

Understanding Variable Interactions for Better Scorecards

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Some Notes on Interactions

- An “interaction” is when the effect of one variable on a response depends on the level of a second variable
- Example: effect on health of (alcohol) + effect of (drug) \neq effect of (alcohol + drug)
- Example: effect on creditworthiness associated with one 90-day delinquency is larger when income is high (let’s say)

Interactions Are Often Ignored

- In scorecard models, interactions are often ignored. The number of points awarded for one level of one variable is constant, regardless of the values of any other variables
- Experience suggests that a small number of well-chosen interactions can improve model fit substantially

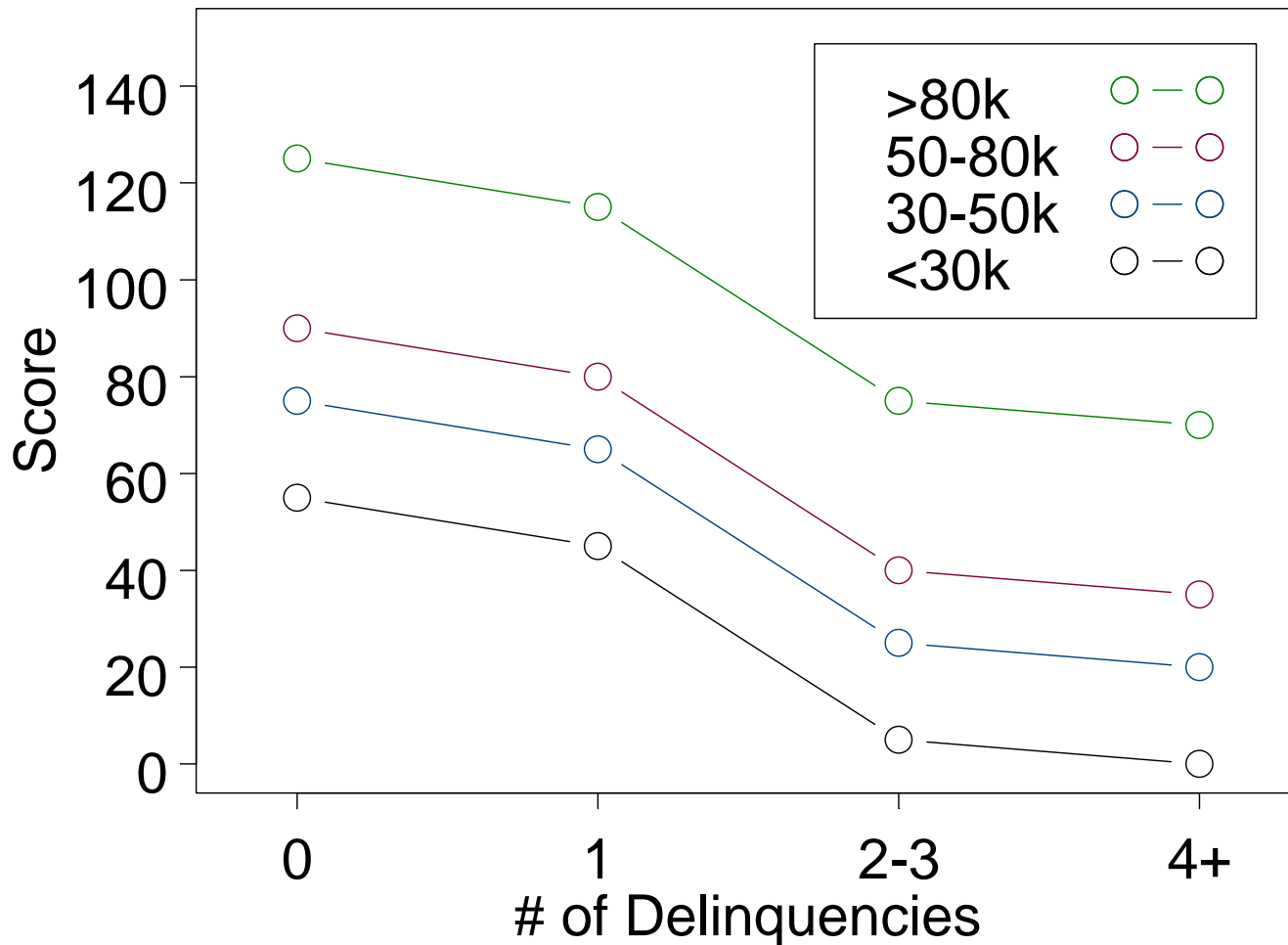
Modeling Interaction

- Logistic regression (widely used, interpretable; interactions must be specified explicitly)
- Classification trees (interactions handled automatically; less statistical theory)
- Neural networks (interactions handled implicitly; no interpretation)

Scorecard With No Interaction

- Household Income Delinquent TLs
 $\leq 29,999$: 0 Pts 0: +55 Pts
 30k-49,999: 20 Pts 1: +45 Pts
 50k-79,999: 35 Pts 2-3: + 5 Pts
 $\geq 80,000$: 70 Pts ≥ 4 : 0 Pts
- Going from the “0 delq” group to the “1 delq” group costs you 10 points, regardless of income

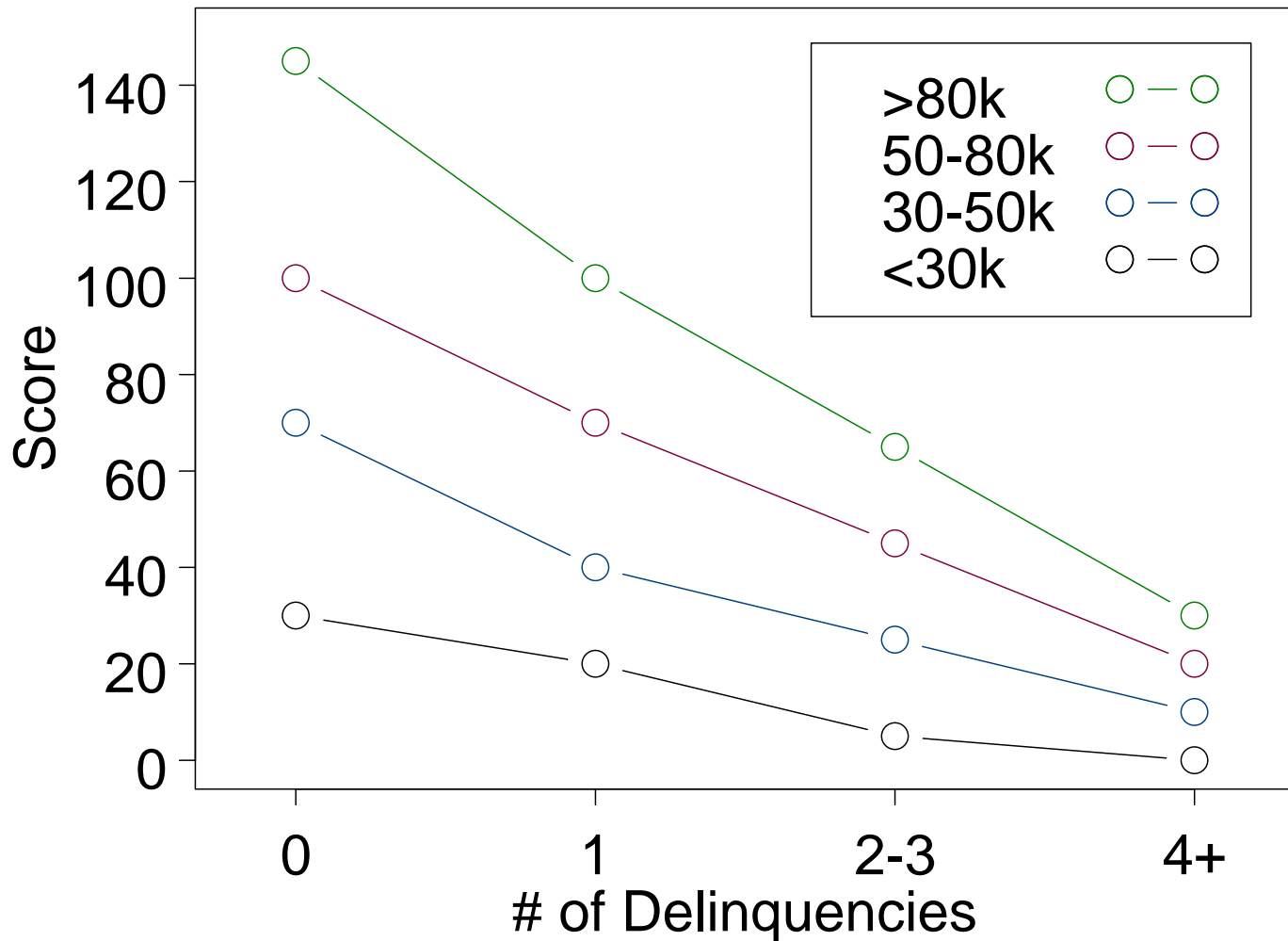
Scorecard With No Interaction



Scorecard With Interaction

Points	0 delq	1 delq	2-3 delq	≥ 4 delq
< 30k	30	20	5	0
30-50 k	70	40	25	10
50-80k	100	70	45	20
$\geq 80k$	145	100	65	30

Scorecard With Interaction



Where Can This Be Used?

- Anywhere interaction exists and isn't currently being modeled
- Potentially large gains in predictive power, in exchange for some loss in interpretability
- Logistic regression techniques and theory continue to apply; other models also available

Segmentation

- One obvious use: segmenting populations into scorecard segments
- We might expect the effect of some variables to be different across segments
- That is, we expect an interaction between that variable and segment

Classification Trees

- This is a statistical model that identifies interactions
- Particularly useful for segmenting population into smaller, more homogeneous groups
- Start with observations in one “box”
- Split into two pieces so as to maximize homogeneity of outcome

Automatic Interaction Detection

- The “Trades < 4 ” group is divided by “EverDQ” (Yes or No), but the “Trades ≤ 4 ” group is divided by “#30+ < 2 ” (Yes or No)
- That is, the effect of “EverDQ” is bigger than “#30 < 2 ” for “Trades < 4 ,” but smaller for “Trades ≥ 4 ”
- This is an interaction!

Conclusion

- Interactions exist in real data, so we should include them in the model
- They show how the effect of one variable on the response can depend on the value of another
- Useful for segmentation, potentially for scorecard construction