

## Case research and design science ...

... as research paradigms for statistical engineering?

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## Why this talk?

Case studies published in journals

- Potentially ...
  - Rich source of information about the challenges of applying statistics to solve real, messy, complex problems
- But often ...
  - Not many lessons being articulated
  - No reflection on whether these lessons can be generalized
  - No attempts to integrate lessons learned into more coherent new theory

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## Not-so-useful case study

### Six Sigma project report

#### 1. Objective

We did an interesting project. The goal is to improve a production process.

#### 2. Case description

We applied some factorial designs, and fitted some regression and ANOVA models. From these we found valuable process improvements.

#### 3. Conclusions

The new process settings saved the company lots of money.



#### What did we learn?

What lessons can we generalize?

How does this help to build a body-of-knowledge about applying statistics?

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## Better-but-still-not-there case study

QUALITY ENGINEERING  
2019, VOL. 31, NO. 3, 484–498

### CASE REPORT

#### Statistical reasoning in diagnostic problem-solving measurements

Jeroen de Mast<sup>a</sup>, Stefan H. Steiner<sup>a</sup>, Rick Kuijten<sup>b</sup>, and

#### Research question (as opposed to goals in the case)

Our purpose in presenting and discussing this case is to demonstrate the roles that statistical thinking can play in diagnostic problem-solving, and identify reasoning patterns that make the application of statistical techniques powerful.

#### Description of what we did to solve the case

#### Conclusions for the case

##### Discussion and

##### Conclusions in the flow-rate case

Based on statistical analysis, we have been able to eliminate many potential causes of the discrepancies,

##### Statistical reasoning patterns and conclusions

This article is more about statistical reasoning than

##### Hypothesis generation, testing, and evaluation

- Principle: Instead of blindly looking for correla-

##### Branch-and-prune strategy

- Principle: When the search space is complex,

##### Pareto heuristic

- Principle: The aim o

#### Lessons learned for applied statistics

##### Explanatory coherence

- Principle: Hypotheses are retained or rejected depending on whether they explain a substantial part of the problem, whether they are parsimoni-

Cases as a research strategy  
to build a body-of-knowledge  
about applying statistics to solve  
problems.

## Why this talk?

### Purpose of this talk

Propose a framework for case study research ...

... to help SE build a body-of-knowledge about applying statistics to solve complex unstructured problems.

*Design Science Research* has the potential to offer such framework.

ISEA website:

"To advance insertion into academic curricula and to enhance the professional qualifications and standing among its members"

Talented young people will go elsewhere unless SE has credibility and standing in Academia

### 2<sup>nd</sup> purpose of this talk

Present *Design Science Research* as a potential and strong foundation for SE, that is recognizable to the wider academic community ...

... to help SE gain credibility in the wider academic community.

## What is *design science*, and how does SE fit in?



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## What is *design science*?

Main research strategy in applied sciences

- Engineering
- Medicine: “evidence-based medicine”
- Now being adopted by business schools



Contents lists available at [ScienceDirect](#)

Journal of Operations Management

journal homepage: [www.elsevier.com/locate/jom](http://www.elsevier.com/locate/jom)

Conducting and publishing design science research  
Inaugural essay of the design science department of the Journal of  
Operations Management

### 1.2. *DSR can be regarded as an engineering approach to OM*

DSR in operations management can be regarded as a conscientious transfer of the strategy used in engineering research, taking into account the fundamental differences between designing and

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## What is design science?

Main research strategy in applied sciences

- Engineering
- Medicine: “evidence-based medicine”
- Now being adopted by business schools

Roots in the work of Herbert Simon

- Book: *The Sciences of the Artificial*
- Framework for sciences not so much seeking *truth* but seeking *actionable knowledge*
  - How to build a bridge?
  - How to cure a patient?
  - How to solve a problem?
- “Means/end” thinking



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## Explanatory, formal and design science

### Explanatory sciences

*Physics, biology, psychology, ...*

- Discover laws of nature
- *How*: empirical studies
- *Goal*: truth finding



### Formal sciences

*Mathematics, logic*

- Develop logical framework for reasoning
- *How*: logical deductions
- *Goal*: logical consistency



### Design sciences

*Engineering, medicine, management science, ...*

- Discover actionable knowledge: techniques, methods, approaches, strategies, principles, ...
- ... as a means to achieve an end
- *Goal*: effectiveness (“it works”)



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## Design science produces actionable knowledge

Product of design science research: *prescriptions*

- Methods, techniques, protocols, strategies, analysis templates, reasoning patterns

Generic form:

- In a context  $C$
- this approach  $A$
- is likely to result in outcome  $O$
- which we can understand as the effect of the working mechanisms  $M$

*Denyer et al. (2008); Van Aken et al. (2016)*

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## Sources of knowledge: 1. Science

### 1. Scientific knowledge

- Mathematical statistics
  - Framework for inferences from data (estimation, hypothesis testing, ...)
- Machine learning and AI
  - Algorithms for (mainly) predictive analytics
- Problem structuring and problem solving
  - Diagnostic problem solving, creative problem solving, structuring messy problems, ...
  - DMAIC, CRISP-DM, pyramid principle, mess maps, ...
- Organizational and management theory
  - Project-mgt, leadership, influencing skills & winning support, organizational development, getting things done, ...

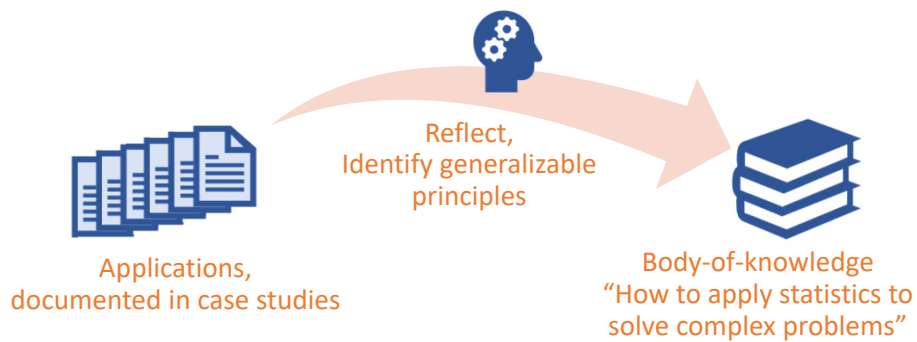
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## Sources of knowledge: 2. cases

### 2. Experiential knowledge

- Learn lessons from earlier applications: “case-based reasoning”
- Case studies!
- Better for dealing with the messier, unstructured challenges of practical statistics that cannot be understood from formal mathematics



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## Case-study research

How to learn from case studies?

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## Case-study research in Management

Eisenhardt K.M. (1989) "Building theories from case study research," *Academy of Management Review* 14(4).

Handfield R.B., Melnyk S.A. (1998) "The scientific theory-building process: A primer using the case of TQM," *Journal of Operations Management* 16.

Voss C., Tsikriktsis N., Frohlich M. (2002) "Case research in operations management," *Int. Journ. of Operations & Production Management* 22(2).

Stuart I., McCutcheon D., Handfield R., McLachlin R., Samson D. (2002) "Effective case research in operations management: A process perspective," *Journal of Operations Management* 20.

Barratt M., Choi T.Y., Li M. (2011) "Qualitative case studies in operations management: Trends, research outcomes, and future research implications," *Journal of Operations Management* 29.

Ketokivi M., Choi T. (2014) "Renaissance of case research as a scientific method," *Journal of Operations Management* 32.

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## Case-study research in Management

Guidelines for case-study research in Management:

- Research question  
(what does the researcher hope to learn from the case?)
- Motivation that a case study is the right research method
- Unit of analysis
- Literature review of relevant theory
  - This yields working hypotheses related to the research question
    - Helps to make the research question more specific, as a choice between alternative hypotheses
    - Helps to determine what observations are needed
- Selection of case or cases
- Systematic collection and documentation of observations
- Within-case analysis
- Cross-case analysis
- Findings and conclusions

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## Case-study research in Management



**The problem of appointment scheduling in outpatient clinics:  
A multiple case study of clinical practice**

Alex Kuiper<sup>a\*</sup>, Jeroen de Mast<sup>b</sup>, Michel Mandjes<sup>c</sup>



Mathematical definition of problem  
as trade-off between *idle times* and *waiting times*

$$(t_1, t_2, \dots, t_n) =$$



Does mathematical problem definition  
make sense in clinical practice?  
Are we solving the right problem?



Extensive literature review

→ interview questions



- Idle time irrelevant
- Complex constraints and large variety
- No optimization, but flexible adjustment

Selection of 10 cases  
(clinics)

| Case | Specialty           | #<br>clinicians |
|------|---------------------|-----------------|
| 1    | Internal medicine   | 27              |
| 5    | Orthopedics         | 26              |
|      | Otorhinolaryngology | 10              |
|      | Ophthalmology       | 7               |
| 8    | Ophthalmology       | 32              |
| 9    | Neurology           |                 |
| 10   | Orthopedics         |                 |

- Within-case analyses
- Cross-case analysis



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## Case studies in SE

Proposed structure for case studies in SE:

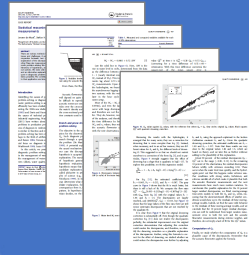
Research question  
*What generalizable  
lessons for SE do we  
hope to learn?*



Literature  
*What's already known  
in scientific literature?*



Presentation of the  
case(s)



Discussion of the case

- Explicate
- Validate
- Understand
- Generalize

... lessons learned



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## Case studies in SE

Discussion of the case:

- *Explicate*  
Articulate what the applied approach was
- *Validate*  
Critically evaluate how effective the chosen approach really was.
  - What was the desired outcome, and to what extent was it achieved?
- *Understand*  
Try to explain the approach's effectiveness from results from mathematics or other sciences
  - What are the working mechanisms of the approach?
- *Generalize*  
Discuss in what situations the approach could or could not be effective?
  - How generalizable is the approach?

*Denyer et al. (2008); Van Aken et al. (2016)*

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## Case studies in SE

Proposed structure for case studies in SE:

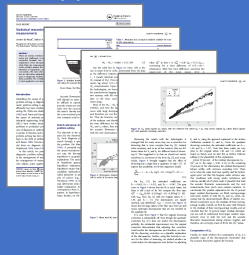
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Literature  
*What's already known in scientific literature?*



Presentation of the case(s)



Discussion of the case

- Explicate
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Result:

"In a context *C*, this approach *A* is likely to result in outcome *O*, which we can understand as the effect of the working mechanisms *M*."

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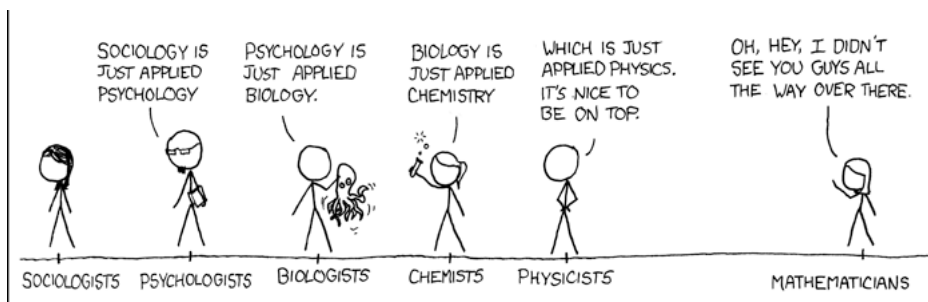
## Applied sciences as autonomous knowledge domains

Is applied science equally respectable as theoretical science?



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## Is SE just applied mathematical statistics?



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## Engineering is just applied physics?

Herbert Simon

- Fed up with scorn of theoretical scientists
- Wrote *The Sciences of the Artificial* to explain the relation between applied and theoretical science.

Two domains of knowledge

Understand internal  
working of tools



Understand functional  
use of tools

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## Two domains of knowledge: cars ...

Under the hood



Understand car's internal working  
Laws of mechanics, electronics

From behind the steering wheel

Understand a car's functional use

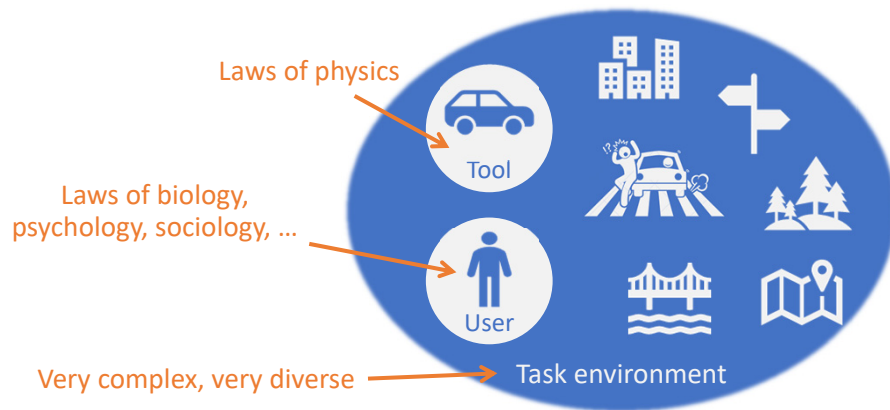
Traffic rules, driving skills, motor skills,  
knowledge of functions of pedals and  
meaning of indicator lights, ...



The *functional use* of a car, and its *internal working*,  
are two *disjoint* domains of knowledge.  
Mechanics and electronics do not help much in learning how to use a car.

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## Tools in isolation vs. as purposive system



Design sciences

are about *interaction* between user, tool and task environment

- What sort of purposes do users have?
- What sort of environments are there?
- What tool would be effective for those purposes in those environments?

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## Engineering is just applied physics?

- A car understood purely from laws of physics is but a purposeless metal object with rubber, plastics and copper wire.
- Understand a car as a functional tool for a user with purposes within a variety of task environments
  - Does not make sense to *design* a car without knowledge about driving behavior and challenges, traffic rules, ergonomics, etc.
  - Does not make sense to limit *driving lessons* to the internal working of cars (mechanics and electronics)
- Engineering *uses* physics, but is much more!

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## SE is just applied mathematics?

The internal logic and algorithmics  
of statistical techniques

Mathematical stats.  
(deductive logic)



The logic of their functional  
application in inquiry, decision making  
and problem-solving

Stat. engineering  
(means/end logic)

- User:
  - What sort of goals in empirical inquiry, decision making, problem solving?  
(*delicate point: many mathematicians have no first-hand experience with empirical inquiry*)
- Task environment:
  - What sort of challenges, practical complications, issues, ...  
do users encounter when applying statistics?
- Which tools are effective for those purposes in those environments?

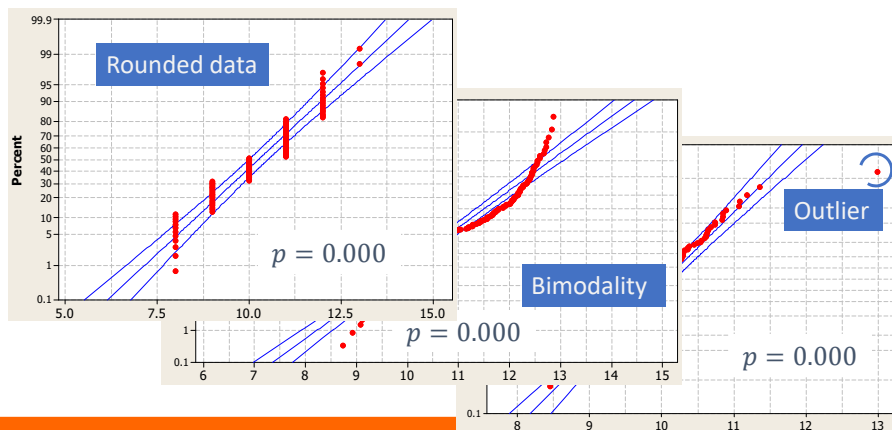
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## Example: normality tests

Normality tests ...

- Mathematical logic:
  - Inferential framework for rejecting or not rejecting  $H_0$ : normality
  - How to define optimal test statistic and calculate  $p$ -value



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## Example: *normality tests*

Normality tests ...

- Mathematical logic:
  - Inferential framework for rejecting or not rejecting  $H_0$ : normality
  - How to define optimal test statistic and calculate  $p$ -value
- Statistical engineering logic:
  - $H_0$  usually implausible on a priori grounds

More interesting questions:

- Is the normal distribution a useful approximation? How can you tell?
- What kind of departures from normality are there?  
(outliers, multimodality, rounded data, skewness, ...)
- And how to deal with them in your study?  
(data cleaning, transformation, other probability distribution, nonparametrics, ...)

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## Take-aways

- Case-study research is useful for building a body-of-knowledge about statistical engineering
  - Detail-rich, empirical sources of knowledge about realistic challenges in applied statistics and problem solving
  - Opportunities to learn about what does and doesn't work in the complex and messy world of application
  - But ... lessons in cases should be explicated, validated, understood and generalized ...  
otherwise we end up with little but an enumeration of case descriptions, but without generalizable lessons and a growing body-of-knowledge
- The art of applying statistics in inquiry, decision-making and problem-solving merits scientific study in itself
  - "Statistical engineering"
  - Altogether different knowledge domain than mathematical statistics
  - Answers questions of the form:  
In which contexts and for what outcomes are which approaches effective?

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