



Tricks and tweaks for the global prediction of high radon levels and local pitfalls

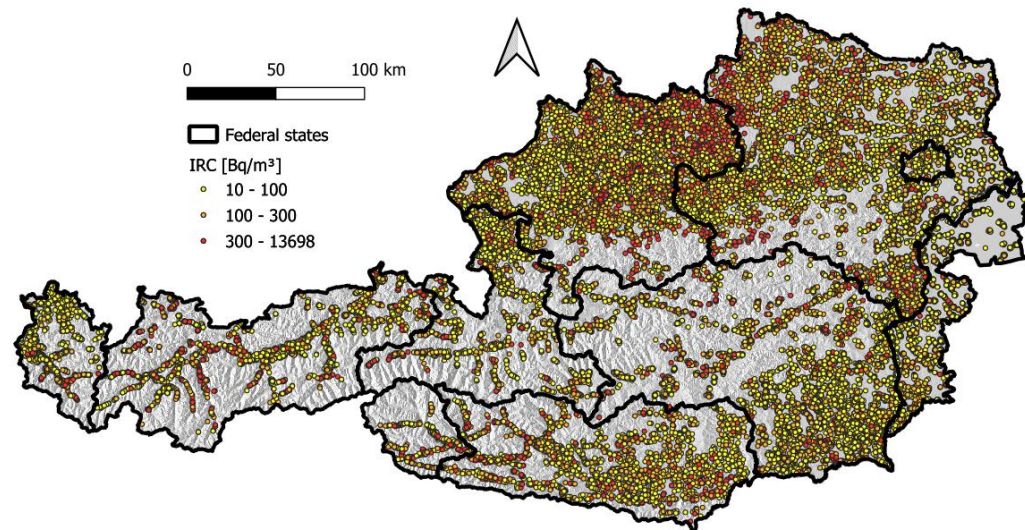
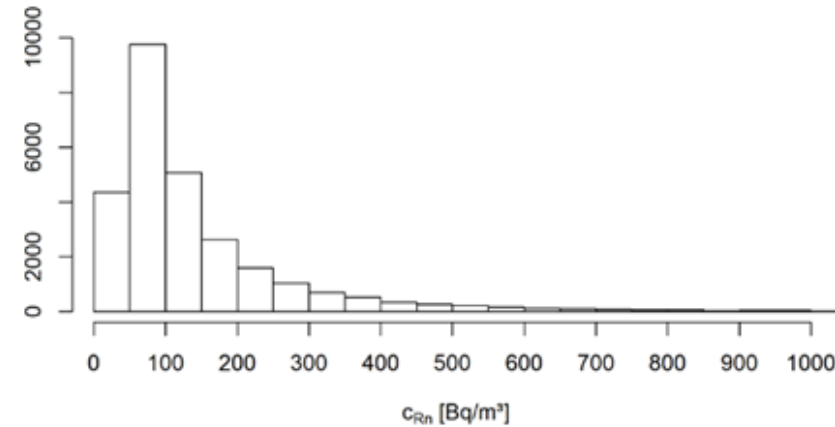
Sebastian Baumann, Sara Antignani, Joan Frédéric Rey,
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17.09.2025, GARRM, Prague



Tricks and tweaks for the global prediction of high radon levels and local pitfalls

- Concepts
- Setup
- Performance
- Conclusion
- Suggestions



- Radiation protection measures usually follow conservative approaches
- Consequences for the prediction of indoor radon concentrations?
- What are the aims of predicting IRC?
- Are high radon levels more relevant as low levels?
- Are there models and techniques better suitable to predict high IRC?






- In Rey et. Al (2024) we suggested modifications of the target (IRC) in a predictive model to focus on the prediction of high radon levels.
- Today some of these model modifications and their performance are presented





SYSTEMATIC REVIEW article

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This article is part of the Research Topic
Radon and Related Health Effects: From Exposure to Risk
Assessment and Policies
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Systematic review of statistical methods for the identification of buildings and areas with high radon levels

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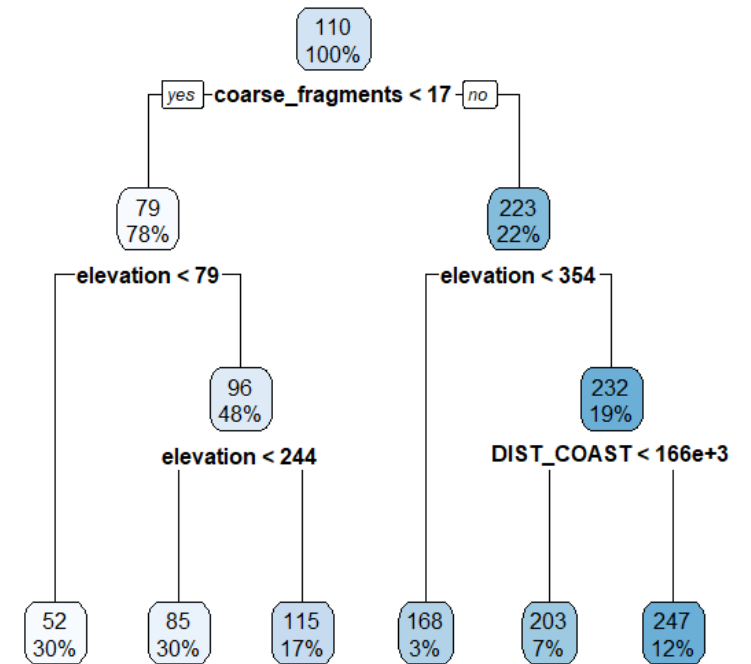
Radon is a natural and radioactive noble gas, which may accumulate indoors and cause lung cancers after long term-exposure. Being a decay product of Uranium 238, it originates from the ground and is spatially variable. Many environmental (i.e., geology, tectonic, soils) and architectural factors (i.e., building age, floor) influence its presence indoors, which make it difficult to predict. However, different methods have been developed and applied to identify radon prone areas and buildings. This paper presents the results of a systematic literature review of suitable statistical methods willing to identify buildings and areas where high indoor radon concentrations might be found. The application of these methods is particularly useful to improve the knowledge of the factors most likely to be connected to high radon concentrations. These types of methods are

● Target Variable ~ Model (Predictors)

IRC ~ Random Forest { Building Characteristics
Geogenic Factors
Climate

Possible Goals:

- Estimate the average IRC of a country
- Develop a radon map
- Estimate the amount of buildings with IRC greater 300 Bq/m³
- Evaluation of influencing factors

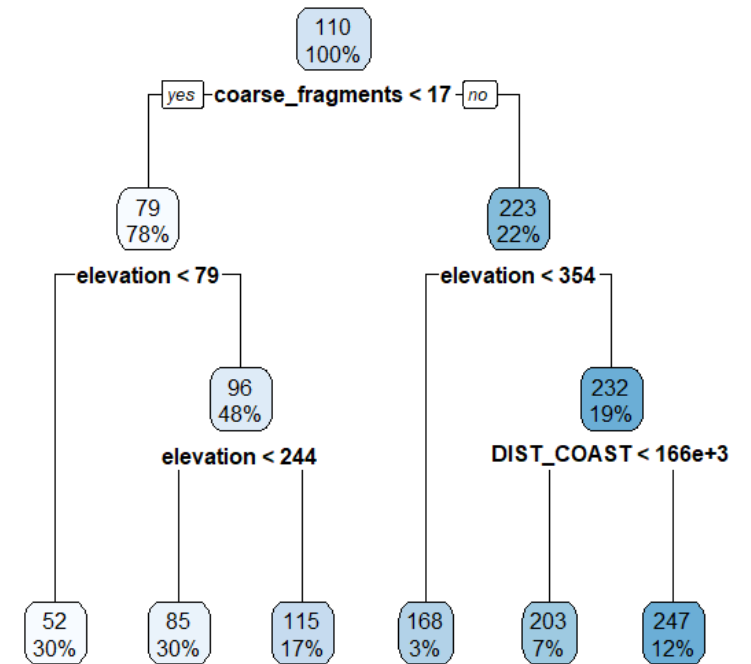


● Target Variable ~ Model (Predictors)

IRC ~ Random Forest { Building Characteristics
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Model characteristics

- ~ 50.000 indoor radon measurements
- Performance metrics on hold out random test set
- 20 Predictors from different categories: coordinates, floor level, geology, temperature, soil characteristics, building material ...



The model - Predicting high radon levels

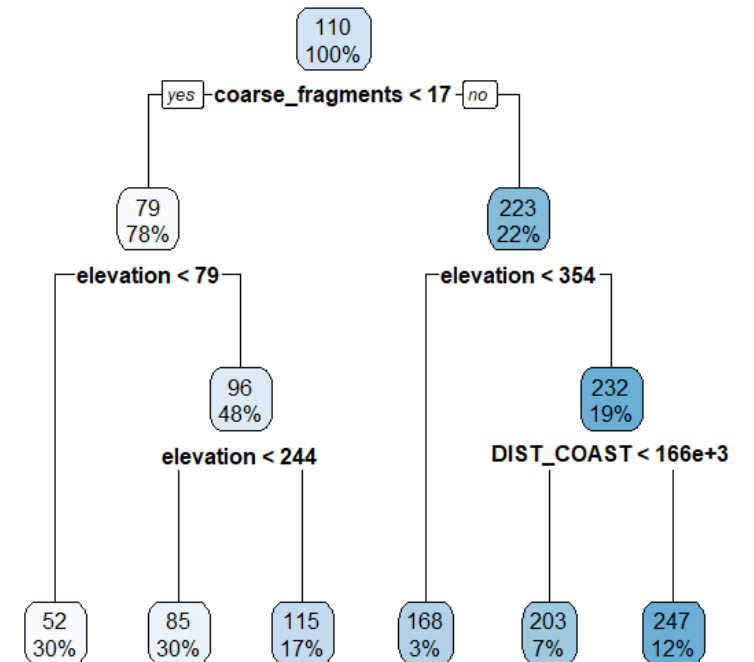
● Target Variable ~ Model (Predictors)

IRC ~ Random Forest

Building Characteristics
Geogenic Factors
Climate

Modifications:

- No modification
- Log-transforming
- Quantile regression forest
- Classification instead of regression



Performance – Example log-transformation

- No Modification of the target variable compared to log-transformation
- MAE = Mean absolute error of prediction and observation (the lower the better)

	Non modification	Log-transformed
Overall MAE Bq/m ³	90	79

Performance – Example log-transformation

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Transforming the target value:

- Usually, IRC follow a log-normal distribution and for prediction the IRC are often log-transformed

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	Non modification	Log-transformed
Overall MAE Bq/m^3	90	79
MAE IRC $> 300 \text{ Bq/m}^3$	333	398

Performance – Example log-transformation

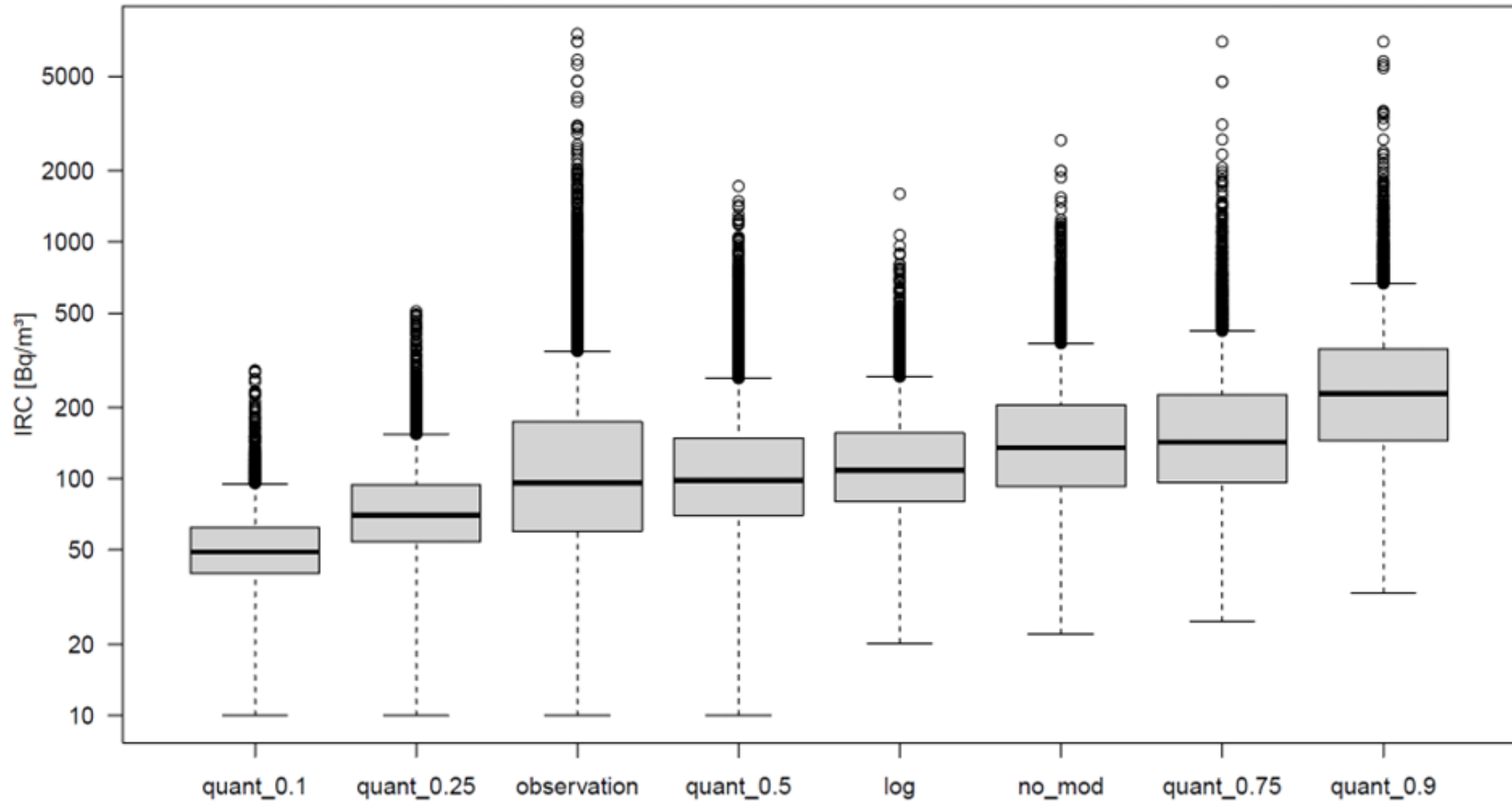
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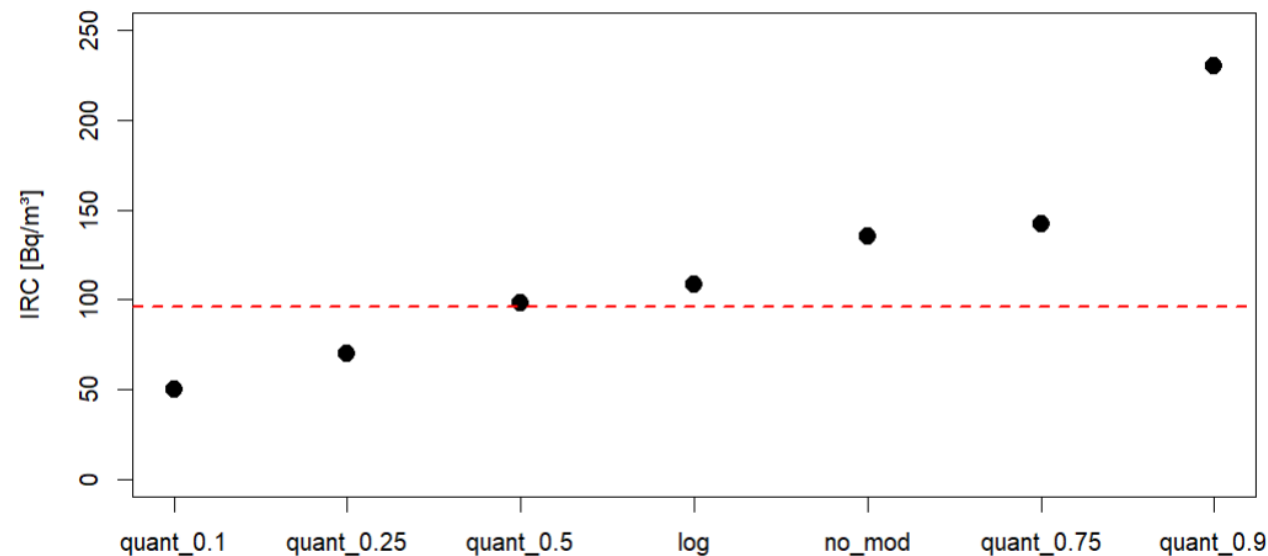
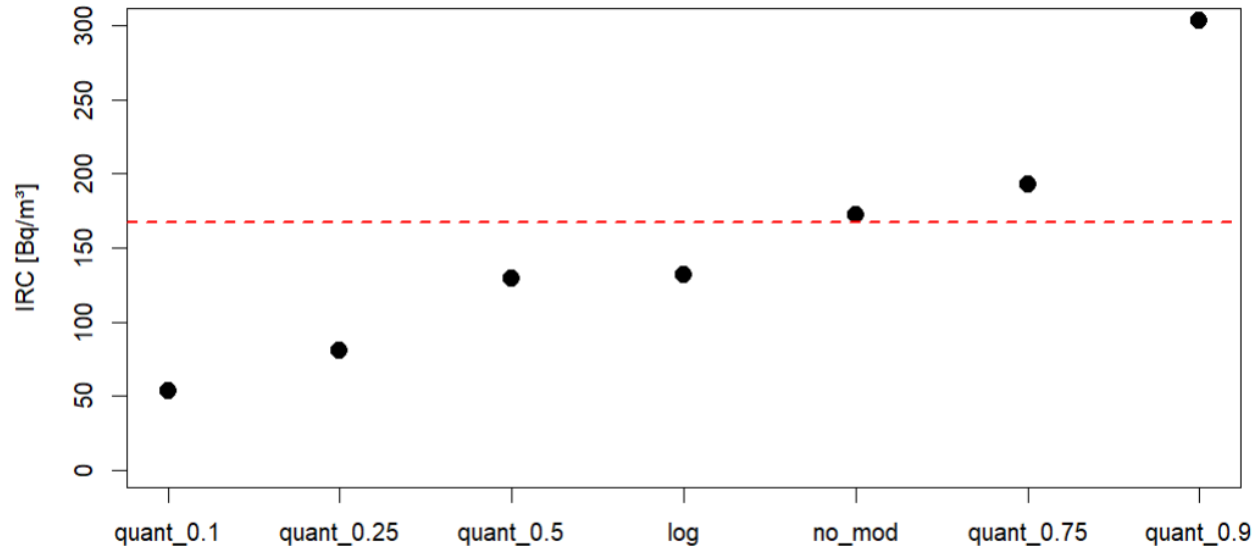
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Overall MAE Bq/m^3	90	79
MAE IRC $\geq 300 \text{ Bq/m}^3$	333	398
MAE IRC $< 300 \text{ Bq/m}^3$	58	37

Performance - Distributions of predictions



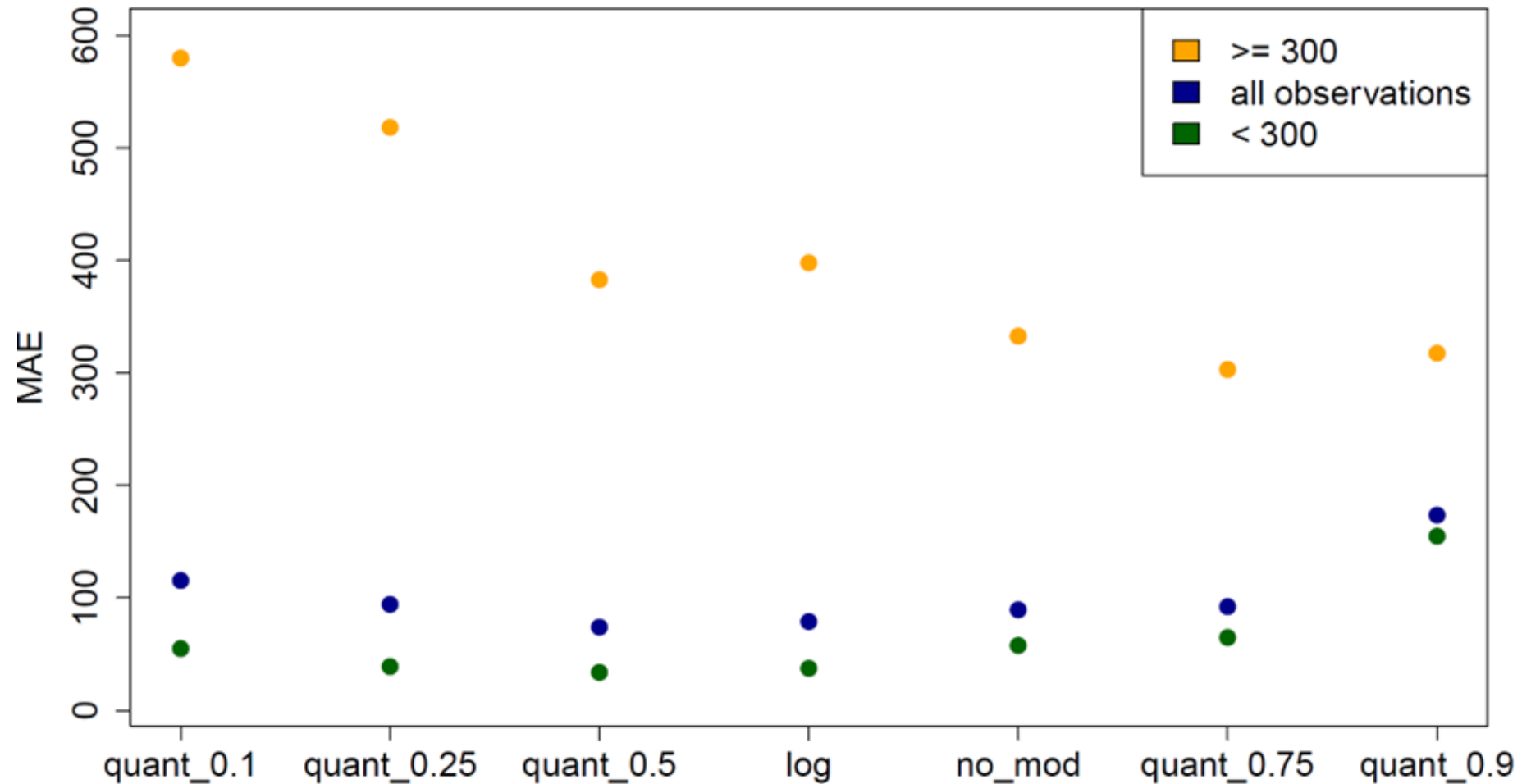
Performance – Aggregates

Arithmetic means

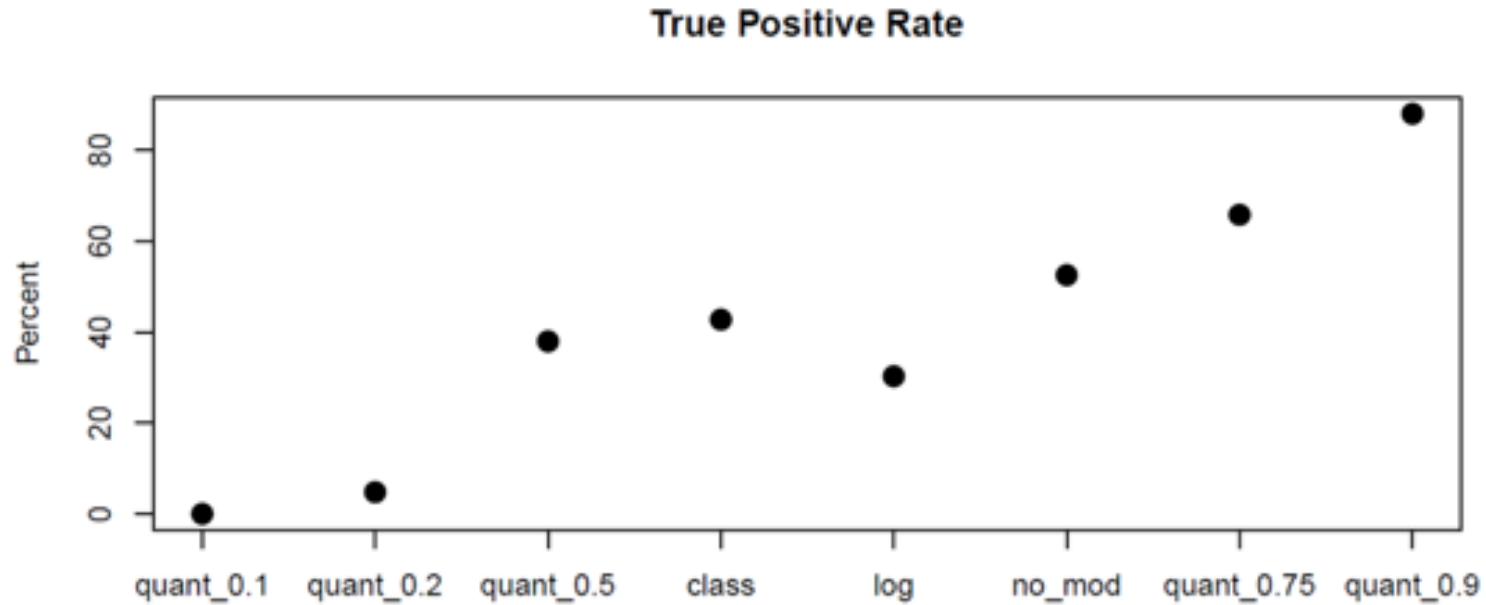


- Arithmetic means of the model predictions, compared to the observation
- Medians of the model predictions compared to the observation

Performance - MAE for different models



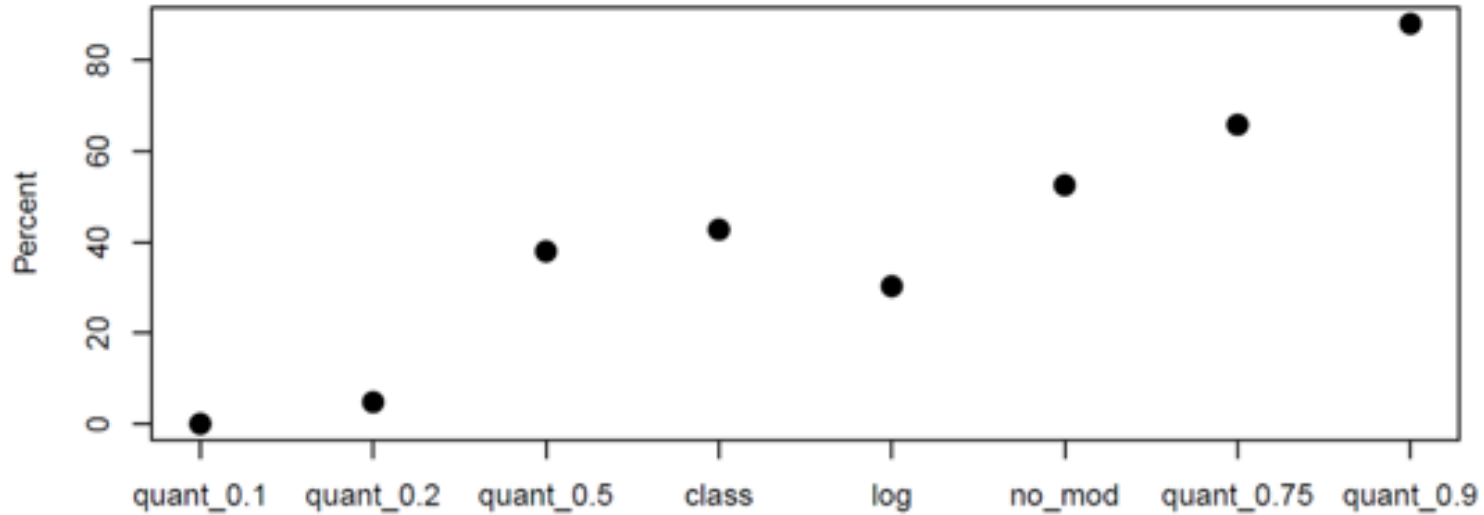
Performance – TPR | TNR



- TPR: how many observations $\geq 300 \text{ Bq/m}^3$ were classified correctly [%]

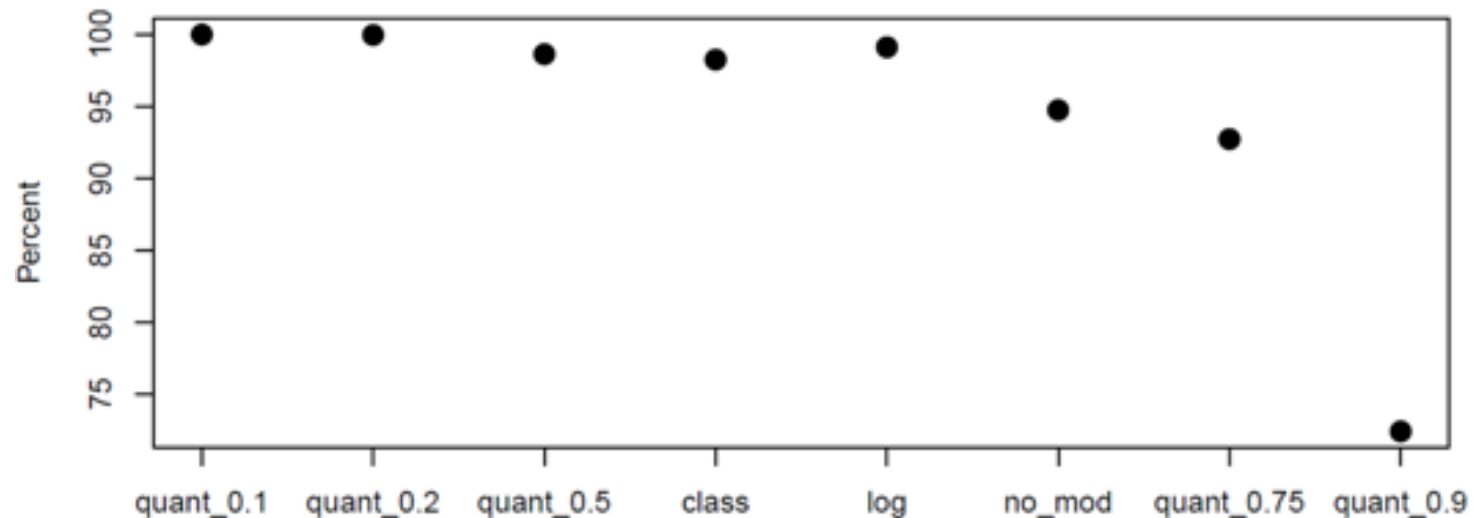
Performance – TPR | TNR

True Positive Rate



- TPR: how many observations $\geq 300 \text{ Bq/m}^3$ were classified correctly [%]

True Negative Rate



- TNR: How many observations $< 300 \text{ Bq/m}^3$ were classified correctly [%]

Conclusions

- Transforming the target (IRC) in a predictive model has consequences, especially if the target follows a log-normal distribution.
- This is seldom discussed when using predictive models
- There may be aims where the overall performance might be less relevant, than the performance for certain IRC – Populations (e.g. $> 300 \text{ Bq/m}^3$)
- What is the goal of my prediction?

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Overall MAE Bq/m^3	90	79
MAE IRC $\geq 300 \text{ Bq/m}^3$	333	398
MAE IRC $< 300 \text{ Bq/m}^3$	58	37
TPR %	53	30
TNR %	94	99
MEAN DIFF $[\text{Bq/m}^3]$	+ 6	- 35
MEDIAN DIFF $[\text{Bq/m}^3]$	+ 39	+ 12

Suggestions

- There is no jack of all trades
- Use well working models (tree based) and focus on overall workflow:
 - DATA handling
 - Feature selection
 - Train / Test split (**Generalization!**)
 - Model evaluation
- Focus on the research question: What do I want to accomplish with my prediction? (Map, Mean, Distribution)



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Thank you!

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