auditor: an R package and methodology for validation of any statistical model

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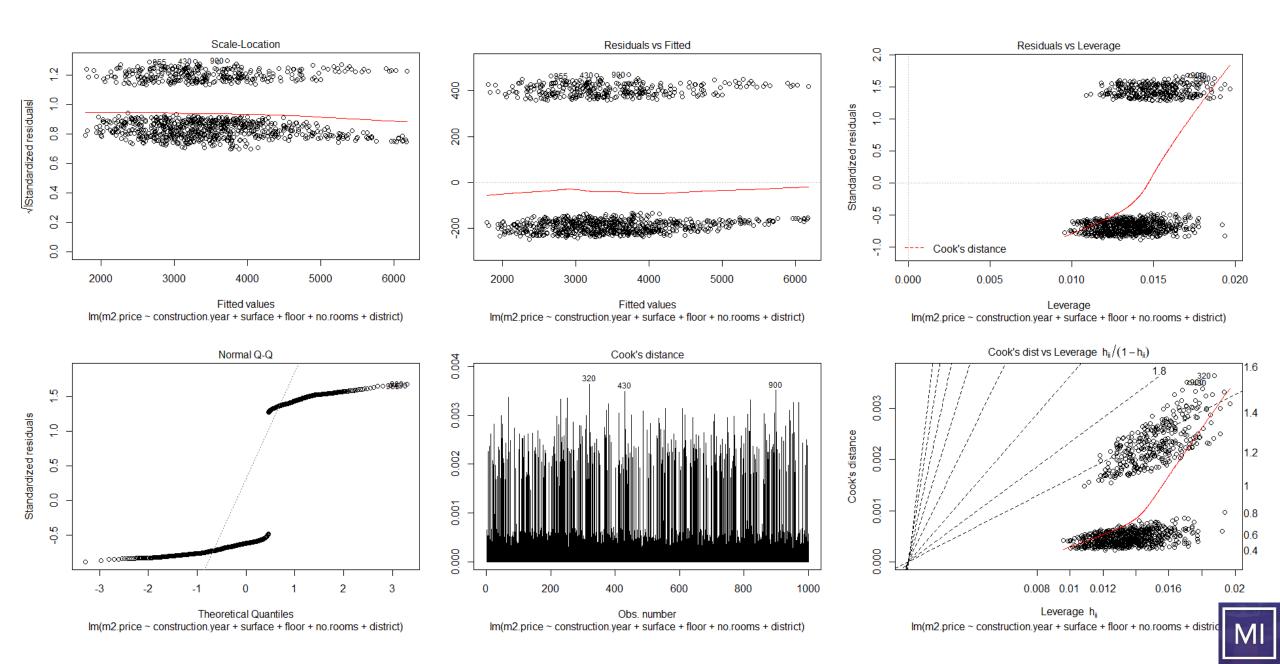


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Motivation – linear models



auditor



- Model-agnostic
- Model comparisons
- Consistency

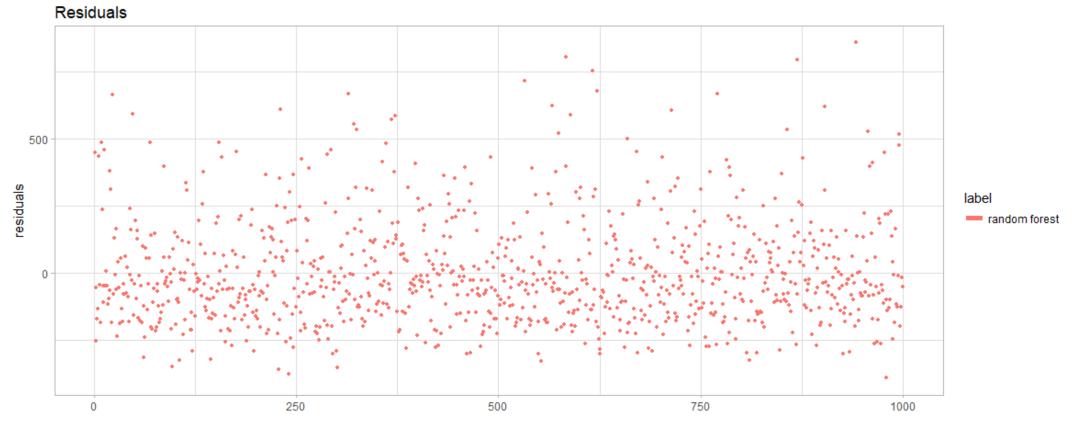
model %>% audit() %>% plot(type =)



model %>% audit() %>% plot(type =)



```
library(auditor)
apartments_rf_model <- randomForest(m2.price ~ ., data = apartments)
au_rf <- audit(apartments_rf_model, data = apartmentsTets, y = apartmentsTest$m2.price)
plot(au_rf, type = "Residual")</pre>
```

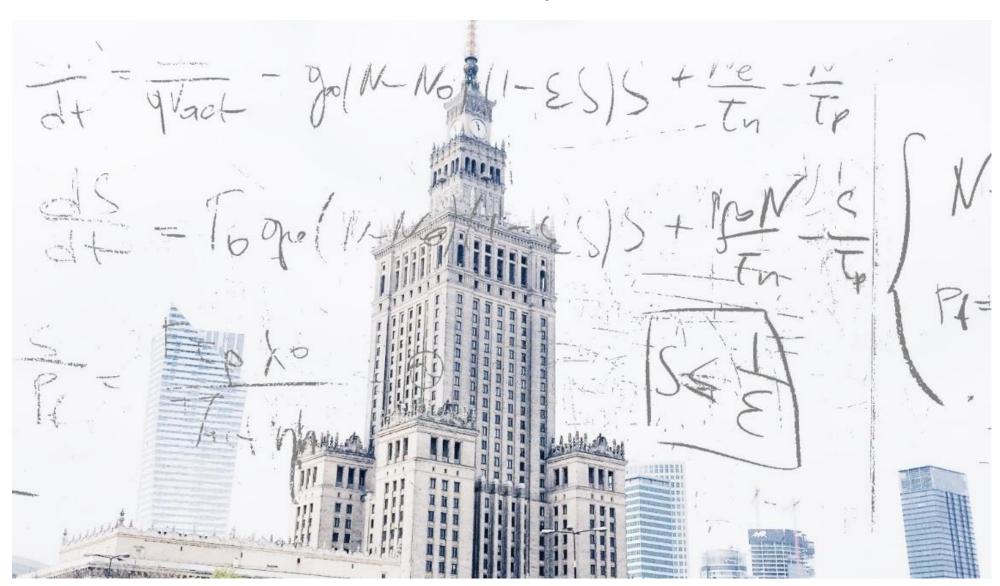




Do we really need model audit?

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A story about two models!





Do we really need model audit?



A story about two models!

m2.price	construction.year	surface	floor	no.rooms	district
5897	1953	25	3	1	Srodmiescie
1818	1992	143	9	5	Bielany
3643	1937	56	1	2	Praga
3517	1995	93	7	3	Ochota
3013	1992	144	6	5	Mokotow

Do we really need model audit?



A story about two models!

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Accuracy is not enough!



Root Mean Square error

```
predicted_mi2_lm <- predict(apartments_lm_model, apartmentsTest)
sqrt(mean((predicted_mi2_lm - apartmentsTest$m2.price)^2))</pre>
```

[1] 283.0865

```
predicted_mi2_rf <- predict(apartments_rf_model, apartmentsTest)
sqrt(mean((predicted_mi2_rf - apartmentsTest$m2.price)^2))</pre>
```

[1] 283.1138



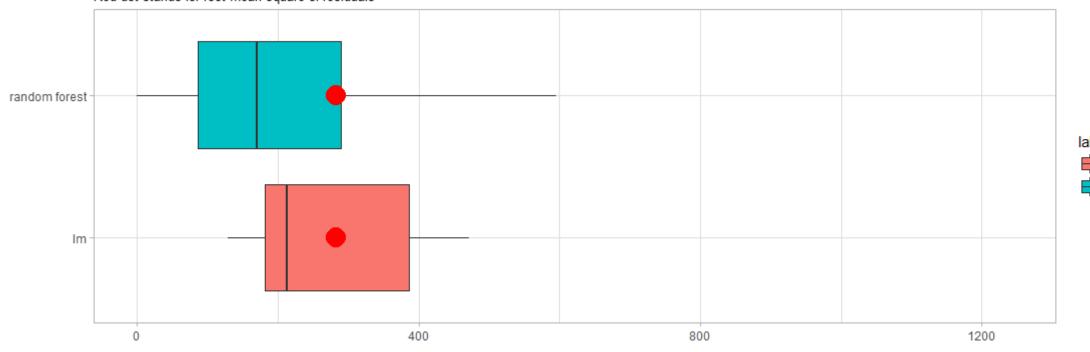
Model Performance



```
au_rf <- audit(apartments_rf_model, data = apartments, y = apartments$m2.price)
au_lm <- audit(apartments_lm_model, data = apartments, y = apartments$m2.price)
plot(au_rf, au_lm, type = "ResidualBoxplot")</pre>
```

Boxplots of | residuals |

Red dot stands for root mean square of residuals



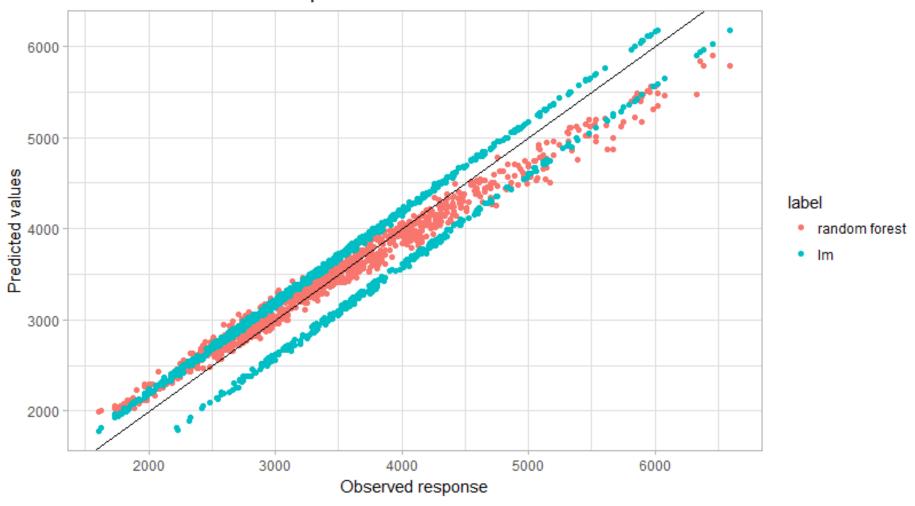




plot(au_rf, au_lm, type = "Performance")

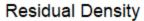


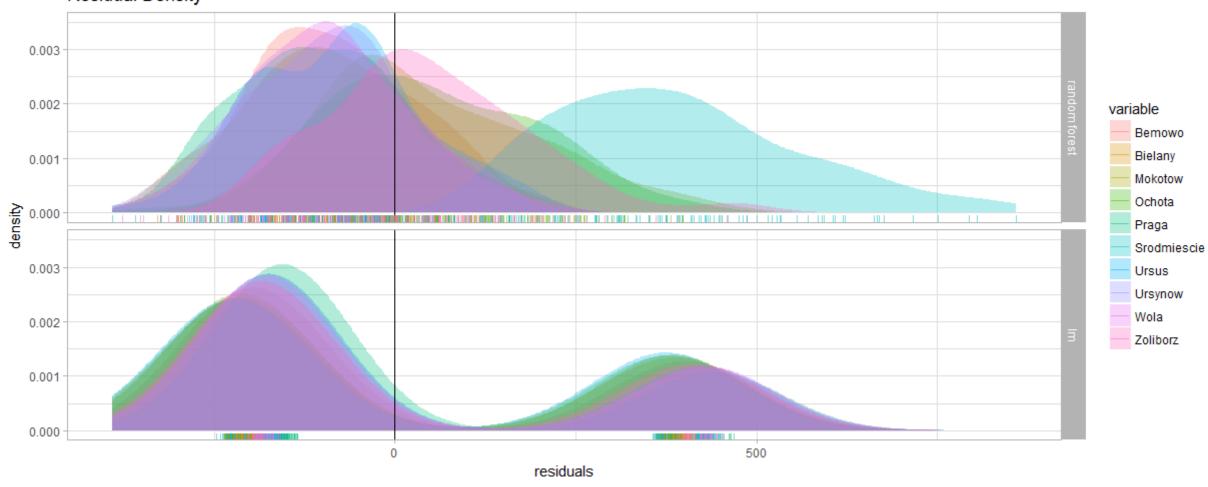
Predicted vs Observed response



plot(au_rf, au_lm, type = "ResidualDensity", variable = "district")





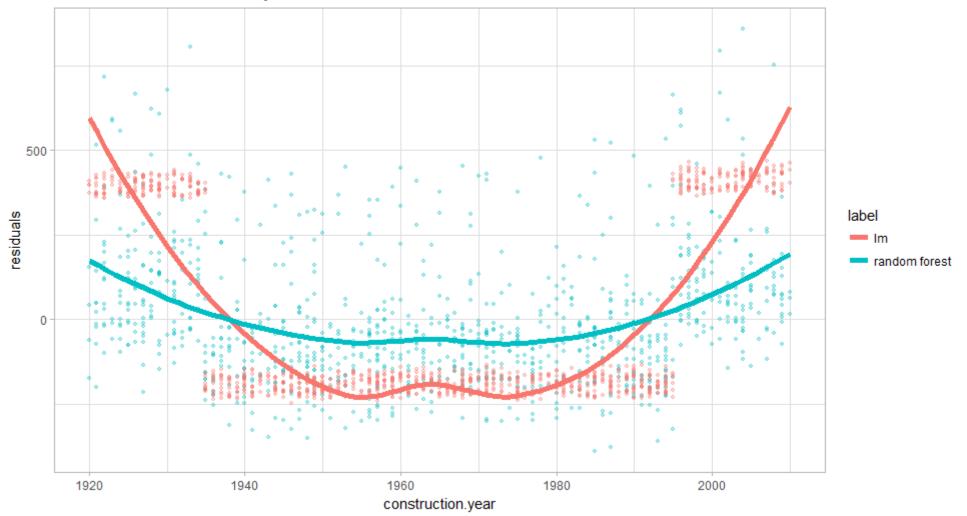




plot(au_rf, au_lm, type = "Residual", variable = "construction.year")



Residuals vs construction.year





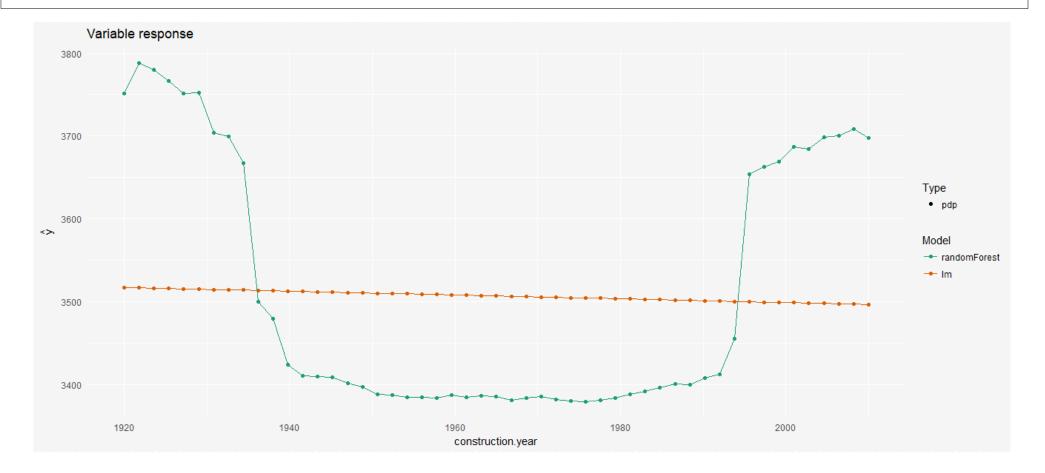
Partial Dependence Plot

```
dalex
```

```
library(DALEX)
```

```
exp_rf <- explain(audit(apartments_rf_model, data = apartments, y = apartments$m2.price)
exp_lm <- explain(audit(apartments_lm_model, data = apartments, y = apartments$m2.price)</pre>
```

plot_model_performance(exp_rf, exp_lm, type = "pdp", variable = "construction.year")





Let's sum up!

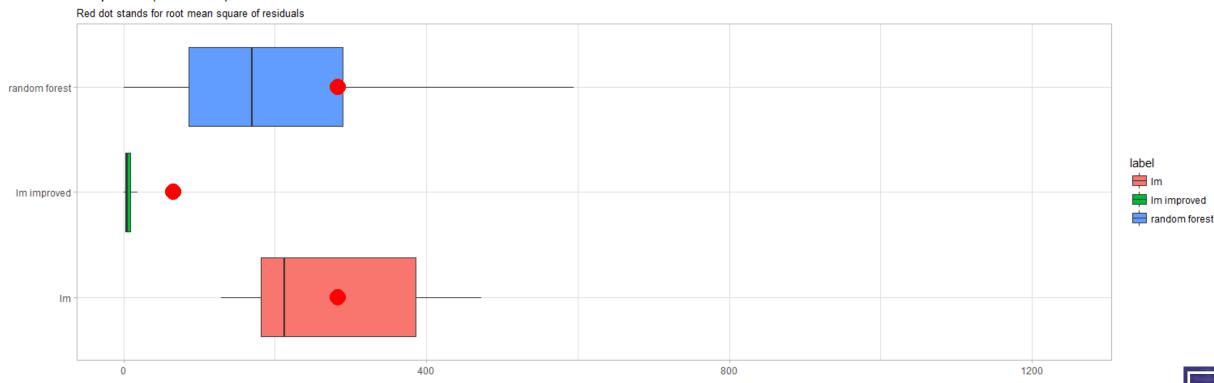


- Both models have similar performance.
- Random forest model has smaller residuals than the linear model but there is a small fraction of very large residuals.
- Random forest model under-predicts expensive apartments. It is not a model that we would like to employ.
- The relation between construction year and the price of square meter is non linear.



Let's improve our model!

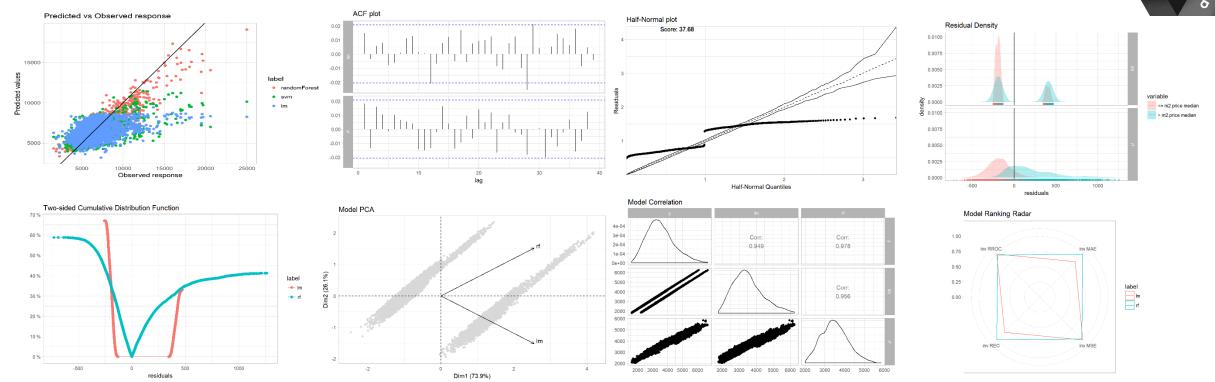
Boxplots of | residuals |





What's more?





GitHub

https://github.com/mi2-warsaw/auditor



Thank you!

auditor

GitHub

https://github.com/mi2-warsaw/auditor

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