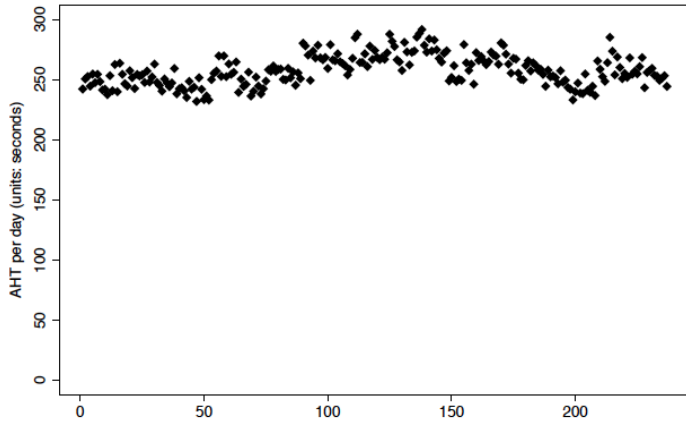
## Solution:

- Use actuals and estimate impact of cheating
- By a simulation experiment with the (bad) FC

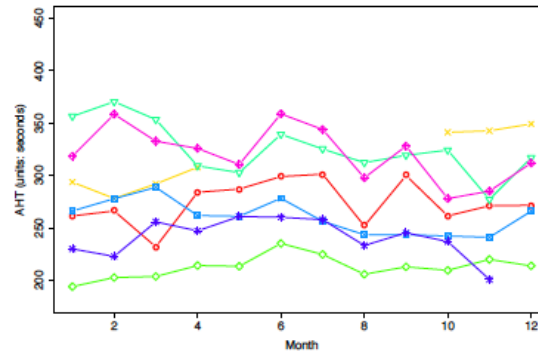


15-min SL with IPP based on rates and actuals

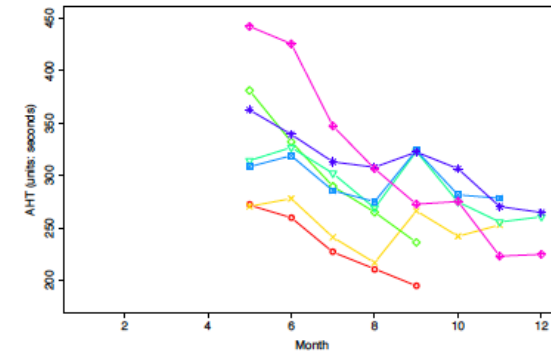
# Results: AHT



Overall AHT fluctuates....

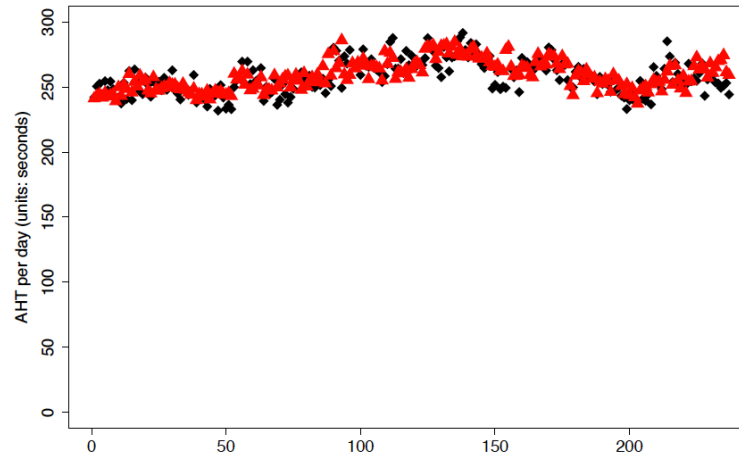


also per agent



$$EAHT_{j,m_i} = \alpha_j e^{\omega_j m_i}$$

model per agent and  
function of time

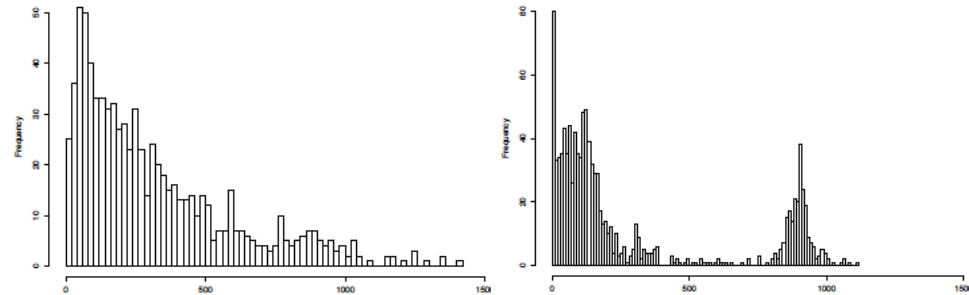
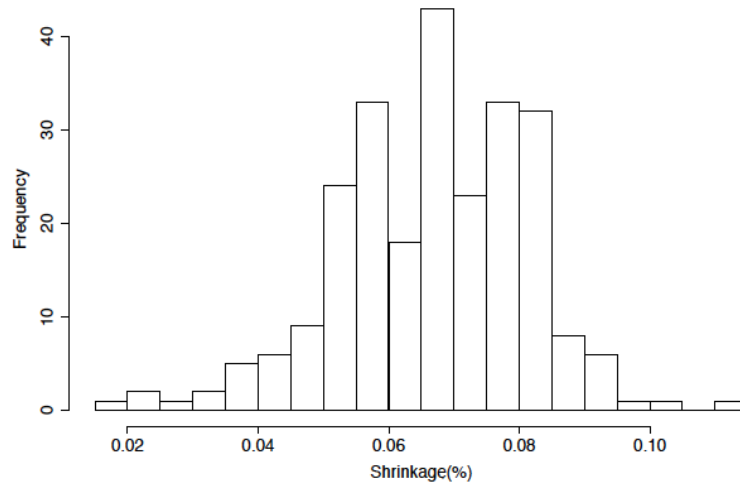


new AHT prediction ( $R^2 = 50\%$ )

WAE  
reduction of  
2.6%



# Results: breaks



durations are also considerable and vary between agents

loss in availability due to breaks:  
substantial + varies between agents

WAE  
reduction of  
5.4%

other issues looked into:

- patience
- wrap-up time
- handling time distribution

# Conclusions (so far)



- Agent variability in handling times matter
- Breaks should be modeled