Quick view the Data

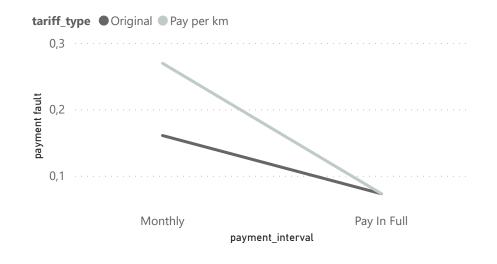
(I) Quickview on Data

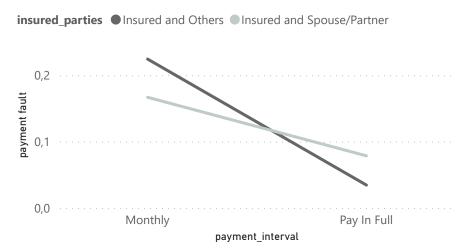
- Get a first idea, which fields might have a an impact on payment faults
- Find areas with weak exposure and prepare binning and grouping values
- Check for interactions between fields

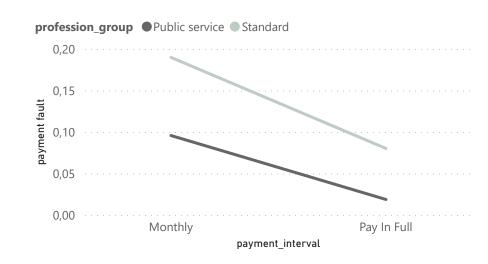


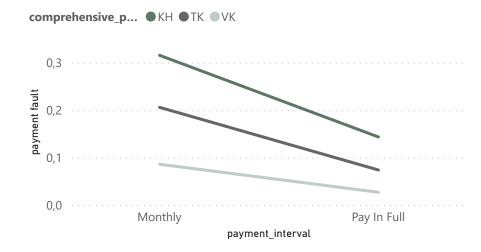
Test Data for Interactions

Possible interactions with Payment type

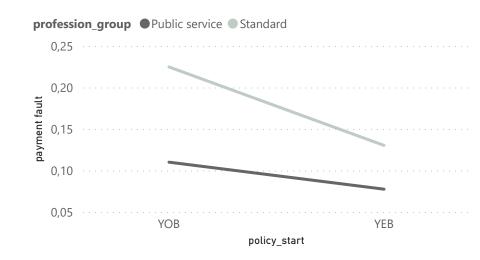




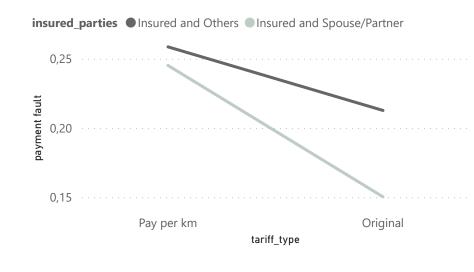




Possible interactions with policy start



Possible interactions with tariff type



Summary: quick view the Data

(I) Pricing:

As expected, payment faults seem to correlate with a higher price tag. This shows by a high correlation between fields that usually have a greater impact in pricing models.

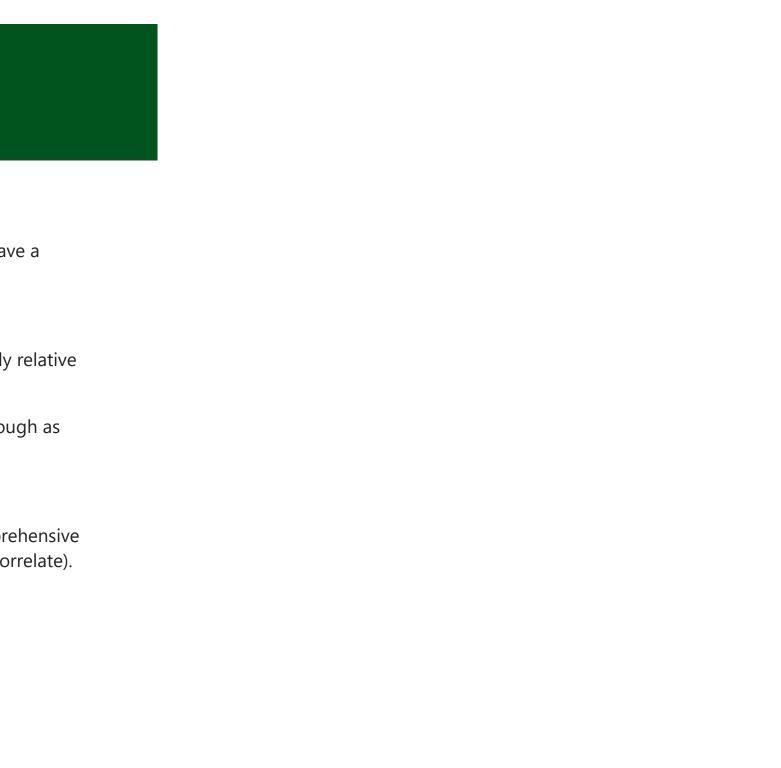
(II) Customer type

But there is another component, the type of customer. It seems, that the profession, the number of installments as well as the (usually relative expensive) full comprehensive coverage does have a positive impact on payment faults.

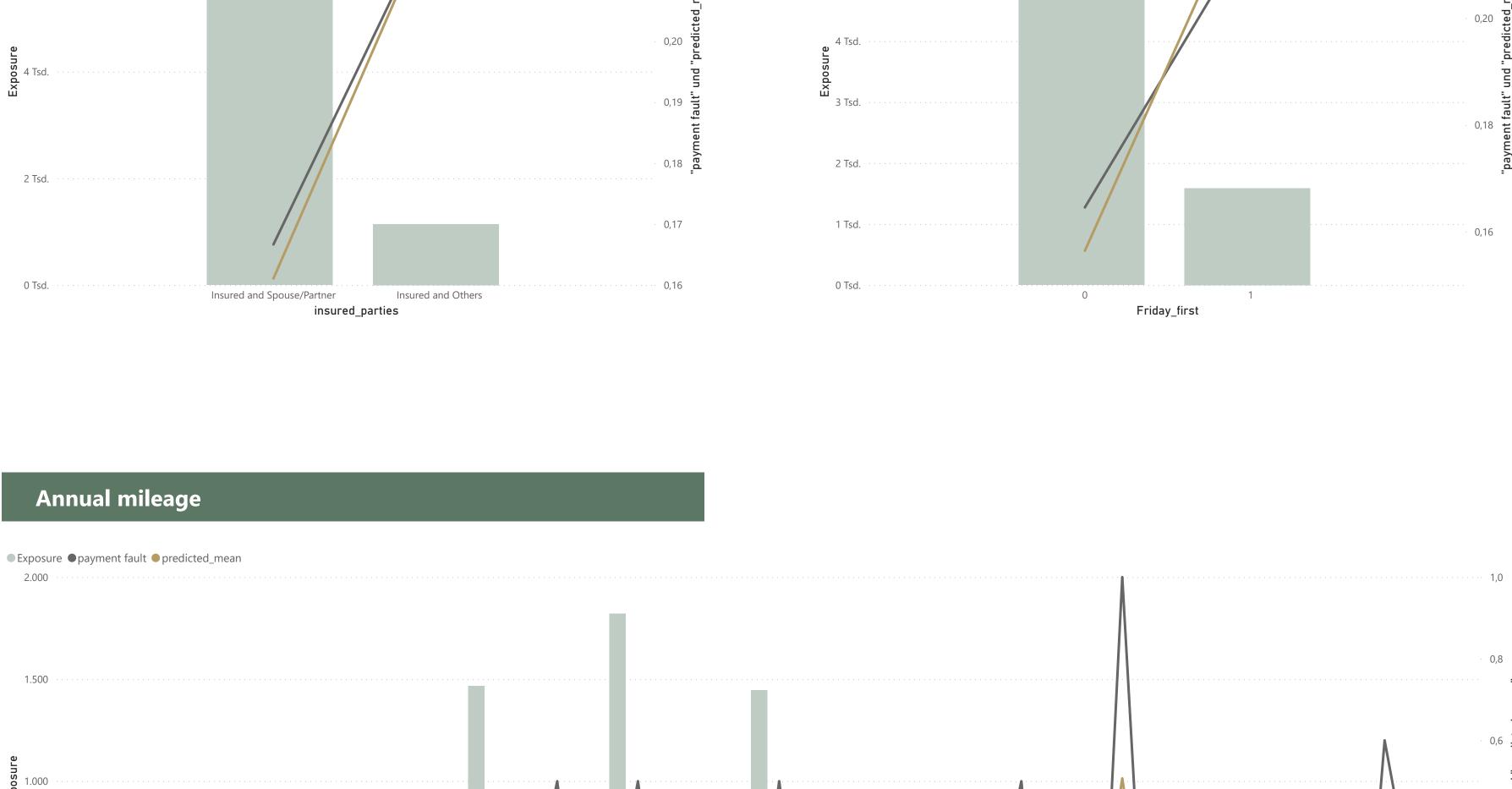
This makes sence, because price sensitive customers tend not to buy such a coverage (porbably because the car is not expensive enough as well).

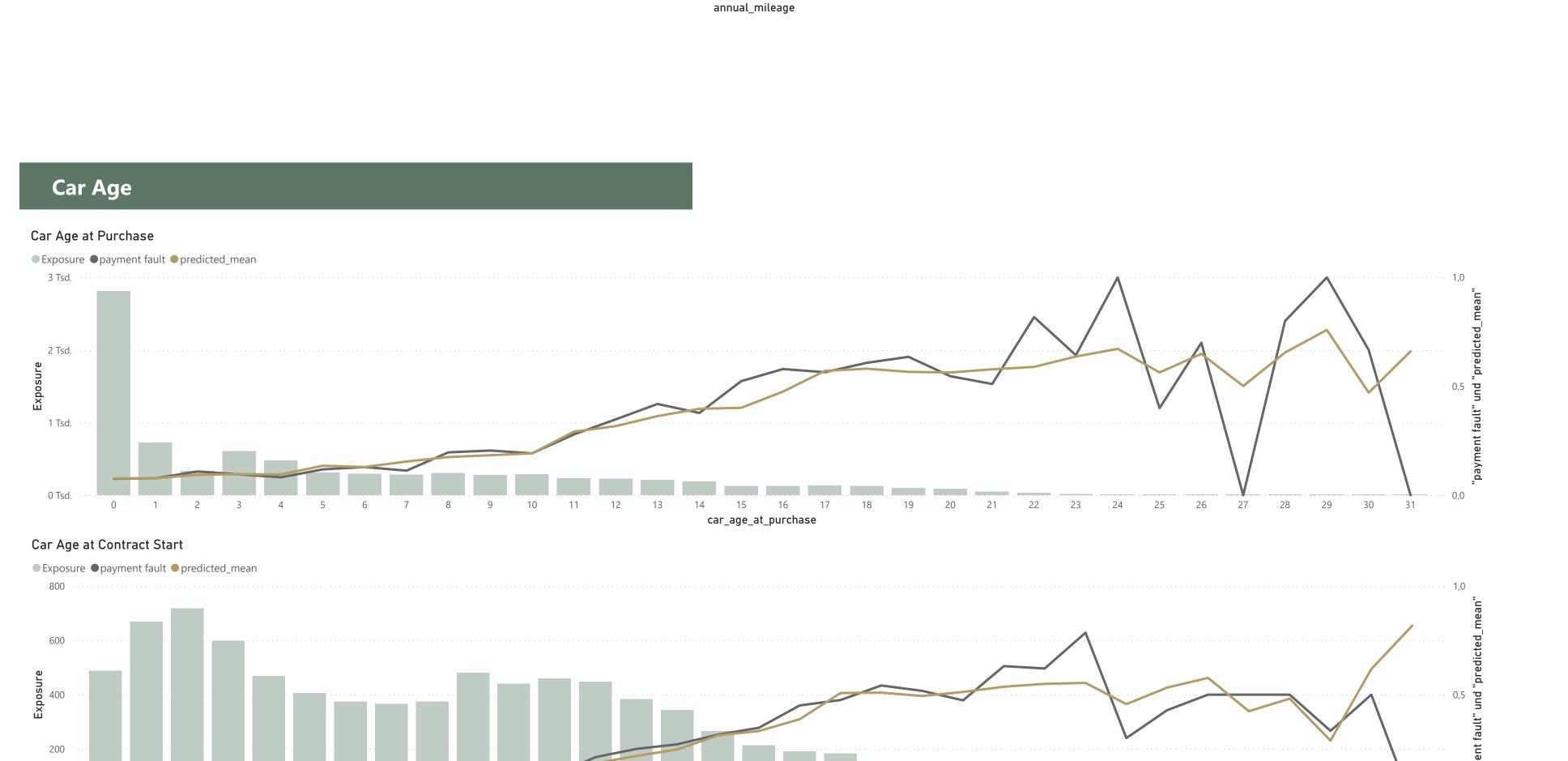
Notes:

the given regional classes are for full comprehensive cover and typically do not affect the pricing of a third liability or partially comprehensive covers. So therefore the effect would be stronger if the regional classes for third liability would have been given (even though they correlate).



Evaluating fit on test data only Age of insured person ■ Exposure ■ payment fault ■ predicted_mean 250 age_insured_person SF-Class third party liability ■ Exposure ■ payment fault ■ predicted_mean 800 SF32 SF33 SF34 SF35 SF36 SF37 SF38 SF39 SF40 SF41 SF42 SF43 SF44 bonus_malus_class_liability SF-Class full comprehensive cover ■ Exposure ■ payment fault ■ predicted_mean 0,5 0,4 bonus_malus_class_comprehensive **Profession group Payment** ■ Exposure ■ payment fault ■ predicted_mean ■ Exposure ■ payment fault ■ predicted_mean 0,18 0,18 0,16 6 Tsd. "payment fault" und "predicted_mean" 2 Tsd. 0,08 0 Tsd. Standard Public service Pay In Full profession_group payment_interval Tarif type **Policy start** ■ Exposure ■ payment fault ■ predicted_mean ■ Exposure ■ payment fault ■ predicted_mean 0,24 2 Tsd. 0,16 0 Tsd. 0 Tsd. Original Pay per km policy_start tariff_type Type of insurence Cover ■ Exposure ■ payment fault ■ predicted_mean ■ Exposure ■ payment fault ■ predicted_mean 0,30 0,5 4 Tsd. 0,25 l "predicted_mean" 3 Tsd. 4 Tsd. 0,20 no.15 no.15 no.15 no.15 no.15 no.15 2 Tsd. 0,2 1 Tsd. 1 Tsd. 0,10 0 Tsd. 0 Tsd. Change of Insurer New Vehicle First Vehicle Cover type_of_insurance **Deductibles** Deductable Partially Comprehensive (VK) Deductable Full Comprehensive (VK) Deductable Partially Comprehensive (TK) ■ Exposure ■ payment fault ■ predicted_mean ■ Exposure ■ payment fault ■ predicted_mean ■ Exposure ■ payment fault ■ predicted_mean 1.500 0,22 0,09 3.000 0,08 3 Tsd. "payment fault" und "predicted_mean" 2.500 2.000 Exbosine 1.500 1.000 1 Tsd. 0,02 500 0,14 0,05 0 Tsd. 0,00 deductible_partially_comprehensive deductible_fully_comprehensive deductible_partially_comprehensive Regional class (FC) ■ Exposure ■ payment fault ■ predicted_mean 0,22 0,21 0,20 **...** 800 200 0,15 risk_predictor_zip_code Friday first insurer (car age purchase = contract start) **Insured parties** ■ Exposure ■ payment fault ■ predicted_mean ■ Exposure ■ payment fault ■ predicted_mean 0,22 0,22 6 Tsd.





car_age_contract_start

Next Steps:

- (I) Checking, that all used Features are actually available at the point of the application
- (II) Extracting the factors and bounds for each field to generate a rating matrix
- (III) Testing the rating matrix for holes