# A categorization and comparison of performance measures for estimated nonlinear associations with an outcome

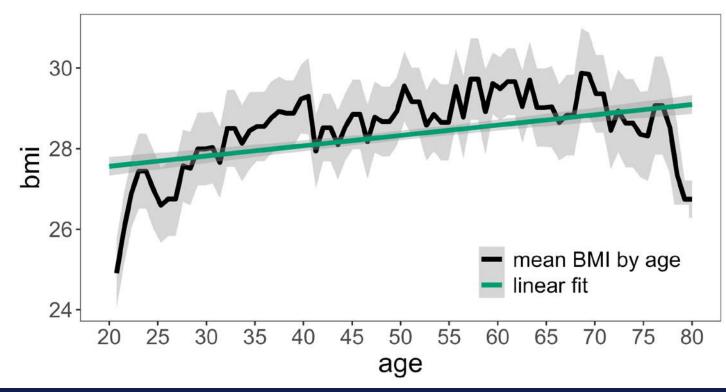
Theresa Ullmann, Georg Heinze, Michal Abrahamowicz, Aris Perperoglou, Willi Sauerbrei, Matthias Schmid, Daniela Dunkler, for TG2 of the STRATOS initative

Presenter: Georg Heinze Institute of Clinical Biometrics, Center for Medical Data Science, Medical University of Vienna



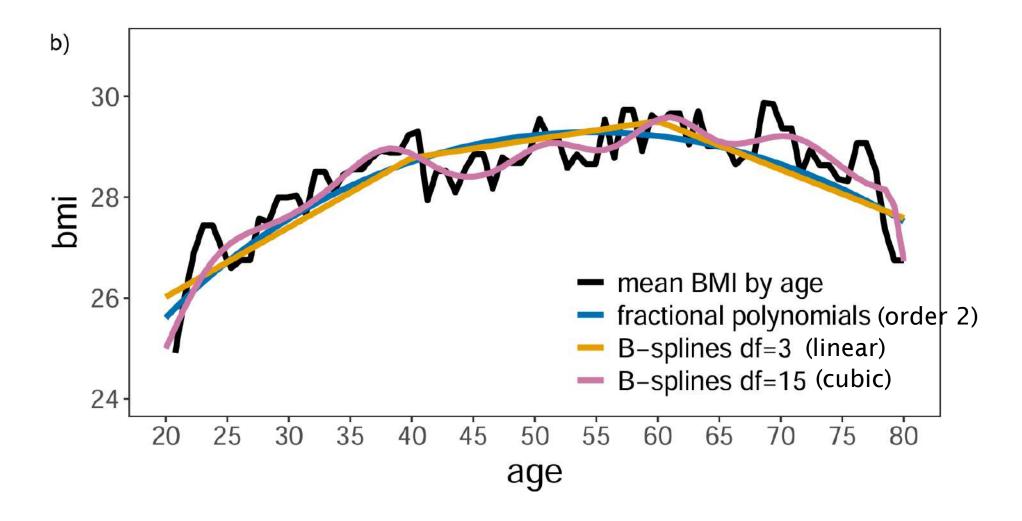
## **Background & motivation**

- Consider the association of BMI with age (NHANES)
- How to separate systematic from unsystematic variation?
- Linear model probably a poor smoother



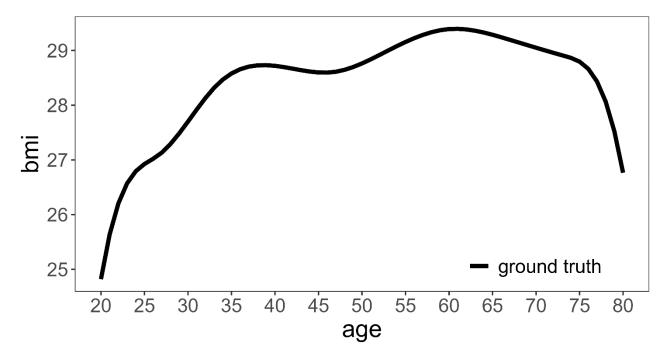


### More smoothers...





- Aim: to compare the performance of different methods of nonlinear modeling
- Data generation mechanism:

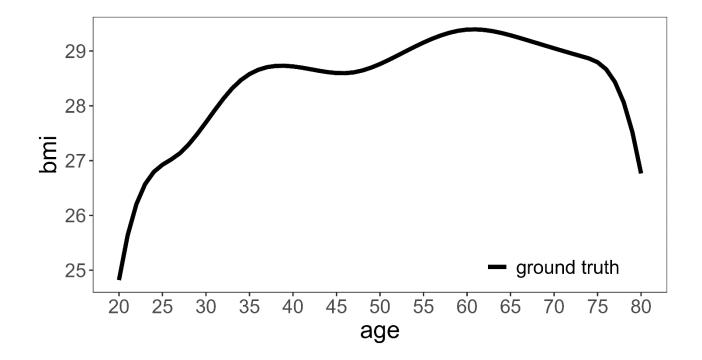


• Estimand: predicted BMI

ADEMP: Morris et al, StatMed 2019

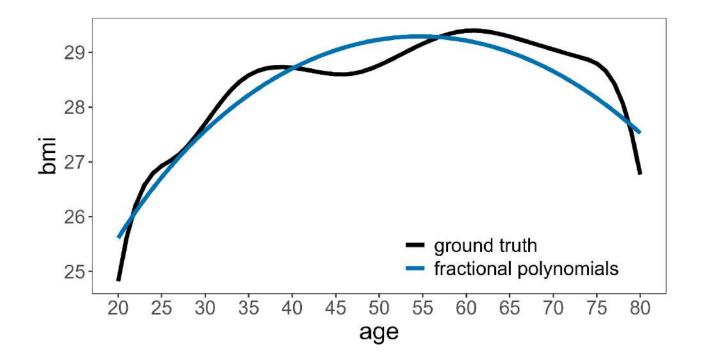


• Methods:



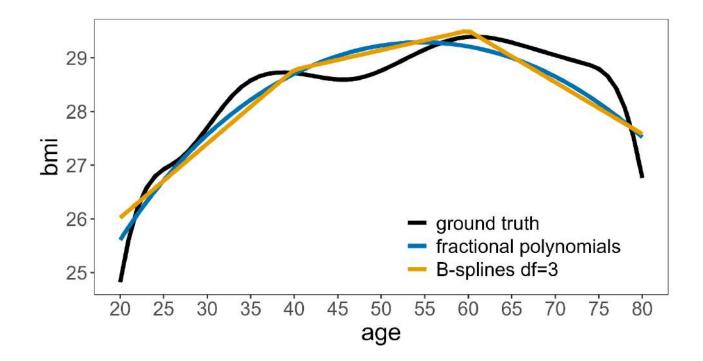


• Methods:



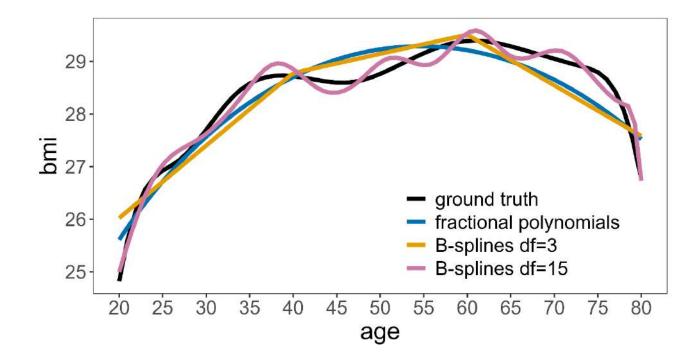


• Methods:



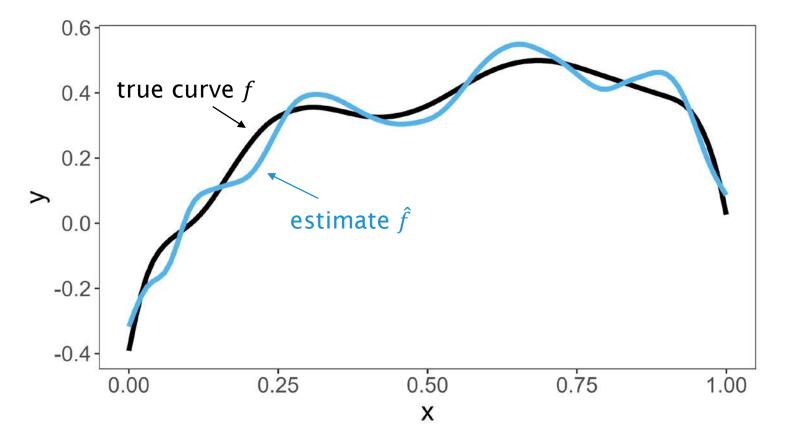


• Methods:





• Performance measures: compare estimated with true curve





• Performance measures:

 $\int_{F_X^{-1}(0.01)}^{F_X^{-1}(0.99)} |\hat{f}(x) - f(x)|\hat{p}(x)dx \quad \text{Buchholz et al. (2014) (see also Govindarajulu et al., 2007)}$ 



• Performance measures:

 $\int_{F_X^{-1}(0.01)}^{F_X^{-1}(0.99)} |\hat{f}(x) - f(x)|\hat{p}(x)dx$  Buchholz et al. (2014) (see also Govindarajulu et al., 2007)

$$\int_{F_X^{-1}(0.05)}^{F_X^{-1}(0.95)} \left(\hat{f}'(x) - f'(x)\right)^2 \mathrm{d}F_X(x) \quad \text{Binder et al. (2011)}$$



• Performance measures:

### Region of interest: 1st to 99th percentile of $F_{\chi}$

 $\int_{F_X^{-1}(0.01)}^{F_X^{-1}(0.99)} |\hat{f}(x) - f(x)|\hat{p}(x)dx$  Buchholz et al. (2014) (see also Govindarajulu et al., 2007)

### Region of interest: 5th to 95th percentile of $F_{\chi}$

$$\int_{F_X^{-1}(0.05)}^{F_X^{-1}(0.95)} \left(\hat{f}'(x) - f'(x)\right)^2 \mathrm{d}F_X(x) \quad \text{Binder et al. (2011)}$$



### • Performance measures:

### Absolute loss

 $\int_{F_X^{-1}(0.01)}^{F_X^{-1}(0.99)} |\hat{f}(x) - f(x)| \hat{p}(x) dx \qquad \text{Buchholz et al. (2014) (see also Govindarajulu et al., 2007)}$ 

$$\int_{F_X^{-1}(0.05)}^{F_X^{-1}(0.95)} \left(\hat{f}'(x) - f'(x)\right)^2 \mathrm{d}F_X(x) \quad \text{Binder et al. (2011)}$$

Quadratic loss



• Performance measures:

### function

 $\int_{F_X^{-1}(0.09)}^{F_X^{-1}(0.99)} |\hat{f}(x) - f(x)| \hat{p}(x) dx \qquad \text{Buchholz et al. (2014) (see also Govindarajulu et al., 2007)}$ 

$$\int_{F_X^{-1}(0.05)}^{F_X^{-1}(0.95)} \left(\hat{f}'(x) - f'(x)\right)^2 \mathrm{d}F_X(x) \quad \text{Binder et al. (2011)}$$

first derivative



• Performance measures:

Integral weighted with precision

 $\int_{F_X^{-1}(0.01)}^{F_X^{-1}(0.99)} |\hat{f}(x) - f(x)| \hat{p}(x) dx \qquad \text{Buchholz et al. (2014) (see also Govindarajulu et al., 2007)}$ 

$$\int_{F_X^{-1}(0.05)}^{F_X^{-1}(0.95)} \left(\hat{f}'(x) - f'(x)\right)^2 \mathrm{d}F_X(x) \quad \text{Binder et al. (2011)}$$

Integral over distribution of X

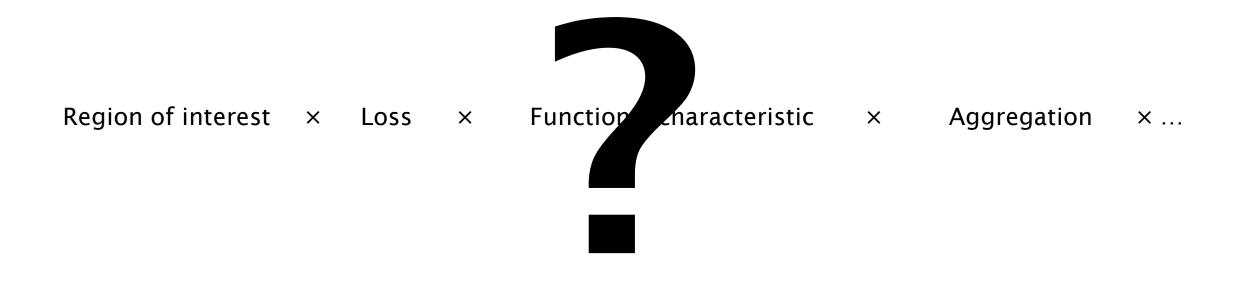


• Performance measures:

Region of interest  $\times$  Loss  $\times$  Functional characteristic  $\times$  Aggregation  $\times \dots$ 



• Performance measures:





# Aims of this project

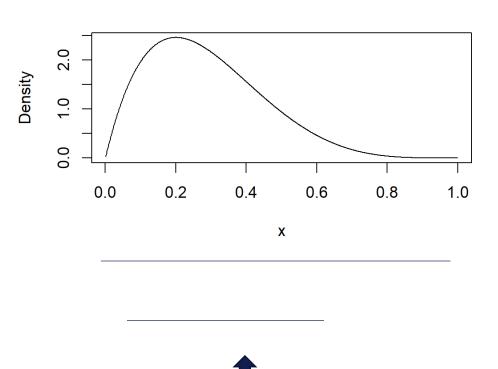
- To provide a comprehensive characterization of performance measures to be used in methods comparison studies
  - Define aspects of such measures
  - Suggest sensible combinations of choices for each of the aspects
- To demonstrate with simple illustrative examples and some hypothetical ,methods'
  - How the resulting performance measures behave
  - That different performance measures capture different aspects of behaviour



### The aspects:

• Localization: Where are we looking at?

- The full range of values (global)
- A subrange (region)
- A single value (point)





#### The aspects: 0.15 0.10 Function • Functional characteristic: 0.05 • The function itself • First derivative 00.0 • Second derivative 0.2 0.4 0.6 0.8 0.0 X 2 0.2 -0.0 Second derivative First derivative 0 -0.2 $\overline{\gamma}$ -0.4 -0.6 2 -0.8 ç -1.0 4 0.2 0.8 1.0 0.0 0.4 0.6 0.0 0.2 0.4 0.8 0.6 х х

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### The aspects:

- Type of loss:
  - Difference:  $m(x) = \hat{f}(x) f(x)$
  - Absolute difference:  $m(x) = |\hat{f}(x) f(x)|$
  - Quadratic difference:  $m(x) = (\hat{f}(x) f(x))^2$
  - $\epsilon$ -level accuracy:  $m(x) = I(|\hat{f}(x) f(x)| \le \epsilon)$



# If we consider the range or a region:

- Axis of aggregation:
  - Y
    - Integration over dx:  $\int m(x) dx$
    - Integration over dF(x) [=expected value):  $\int m(x) dF(x)$
  - X
    - Location of maximum/minimum  $f(x) (= argmax(\hat{f}(x)), argmin(\hat{f}(x)))$
    - Number of roots (e.g. of  $\hat{f}'(x)$ )



### Select the performance measure

Localization:

RangePoint

● Y ○ X Type o

Functional characteristic:

f(x)

⊖ f'(x)

○ f"(x)

Loss:

O Difference

○ Absolute

○ Squared

Epsilon-level
 accuracy

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<b>•</b> Y
ΟX
Type of aggregation:
$\bigcirc$ Integration over $dx$
$\bigcirc$ Expectation over $dF_X$
$\bigcirc$ Quantile with respect to $F_X$
⊖ Maximum
O Minimum
Scope of aggregration:
$\bigcirc$ whole range $[0,1]$

Axis of aggregation:

) subrange  $[F_X^{-1}(0.05), F_X^{-1}(0.95)]$ 

 $= \int \left( \hat{f}(x) - f(x) \right) dx$ 

"mean deviation"



Localization:	Axis of aggregation:
Range	• Y
⊖ Point	$\bigcirc$ X
Functional	Type of aggregation:
characteristic:	) Integration over $dx$
● f(x)	$\bigcirc$ Expectation over $dF_X$
⊖ f'(x)	$\bigcirc$ Quantile with respect to $F_X$
⊖ f''(x)	⊖ Maximum
Loss:	O Minimum
O Difference	Scope of aggregration:
Absolute	$\bigcirc$ whole range $[0,1]$
⊖ Squared	

### $=\int |\hat{f}(x) - f(x)| dx$

### "mean absolute deviation"



### Select the performance measure

Localization:

Range

○ Point

Functional characteristic:

● f(x)

○ **f**(x)

O **f''(x)** 

Loss:

○ Difference

⊖ Absolute

Squared

Epsilon-level
 accuracy

Axis of aggregation: **O**Y OX Type of aggregation:  $\bigcirc$  Integration over dx $\bigcirc$  Expectation over  $dF_X$  $\bigcirc$  Quantile with respect to  $F_X$ ○ Maximum O Minimum Scope of aggregration:  $\bigcirc$  whole range [0,1]⊖ subrange  $[F_{\chi}^{-1}(0.05), F_{\chi}^{-1}(0.95)]$ 

 $= \int \left( \hat{f}(x) - f(x) \right)^2 dF(x)$ 

", expected (over F(x)) squared deviation"



Localization:	x	
⊖ Range	0,75	
Point		
Functional		
characteristic:		
f(x)		
⊖ f'(x)		
○ f'(x)		
Loss:		
O Difference		
⊖ Absolute		
⊖ Squared		
Epsilon-level		
accuracy		
epsilon		
0,05		

 $= I(\left|\hat{f}(0.75) - f(0.75)\right| \le 0.05)$ 

", within  $f(x) \pm 0.05$  at x = 0.75"



### Select the performance measure

Localization:

Range

○ Point

Functional characteristic:

○ f(x)

○ f'(x)

● f''(x)

#### Loss:

○ Difference

⊖ Absolute

O Squared

 Epsilon-level accuracy

forn	nance measure
Ах	is of aggregation:
$\bigcirc$	Y
0	Х
Ту	pe of aggregation:
$\bigcirc$	Integration over $dx$
0	Expectation over $dF_X$
0	Quantile with respect to ${\cal F}_X$
$\bigcirc$	Maximum
0	Minimum
Sc	ope of aggregration:
$\bigcirc$	whole range $[0,1]$
•	subrange $[F_X^{-1}(0.05),F_X^{-1}(0.95)]$

 $= \int_{Q05}^{Q95} \left( \hat{f}''(x) - f''(x) \right)^2 dx$ 

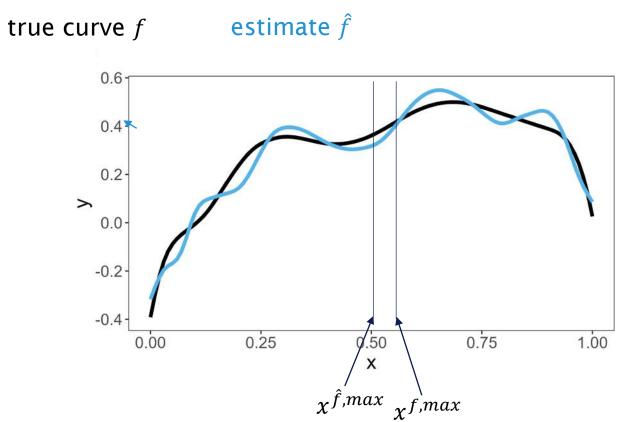
"wiggliness"



Localization: Range	Axis of aggregation:
O Point	ΟY
Functional	• X
characteristic:	Type of
f(x)	aggregation:
⊖ f'(x)	$\bigcirc$ Number of roots
⊖ f"(x)	Location of maximum
Loss:	O Location of
Difference	minimum
O Absolute	Scope of
O Squared	aggregration:
O Epsilon-	<ul> <li>whole range</li> </ul>
level	[0,1]
accuracy	
	$\bigcirc$ subrange $[F_x^{-1}(0.05), F_x^{-1}(0.95)]$

MEDICAL UNIVERSITY OF VIENNA "Deviation of location of maximum":

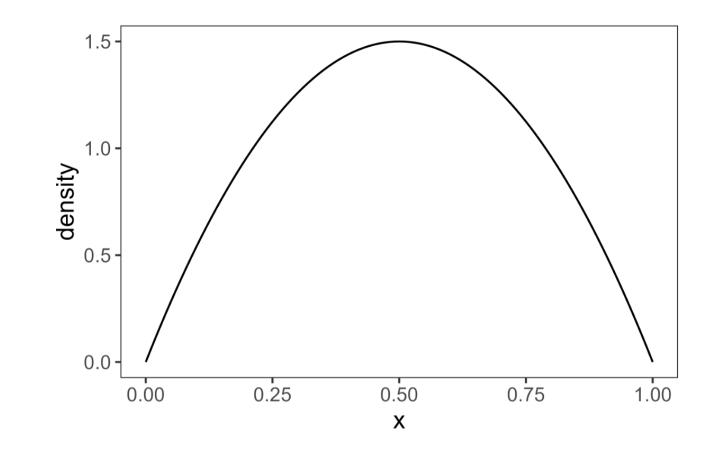
$$x^{\hat{f},max} - x^{f,max}$$



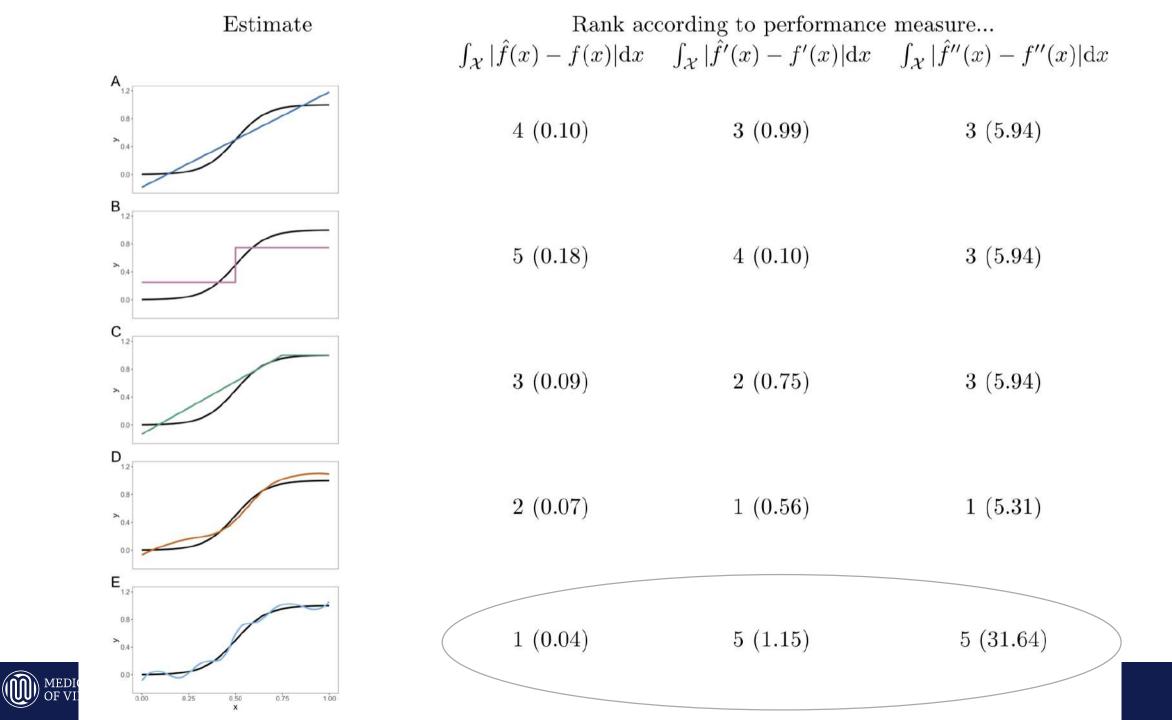
### Some examples

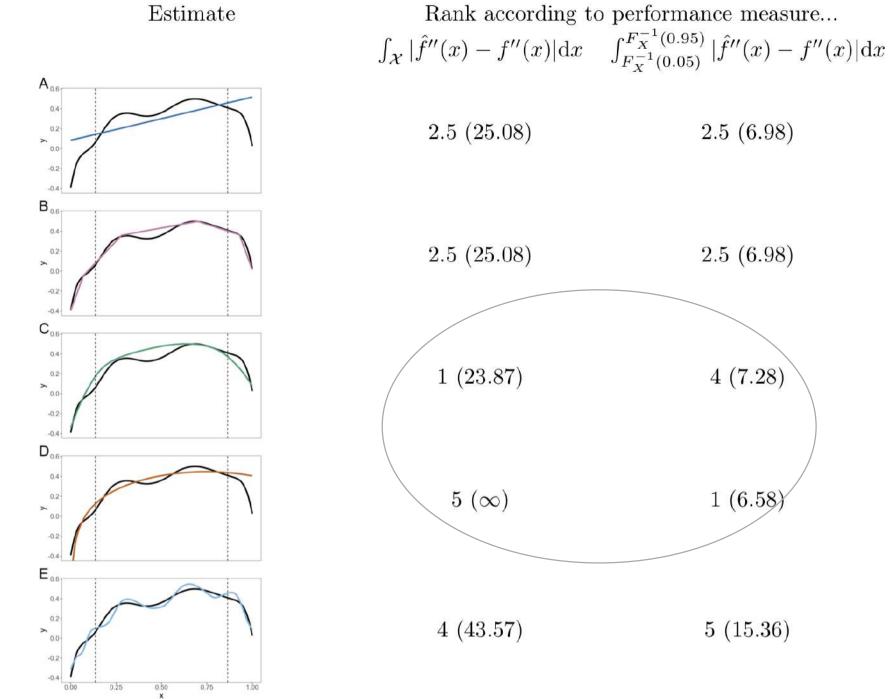
 In these examples, we consider x distributed as Beta(2,2)

- In some examples, we will nevertheless perform the integration over dx
- In others
   we will integrate over
   dF(x)

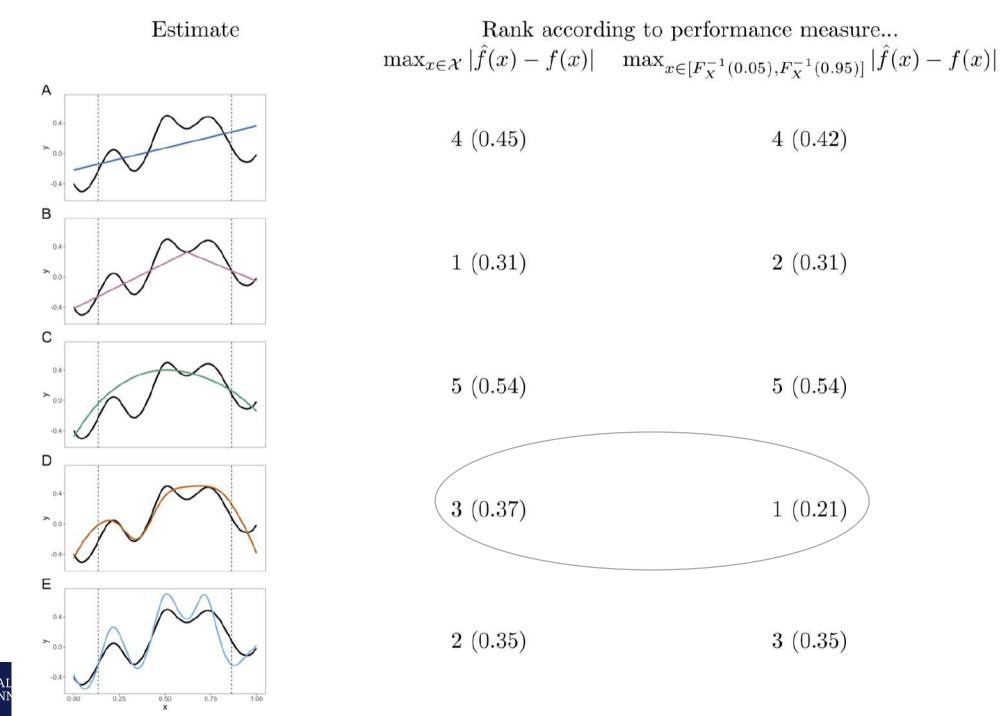


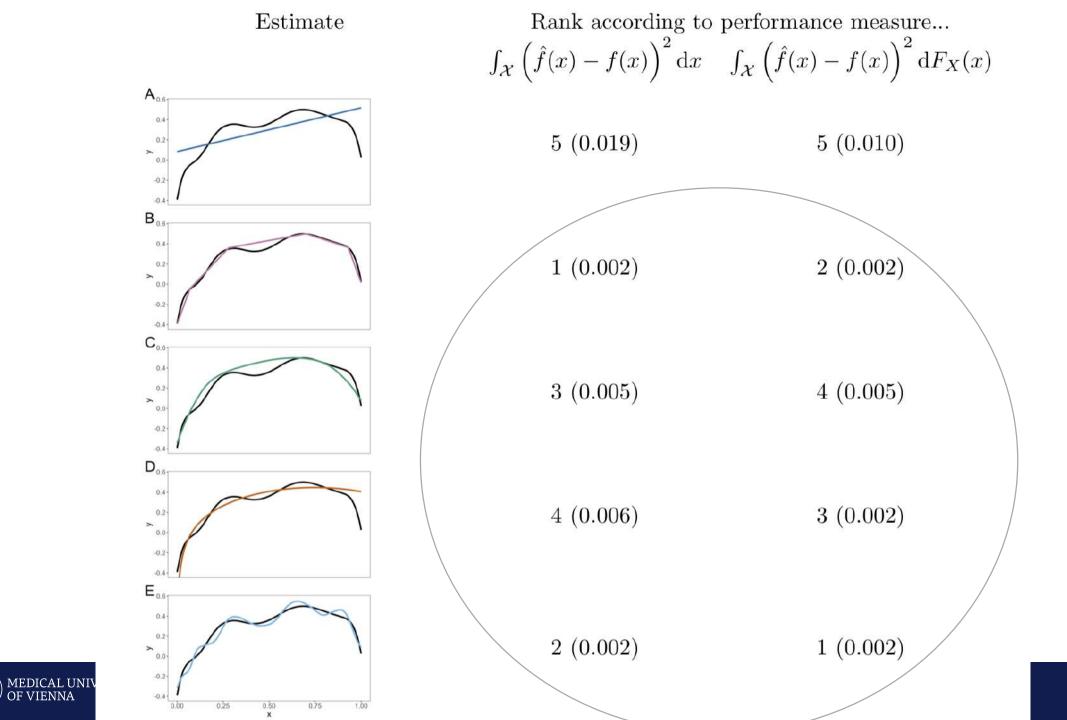


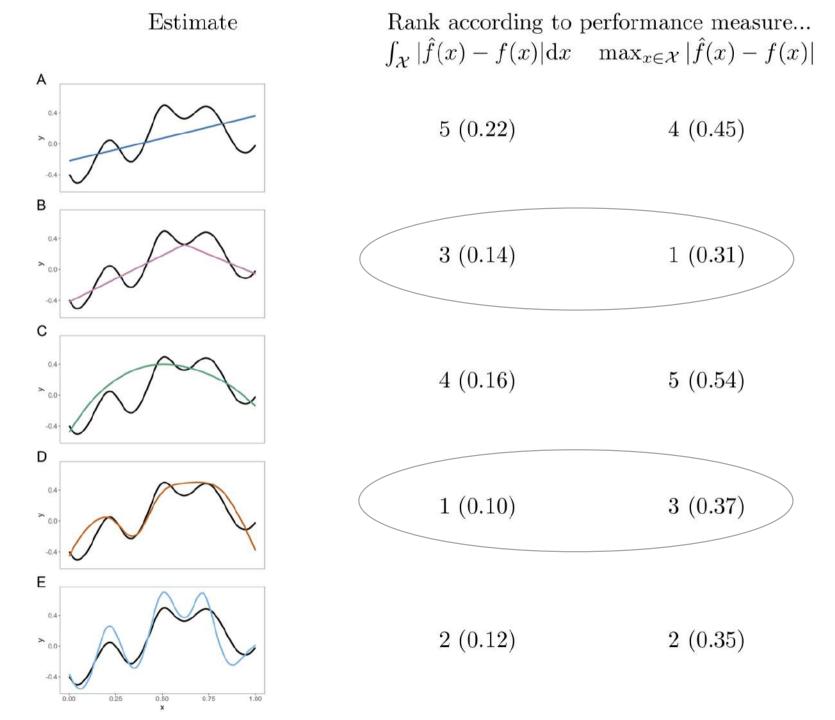




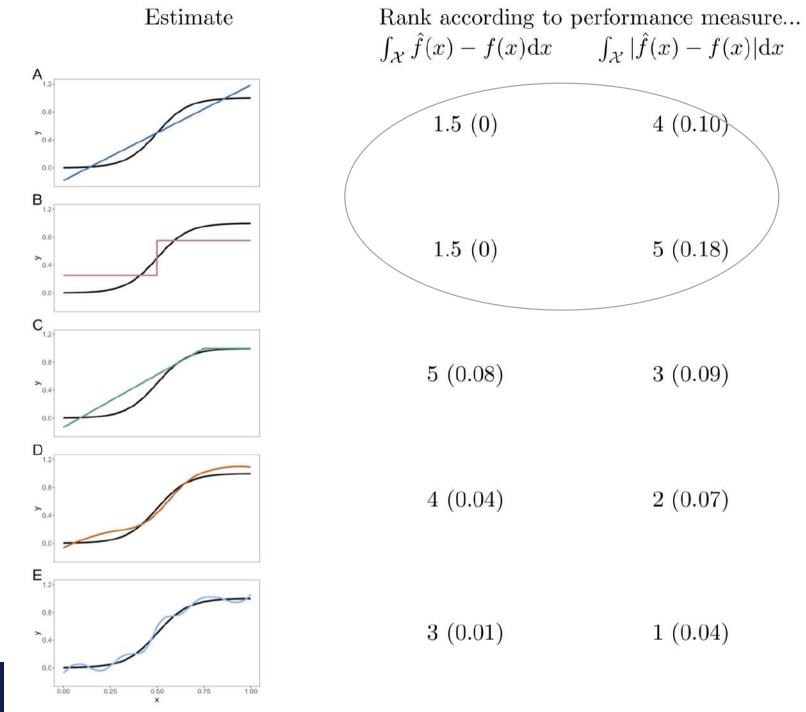




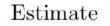


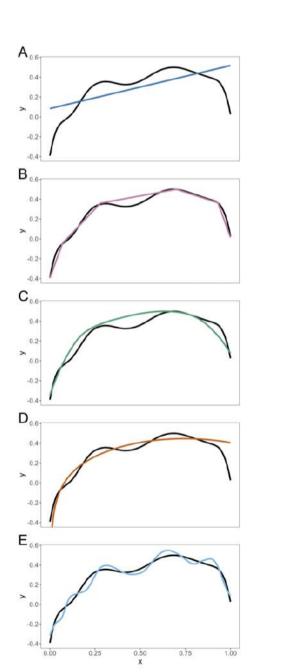




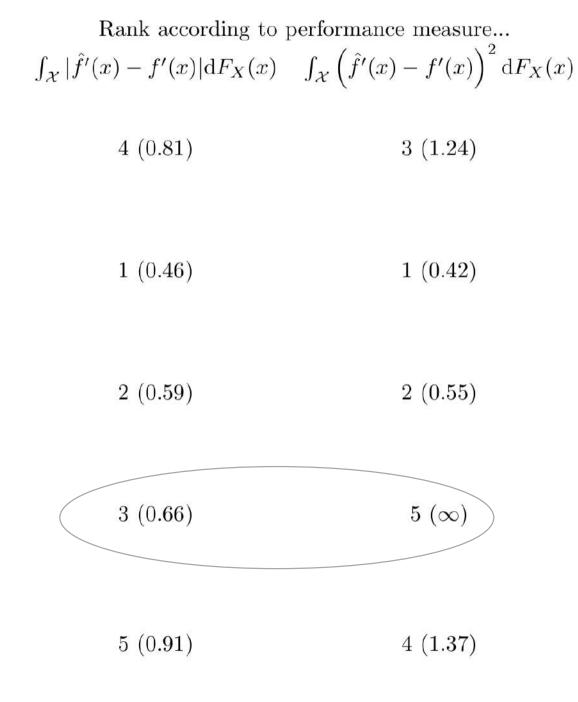


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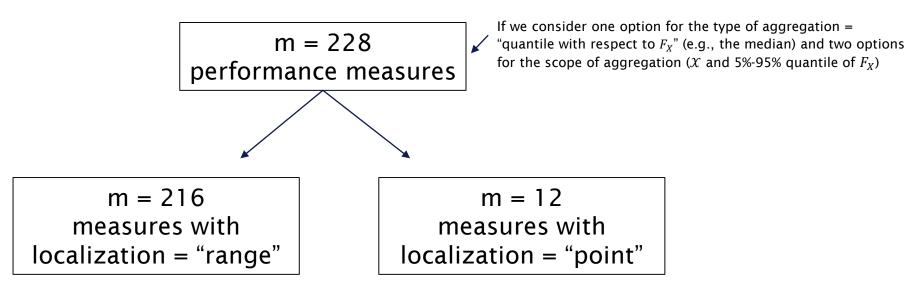


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## How many measures are there?

According to our categorization, there are...



 $\rightarrow$  How to choose a **smaller set** of performance measures for a simulation study?

→ Select those that capture different features (see examples!)



# Aggregation over simulated data sets

- Our performance measures will summarize the quality of the fitted line in 1 simulated data set
- The analyst still has to decide whether
  - Expected value of the performance measure
  - Variance of the performance measure
  - or other population quantity is of interest (e.g., median, *p*<sup>th</sup> quantile etc.)
- If there is a clear optimum value (e.g. expected difference [=bias] should be 0), one could also construct a combination of bias + variance
  - Obvious: MSE = bias<sup>2</sup> + variance



# Applications

- Univariate models: unadjusted association
- Models where the association of interest is adjusted for a (fixed) set of adjustment variables (descriptive-associational)
- Evaluation over a two-dimensional grid on X<sub>1</sub>, X<sub>2</sub>
- Prediction/calibration:
  - agreement of predicted and observed values
  - agreement of predicted and true linear predictor values
- Extensions: comparison to ,null' instead of true f(x)
  - Number of roots
  - General wiggliness



## Preprint is available on Arxiv



Ullmann, T., Heinze, G., Abrahamowicz, M., Perperoglou, A., Sauerbrei, W., Schmid, M., Dunkler, D., for TG2 of the Stratos initiative. (2025). A categorization of performance measures for estimated non-linear associations between an outcome and continuous predictors (Version 1). arXiv. https://doi.org/10.48550/ARXIV.2503.16981



### References

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Buchholz, A., Sauerbrei, W., & Royston, P. (2014). A measure for assessing functions of time-varying effects in survival analysis. Open Journal of Statistics, 4(11), 977998

Govindarajulu, U. S., Spiegelman, D., Thurston, S. W., Ganguli, B., & Eisen, E. A. (2007). Comparing smoothing techniques in Cox models for exposure-response relationships. Statistics in Medicine, 26(20), 3735-3752.

Morris, T.P., White, I.R., Crowther, M.J., 2019. Using simulation studies to evaluate statistical methods. Statistics in Medicine 38, 2074–2102. <u>https://doi.org/10.1002/sim.8086</u>

