

# The effect of Tilburg gold on your health

**Pacmed meetup: Combining causal inference and machine learning in practice**

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# Introduction

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# Example of a dataset dataset

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X_1	X_2	X_3	X_i	I	Y0	Y1
Male	9	14	1	0	67	NA
Female	60	36	0	1	NA	113
Female	7	2	1	1	NA	54
...	...	...	...	...	...	...

# Prediction versus understanding

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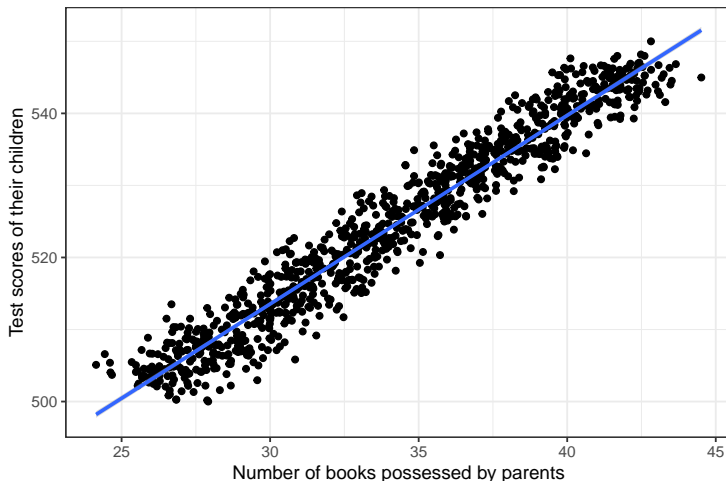
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# Voorspellen versus begrijpen

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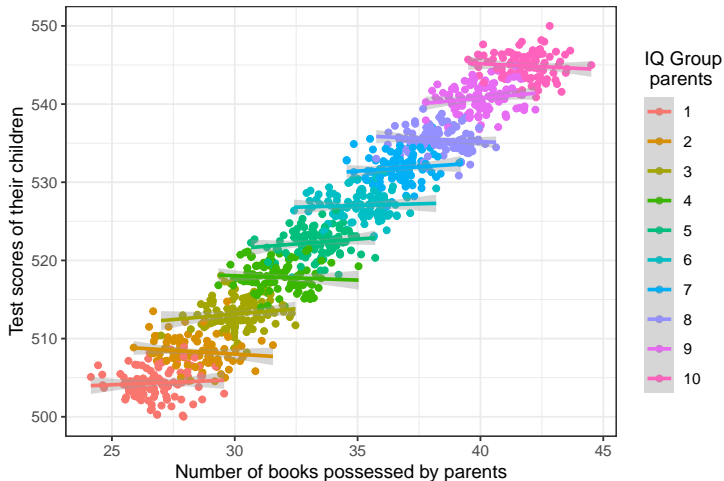
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# DAG building blocks

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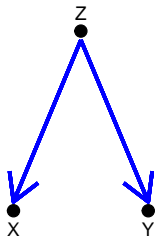
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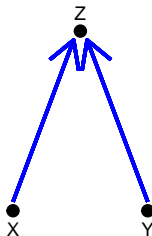
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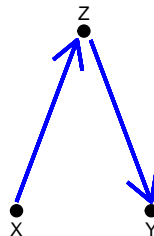
Confounder



Collider



Mediator



# Confounder

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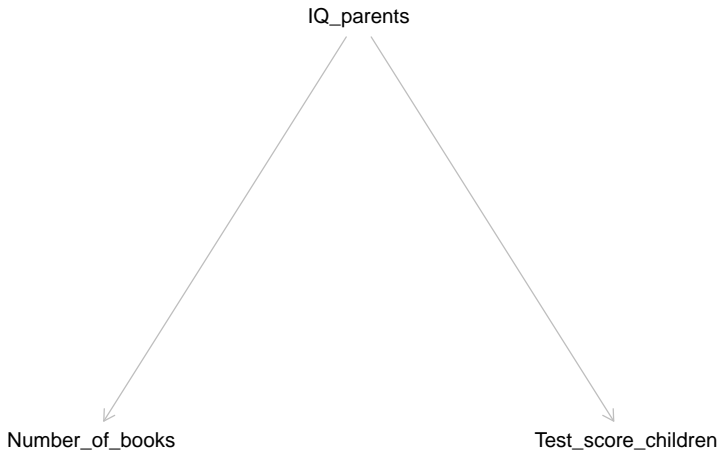
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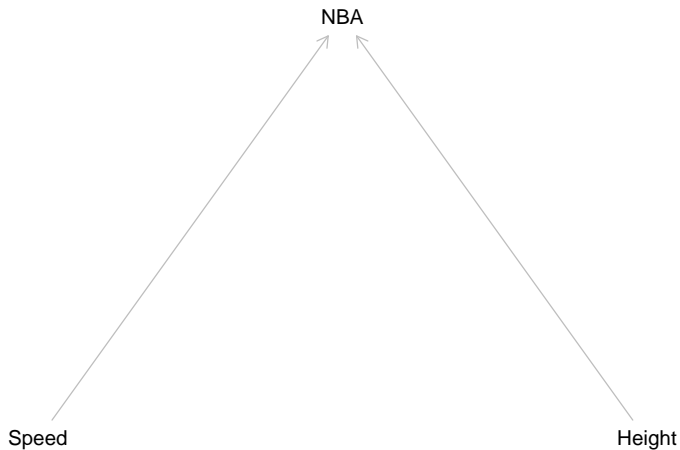
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# Does height causes speed?

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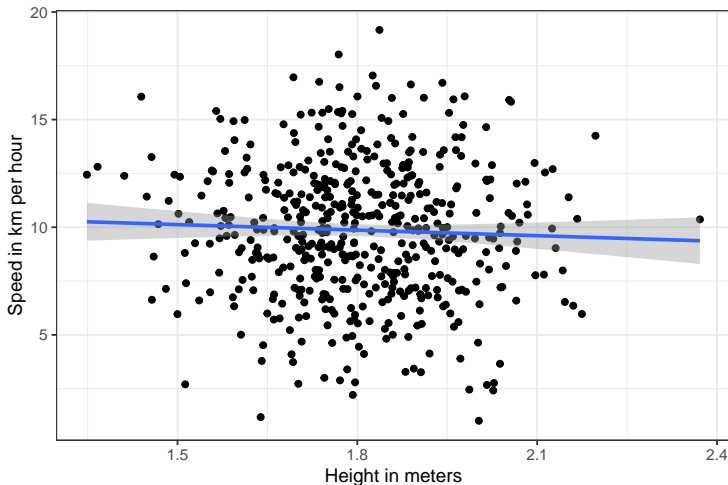
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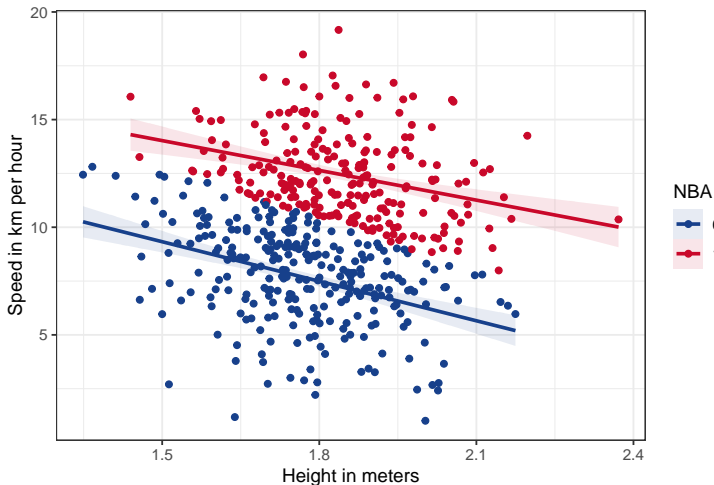
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# Mediator

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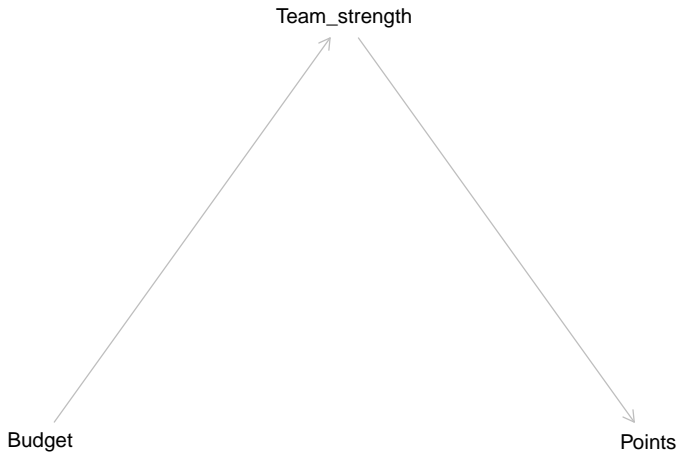
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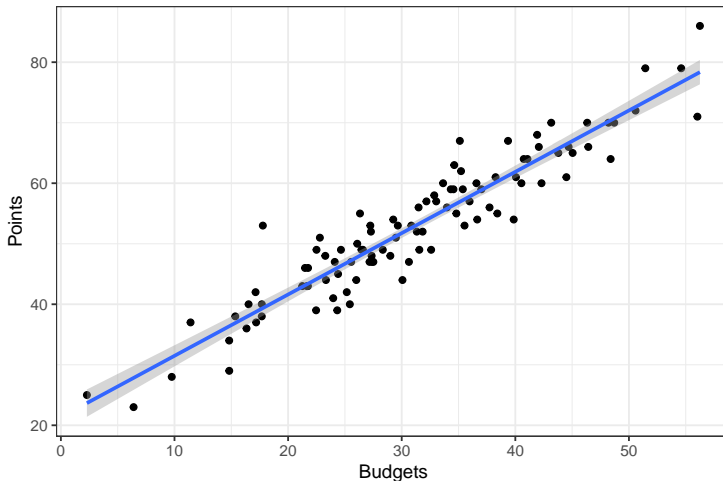
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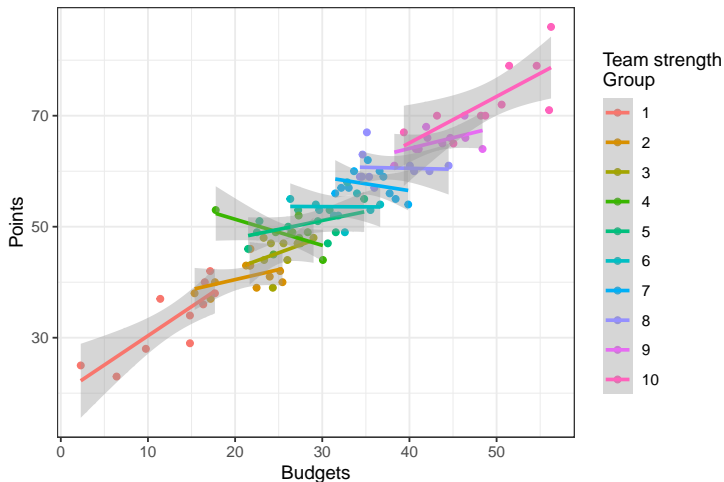
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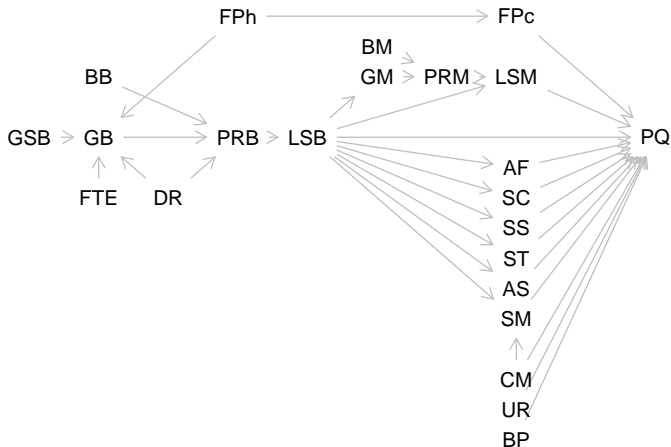
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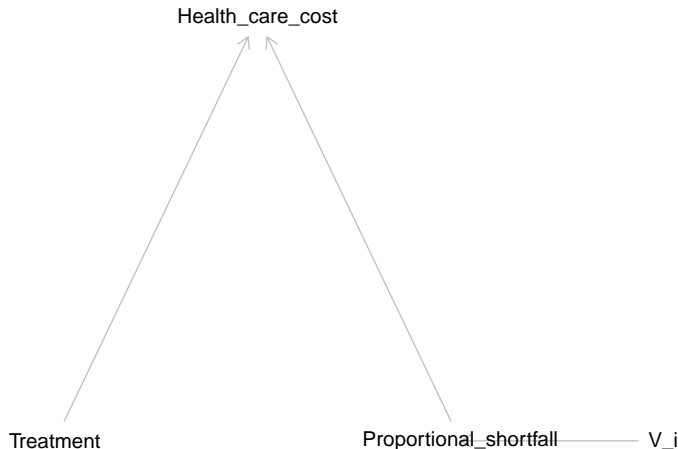
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The variable *Proportional\_Shortfall* is based on:

$$\begin{aligned} \text{Proportional\_Shortfall} = \text{abs}(\text{scale}(V1^3 + 2 * V2 + \\ 3 * V3^2 + 4 * V4 + 5 * V5 \\ + 6 * V6 * V7)) + \epsilon \end{aligned}$$

1. Fit Random Forest model on the data
2. Determine the average treatment effect with generalized random forests (grf)

We will fit 2 models for each of these steps:

- a. An analysis with all variables ("the wrong model")
- b. An analysis with all variables, except the collider *Health\_care\_cost* ("the right model")

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# Summary statistics

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Statistic	N	Mean	St. Dev.	Min	Max
Proportional Shortfall	1,000	1.114	0.859	0.003	4.778
Treatment	1,000	0.474	0.500	0	1
Health care cost	1,000	6.976	4.746	-1.578	27.557
V1	1,000	498.746	288.860	2	1,000
V2	1,000	492.975	293.667	1	1,000
V3	1,000	512.349	293.801	1	1,000
V4	1,000	489.041	288.225	1	1,000
V5	1,000	516.896	292.376	1	1,000
V6	1,000	497.232	284.475	1	1,000
V7	1,000	505.715	289.385	1	1,000
V8	1,000	501.330	276.913	3	998

# Predictions Random Forest

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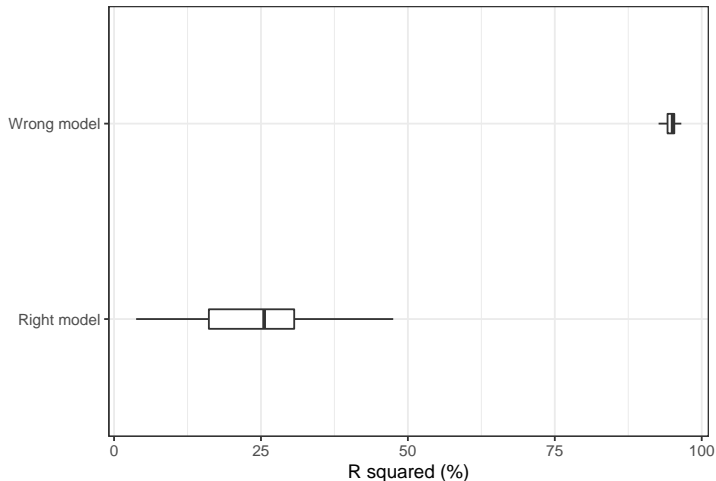
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# Average treatment effect

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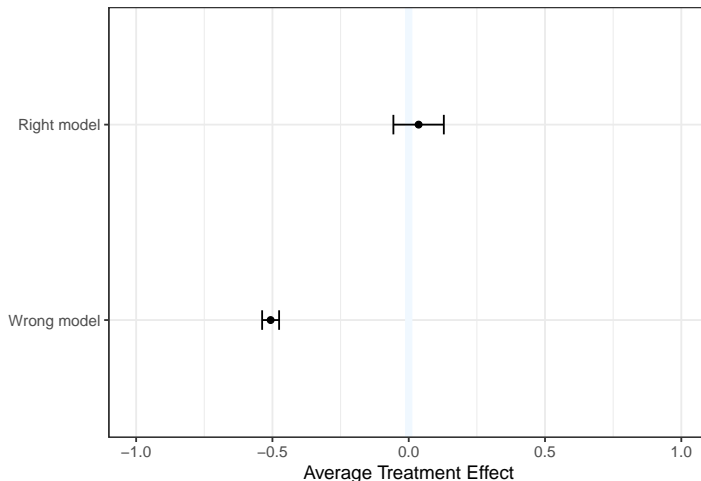
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- ▶ Causal models are necessary for inference
- ▶ It is tempting to use all variables in a machine learning model
- ▶ However, this could lead to misleading conclusions

# Blog and code (in Dutch)

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[https://misjamikkers.github.io/post/  
causaliteit-en-machine-learning/](https://misjamikkers.github.io/post/causaliteit-en-machine-learning/)

[https://github.com/misjamikkers/Meetup\\_Informatieberaad](https://github.com/misjamikkers/Meetup_Informatieberaad)