



# SAS® Club

Der Business Analytics Club für SAS User  
27. November 2019 | Wien

# Agenda

15:30 - 16:00 Uhr	Registrierung
16:00 - 16:15 Uhr	<b>Begrüßung &amp; Einleitung</b> Gerhard Svolba, SAS
16:15 - 16:45 Uhr	<b>Die Kommunikation von analytischen Ergebnissen und die Interpretation von ML-Modellen mit SAS Viya - 5 Tipps und Tricks, die Ihr Leben als Data Scientist vereinfachen</b> Gerhard Svolba, SAS
16:45 - 17:10 Uhr	<b>SAS in der Cloud</b> Phillip Manschek, SAS
17:10 - 17:40 Uhr	<b>Simplification of Routine Data Analyses &amp; Tasks Using SAS</b> Christoph Gruber, Novartis / EBEWE Pharma Ges.m.b.H.
17:40 - 18:00 Uhr	<b>Die Neuerungen im SAS Enterprise Guide 8.1</b> Christian Url, mayato AT GmbH
ab 18:00 Uhr	Gemütliches Get2Gether

- Downloads: <https://github.com/gerhard1050/DACH-SASUserGroups>

# Communicating Analytical Results and Interpreting your ML Models with SAS Viya – 5 Tips and Tricks that will make your life as data scientist easier

Gerhard Svolba

Analytic Solutions Architect

SAS Austria

Credits for Input to:  
Martin Schütz, Tamara Fischer

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<https://www.linkedin.com/in/gerhardsvolba/>

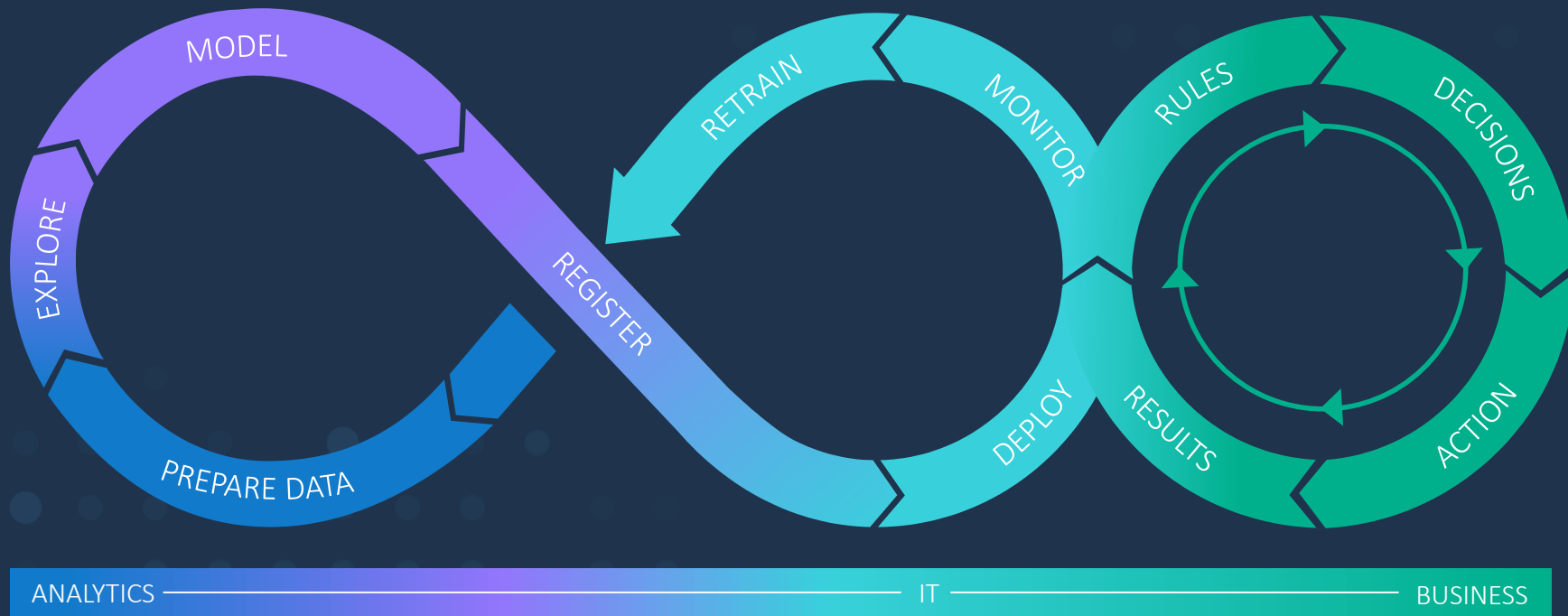


**sas**  
THE POWER TO KNOW®

# We (data scientists) want to communicate our results

- Acceptance of our results
- Better understanding – better usage in the business process
- Less „last minute“ misunderstandings

# THE SAS DECISIONING PROCESS

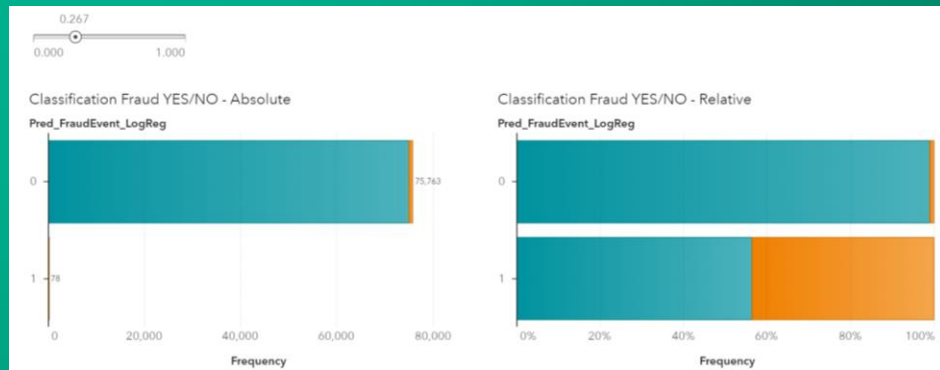


# 5 Tips (featuring SAS Visual Analytics, SAS Model Studio and SAS Coding)

1. Perform interactive cutoff analysis
2. Quantify the importance of explanatory variables
3. Turn on the model interpretability charts
4. Use a decision tree to „explain“
5. Display the (hidden) regression coefficient

# Tip #1:

Perform interactive cutoff analysis to illustrate the consequences on the Good/Bad classification

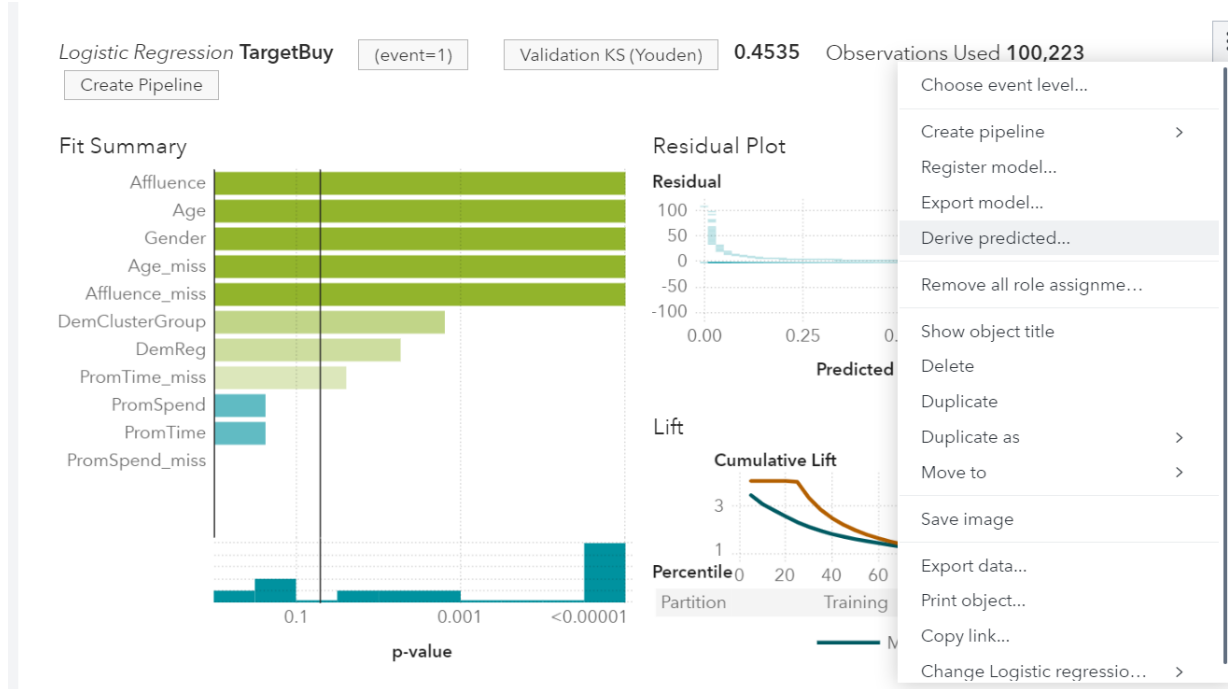


# Illustrate the outcome (deliverable) of a predictive model

- A predictive model
- creates predictions.
- (In case of a binary classification task it outputs the probability the that event takes place.)
- You want to show this!
- And illustrate the consequences of different cutoff values for the business decision.



# Select „Derive Predicted“ in a predictive model created with SAS Visual Analytics



# Name your output variables

×

New Prediction Items

2 new items will be created: 1 predicted value and 1 probability value.

Select the items you want to show in the Data pane.

Predicted values:

☒

Probability values:

☒

# You receive new variables in the data

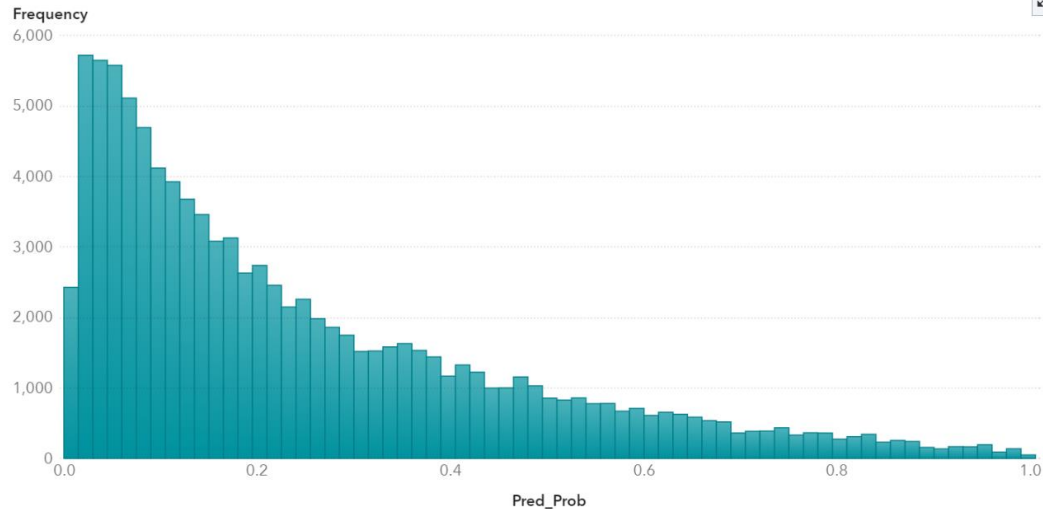
▼ Derived (Logistic Regression: Octob...

☐ Pred\_FraudEvent\_LogReg

☐ Pred\_Prob

☒ Prediction cutoff for Pred\_Fraud...

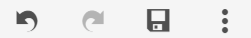
Frequency of Pred\_Prob



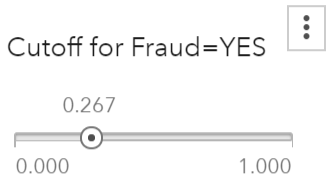
# Also allows you to interactively „play“ with the cutoff point

- Important to illustrate the outcome of a predictive model
- What are the consequences of a certain cutoff point on
  - Number of customers, transactions flagged with YES
  - Expected false positives, false negatives, ...

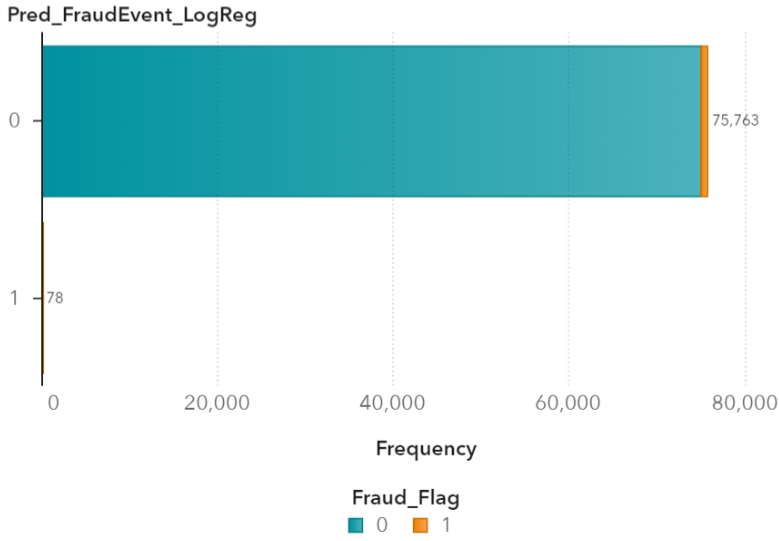
# Fraud\_VA\_01



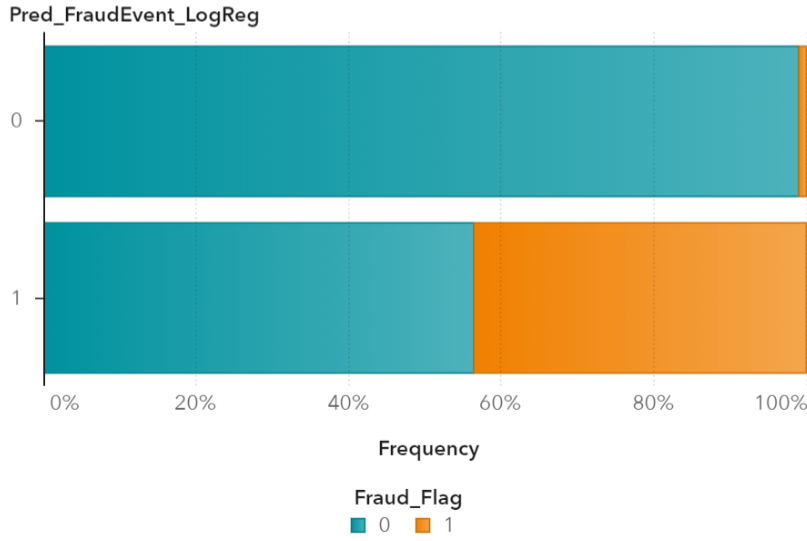
- Demogr
- LogReg
- DecTree
- DecTree Auto
- GradientBoosting
- Model Comparision
- Cutoff Analysis
- +



Classification Fraud YES/NO - Absolute

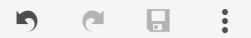


Classification Fraud YES/NO - Relative



- Options
- Roles
- Actions
- Rules
- Filters
- Ranks

# Fraud\_VA\_01



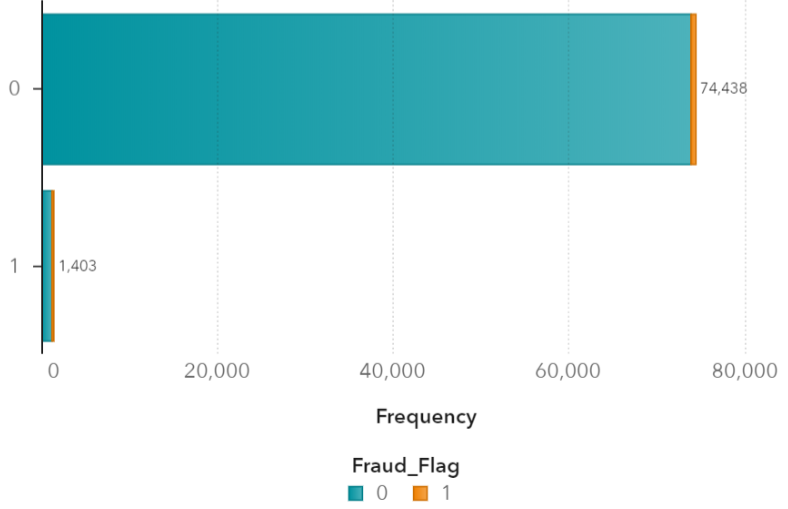
- Demogr
- LogReg
- DecTree
- DecTree Auto
- GradientBoosting
- Model Comparision
- Cutoff Analysis 
- 

## Cutoff for Fraud=YES



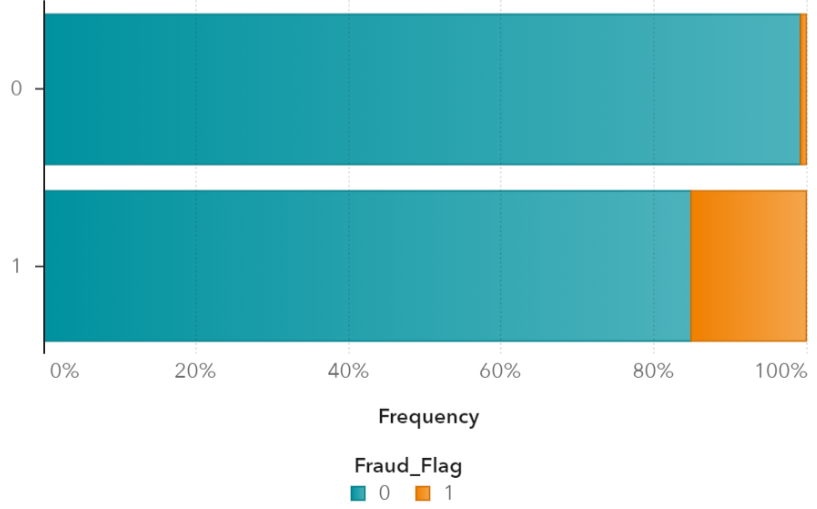
## Classification Fraud YES/NO - Absolute

Pred\_FraudEvent\_LogReg



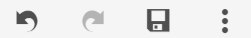
## Classification Fraud YES/NO - Relative

Pred\_FraudEvent\_LogReg



- Options
- Roles
- Actions
- Rules
- Filters
- Ranks

# Fraud\_VA\_01



- Demogr
- LogReg
- DecTree
- DecTree Auto
- GradientBoosting
- Model Comparision
- Cutoff Analysis



- Data
- Objects
- Outline

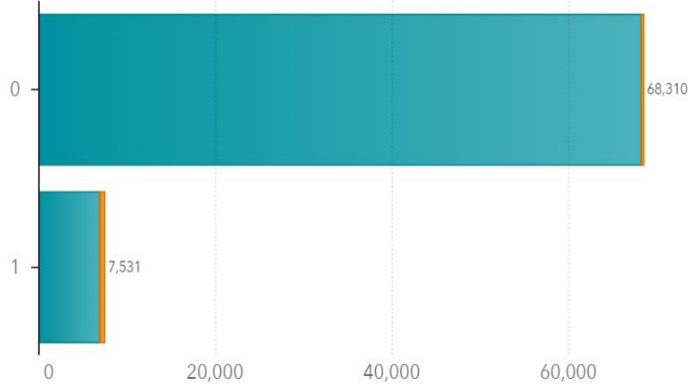
- Options
- Roles
- Actions
- Rules
- Filters
- Ranks

Cutoff for Fraud=YES



Classification Fraud YES/NO - Absolute

Pred\_FraudEvent\_LogReg



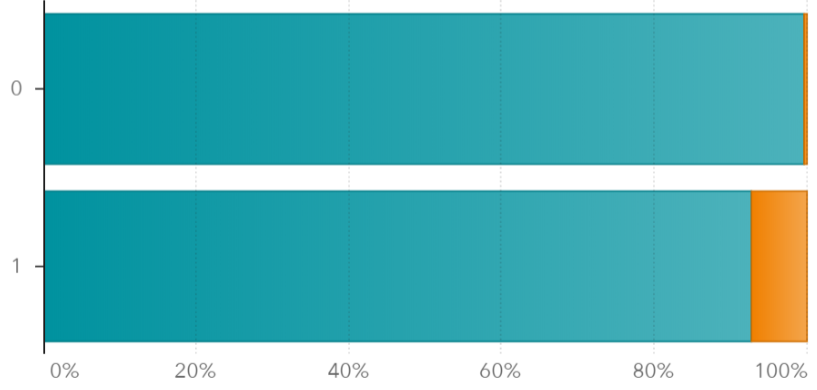
Frequency

Fraud\_Flag

0 1

Classification Fraud YES/NO - Relative

Pred\_FraudEvent\_LogReg



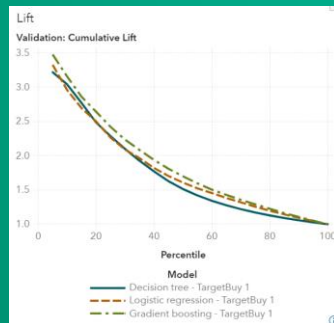
Frequency

Fraud\_Flag

0 1

## Tip #2:

Quantify the importance of explanatory variables  
in a predictive model with a business case





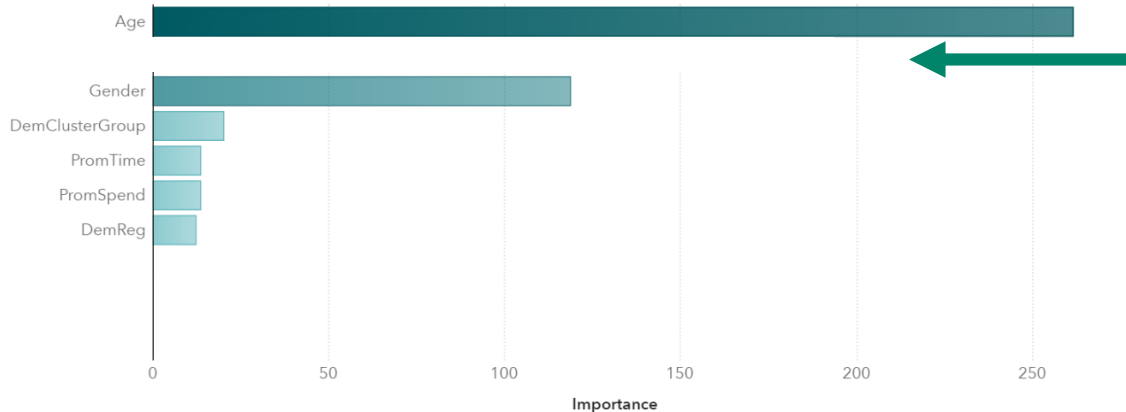
# Variable importance chart in a gradient boosting model

Gradient Boosting TargetBuy (event=1) Validation KS (Youden) 0.5058 Observations Used 100,223

Create Pipeline

< Variable Importance Iteration Assessment >

Variable Importance



What happens, if we do not have variable „AFFLUENCE“ available?

# What happens, if we do not have variable „AFFLUENCE“ available?

- Will other variables substitute the missing content?
  - Will the model quality go down?
1. Create a copy of your model
  2. Remove the variable of interest
  3. Compare the old and the new model

## Gradient boosting - TargetBuy 1

### ▼ Response

 TargetBuy

### ▼ Predictors

 DemClusterGroup

 DemReg

 Gender

 Age

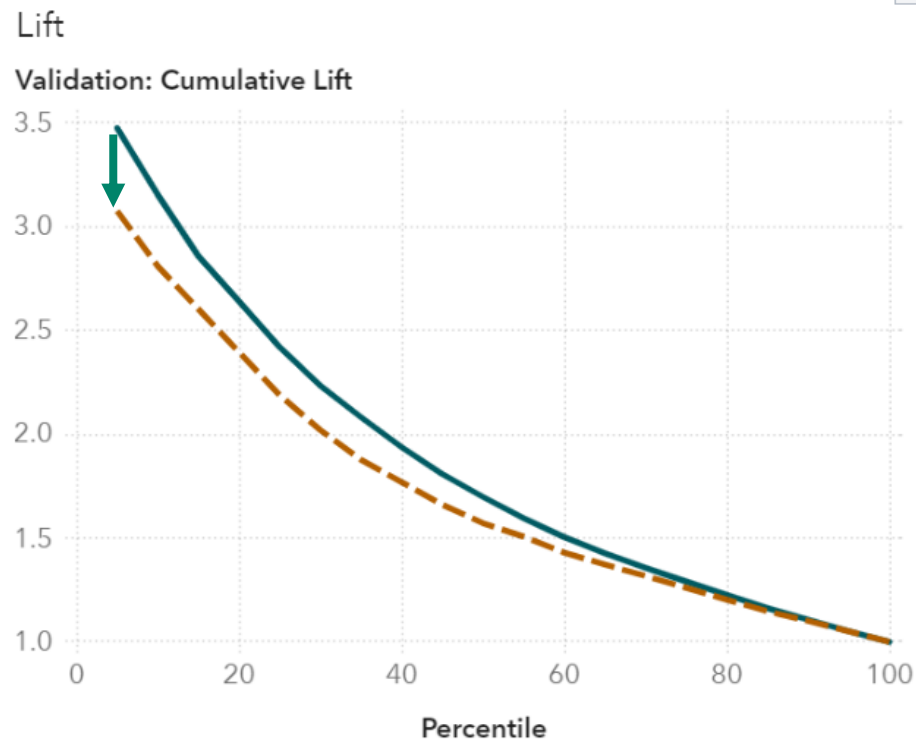
 PromSpend

 PromTime

 + Add

# Compare the **old** and the **new** model

- Lift drops from 3.47 to 3.07
- What does that mean in € ?



# Calculating a business case

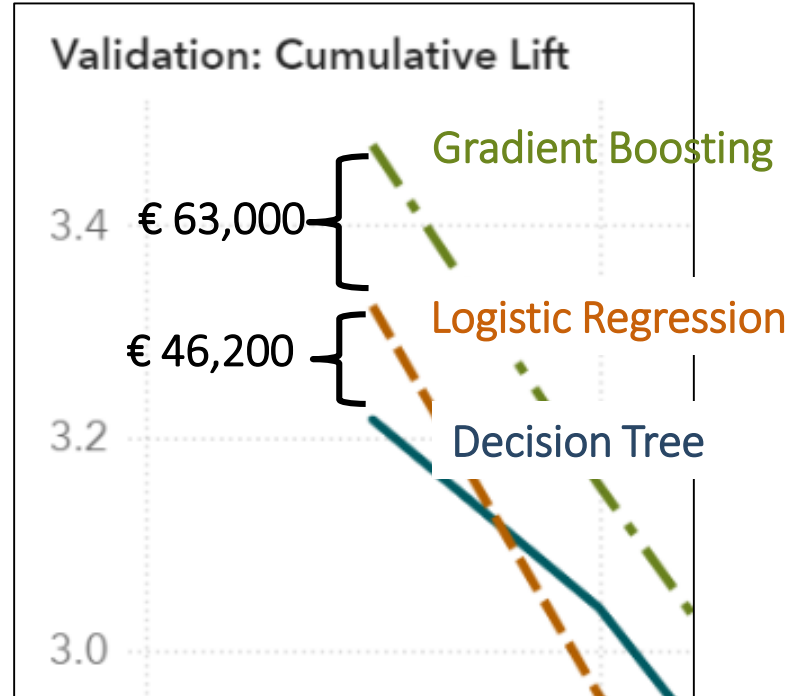
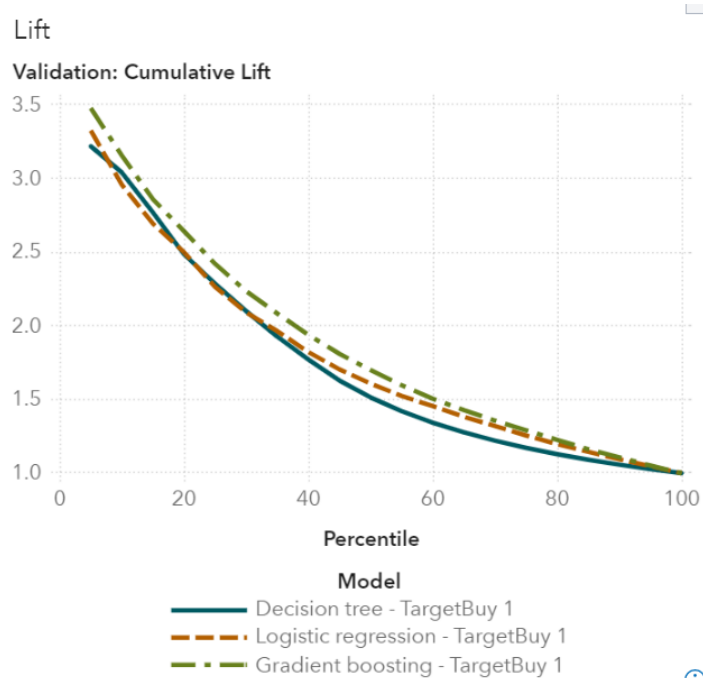
- Assume we have 2 Mio customers
- A campaign offer is sent to the top 5 % (100,000)
- A responding customer contributes a profit of € 35
- Assuming a baseline (autonomous) response of 12 %
  - A lift of 3.47 → 41.64 %
  - A lift of 3.07 → 36.84 %
- Not having variable AFFLUENCE costs us 4.8 % response
  - $100,000 * 4.8 \% = 4800$  missed responders \* €35 = € 168,000

# Quantify the effect of data quality on your business results



- Part III contains simulation case studies for data availability, data quantity, data correctness and data completeness
- Illustrated in € (\$) values based on a business case study
- <http://support.sas.com/svolba>
- <https://github.com/gerhard1050/Data-Quality-for-Data-Science-Using-SAS>

# How much does it cost to use a simpler (better explainable) model for my predictions



## Tip #3:

# Turn on the model interpretability charts in SAS Model Studio

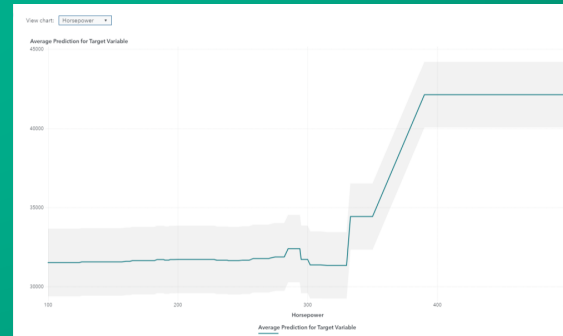
### ▼ Model Interpretability

#### ▼ Global Interpretability

✓ Variable importance

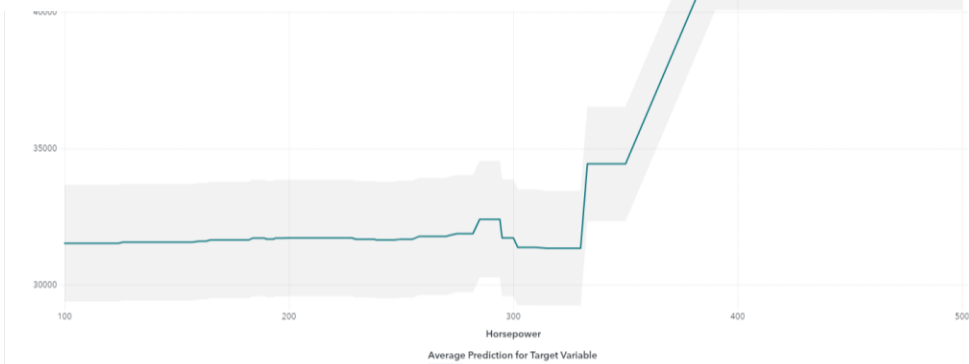
✓ PD plots

Maximum number of variables:



# Partial Dependency Plot (PD)

Horsepower	Average Predicted MSRP
100	\$15400
150	\$17400
...	
...	
...	



Model Interpretability

Global Interpretability

☒ Variable importance

☒ PD plots

Maximum number of variables:



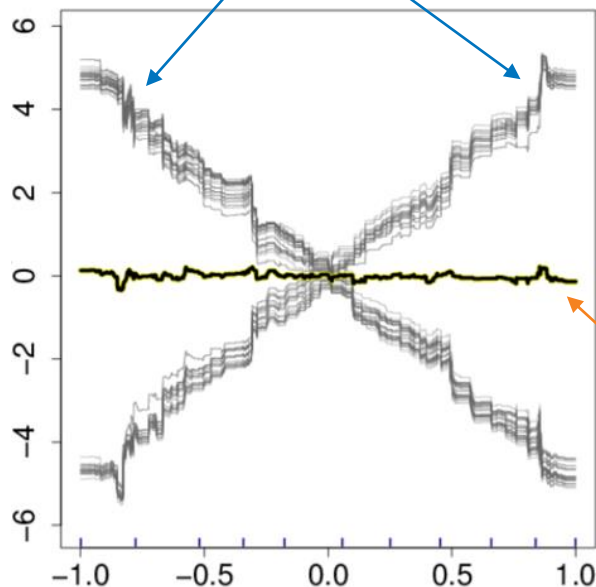
Number of observations:

500



# Individual Conditional Expectation (ICE)

Subgroups of Individual Observations



PD Funktion

Local Interpretability

☒ ICE plots

> LIME



General ICE/LIME Options

Type of instances to explain:

Cluster centroids

Number of clusters:



☐ Create score code for assigning clusters


Number of inputs to use in explanation:

5

# Local Interpretable Model-Agnostic Explanation (LIME)



**Perturbed Instances**



**Local Interpretability**

- ☒ ICE plots
- LIME** ☐

Number of perturbed instances:  
10,000

Number of iterations:  
1 6 10

Distance metric:  
Cosine

Kernel width:  
0

**General ICE/LIME Options**

Type of instances to explain:  
Cluster centroids

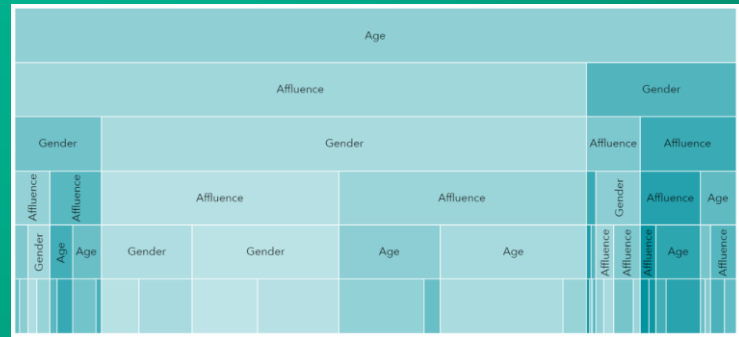
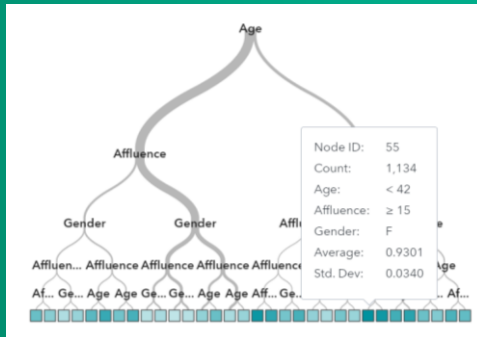
Number of clusters:  
2 6 10

☐ Create score code for assigning clusters

Number of inputs to use in explanation:  
5

## Tip #4:

Use a decision tree to „explain“ why customers received a high/low predicted probability

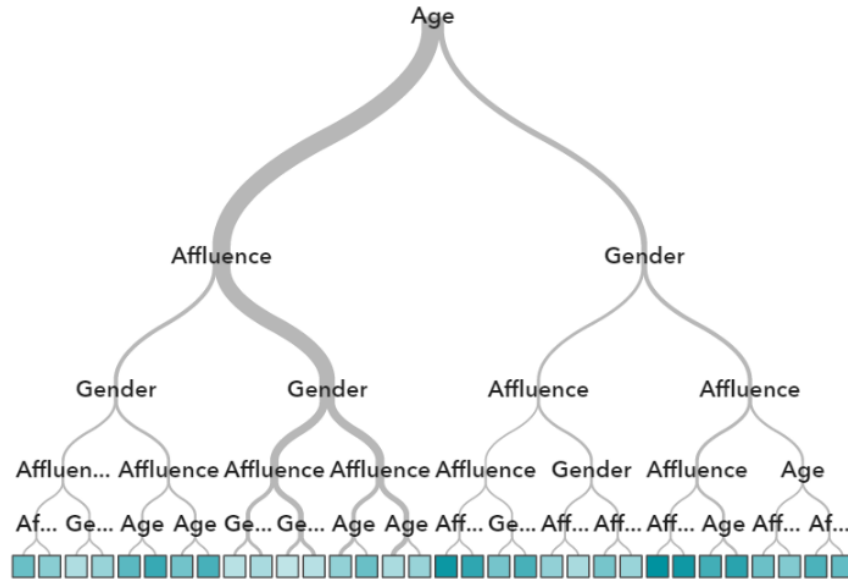


# General Idea

0036 Observations Used 100,223

Create Pipeline

Assessment >



Decision tree - Prob\_Event\_GradBoos...

Response

Prob\_Event\_GradBoost

Predictors

Age

Affluence

Gender

DemReg

DemClusterGroup

PromClass

PromSpend

PromTime

+ Add

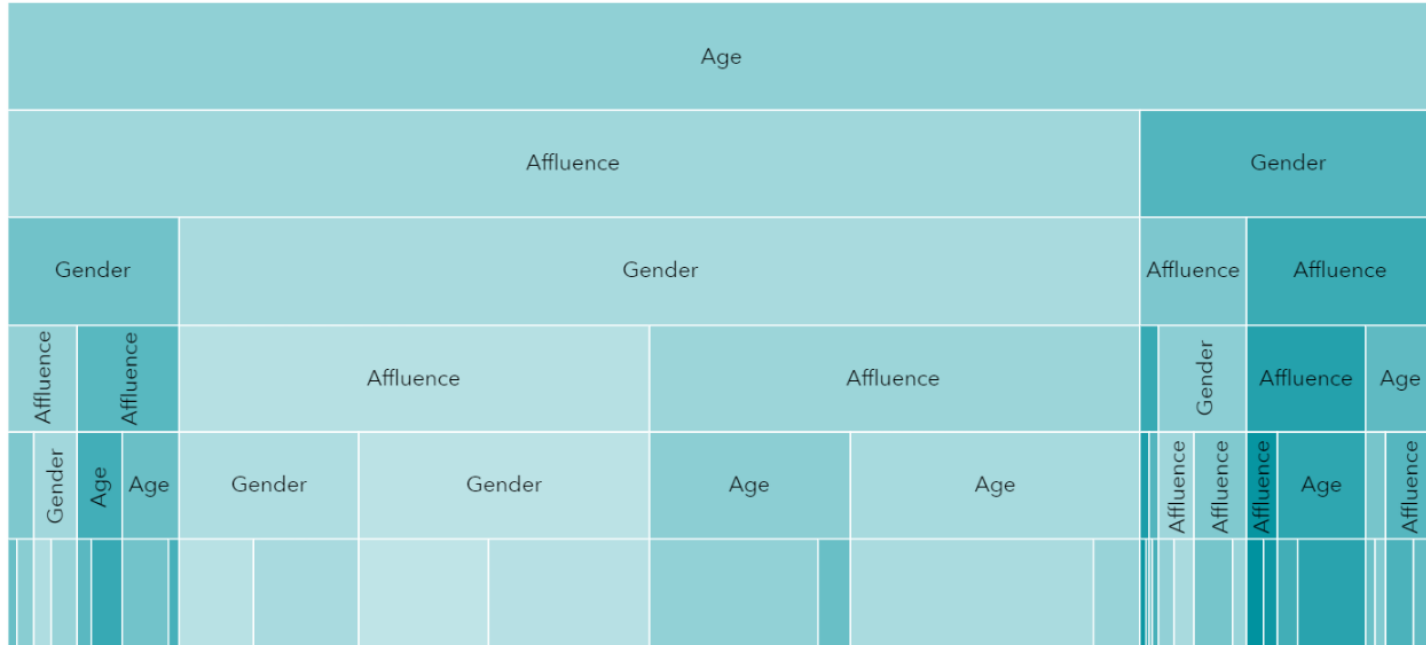
Partition ID

+ Add

Predicted Probability  
from the Gradient  
Boosting Model

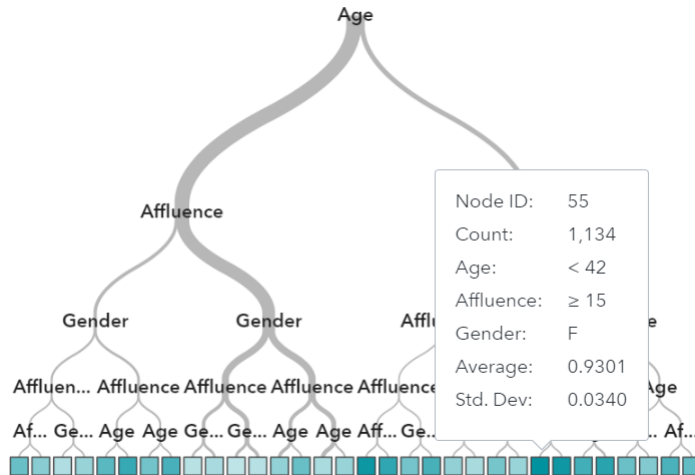
Variables you want to  
use for the explanation

# Decision Tree creates segments with high/low predicted probability

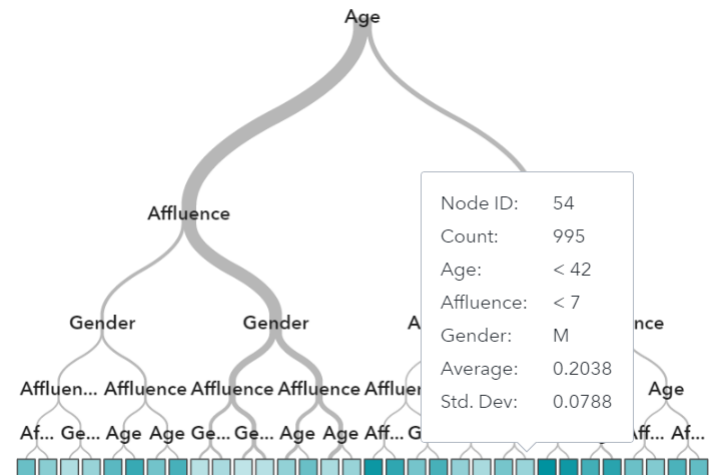


# You can interpret the segments

Young affluent ladies →  
PredictedProb = 93 %



Young non-affluent men →  
PredictedProb = 20 %



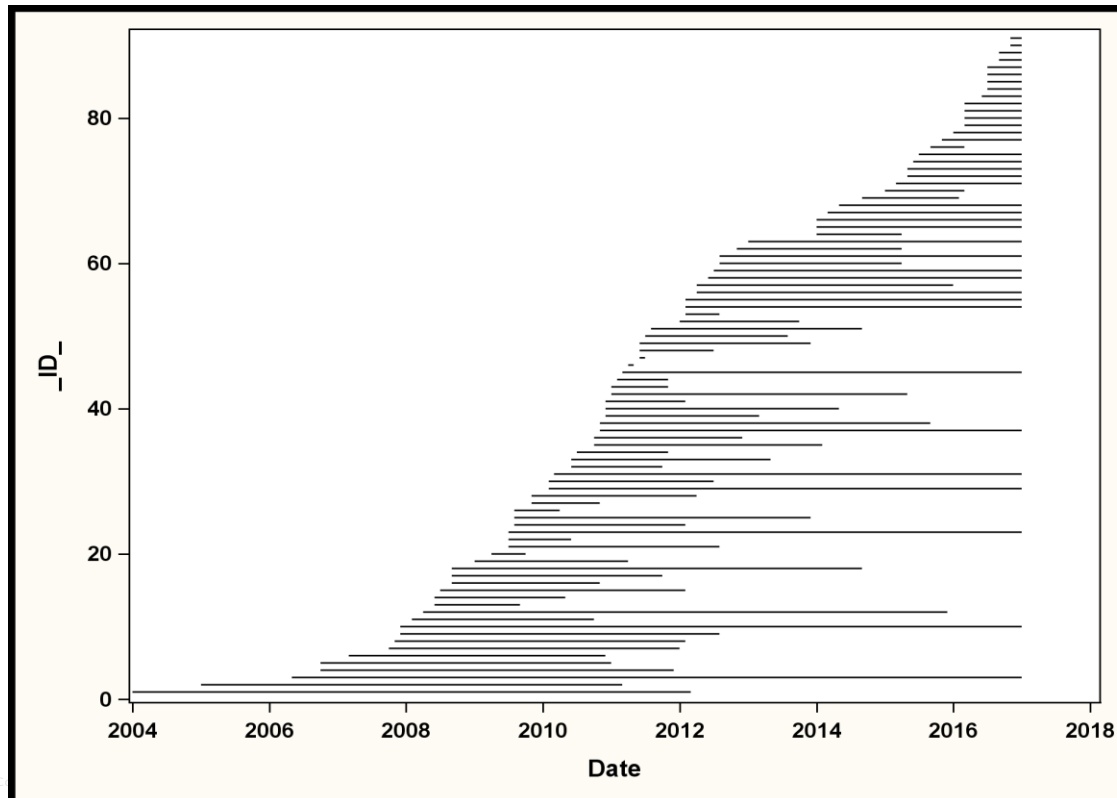
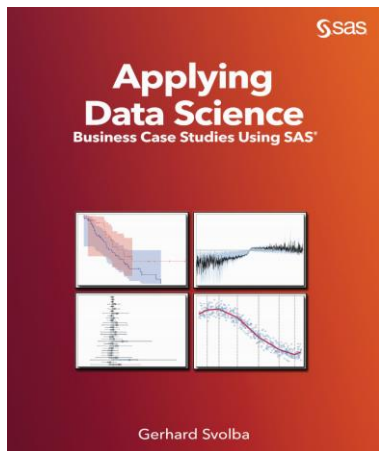


## Tip #5:

Display the (hidden) regression coefficient  
of the reference category

# Survival analysis performed for employee headcount data

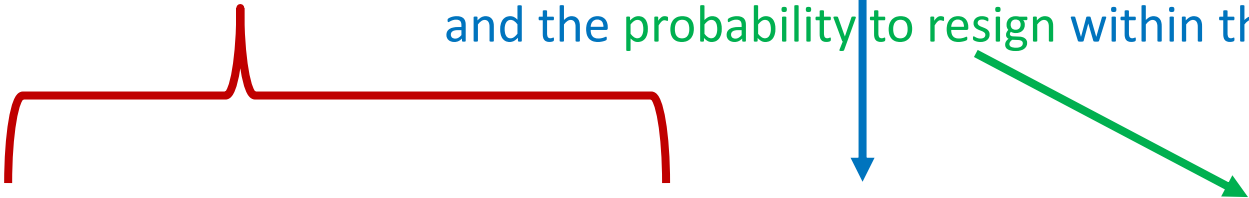
- Observe Careers per Employee
  - Different length
  - „Left company“ or „censored“





# How long will Gerhard still stay in our company?

Given certain risk factors, what is the expected survival in 6 months  
and the probability to resign within the next 6 months.



EmpNo	Department	Gender	TechKnowH...	_T_	EM_SURVFCST	EM_SURVEVENT	T_FCST
1003	TECH_SUPPORT	M	YES	128	0.240	0.000	134
1010	TECH_SUPPORT	M	YES	109	0.240	0.011	115
1023	SALES_ENGINEER	M	YES	90	0.108	0.313	96
1029	TECH_SUPPORT	M	YES	83	0.386	0.133	89
1031	TECH_SUPPORT	F	YES	82	0.177	0.219	88
1037	ADMINISTRATION	M	NO	74	0.471	0.066	80
1045	ADMINISTRATION	M	NO	70	0.494	0.053	76
1054	TECH_SUPPORT	F	YES	59	0.316	0.102	65
1055	SALES_ENGINEER	M	YES	59	0.313	0.103	65

# Regression Analysis – Displaying the (hidden) value of the reference category when using deviation coding

Class Level Information					
Class	Value	Design Variables			
Department	ADMINISTRATION	-1	-1	-1	-1
	MARKETING	1	0	0	0
	SALES_ENGINEER	0	1	0	0
	SALES_REP	0	0	1	0
	TECH_SUPPORT	0	0	0	1
TechKnowHow	NO	-1			
	YES	1			

Analysis of Maximum Likelihood Estimates							
Parameter		DF	Parameter Estimate	Standard Error	Chi-Square	Pr > ChiSq	Hazard Ratio
Department	MARKETING	1	-1.15513	0.47794	5.8414	0.0157	0.606
Department	SALES_ENGINEER	1	0.82336	0.52244	2.4838	0.1150	4.380
Department	SALES_REP	1	0.62976	0.25224	4.6436	0.0312	3.609
Department	TECH_SUPPORT	1	0.35572	0.29940	1.4117	0.2348	2.744
TechKnowHow	YES	1	-0.63474	0.27370	5.3781	0.0204	0.281

$$- [(-1.155) + 0.823 + 0.630 + 0.356] = -0.654$$

```

PROC PHREG DATA=Employees;
  CLASS department gender TechKnowHow / PARAM=effect REF=first;
  MODEL Duration*Status(1) = department gender TechKnowHow /
    SELECTION=stepwise;
RUN;

```

# Download the SAS-macro

<https://github.com/gerhard1050/Applying-Data-Science-Using-SAS>

## Macro Parameters

The following parameters can be specified with the macro.

### **ParmEst**

The name of the data set that contains the ParameterEstimates, created with the ODS OUTPUT statement. Default = ParameterEstimates.

---

### **ClassLevels**

The name of the data set that contains the ClassLevelInfo, created with the ODS OUTPUT statement. Default = ClassLevelInfo.

### **OutputDS**

The name of the data set that shall contain the output data set. Default = \_ParmEst\_XT\_.

```
%CALC_REFERENCE_CATEGORY|( ParmEst      = ParameterEstimates1,  
                           ClassLevels = ClassLevelInfo1);
```

The output format of ParmEst and ClassLevels varies between different regression procedures in SAS. Please contact the author (Email: [sastools.by.gerhard@gmx.net](mailto:sastools.by.gerhard@gmx.net)) in case your output file does not match the requirements of the macro.

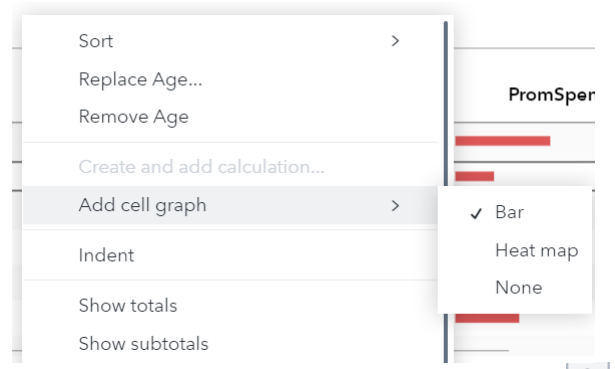


## Bonus Tip:

Use sparklines and bars to illustrate properties of customer segments and clusters

# Cluster Profiling

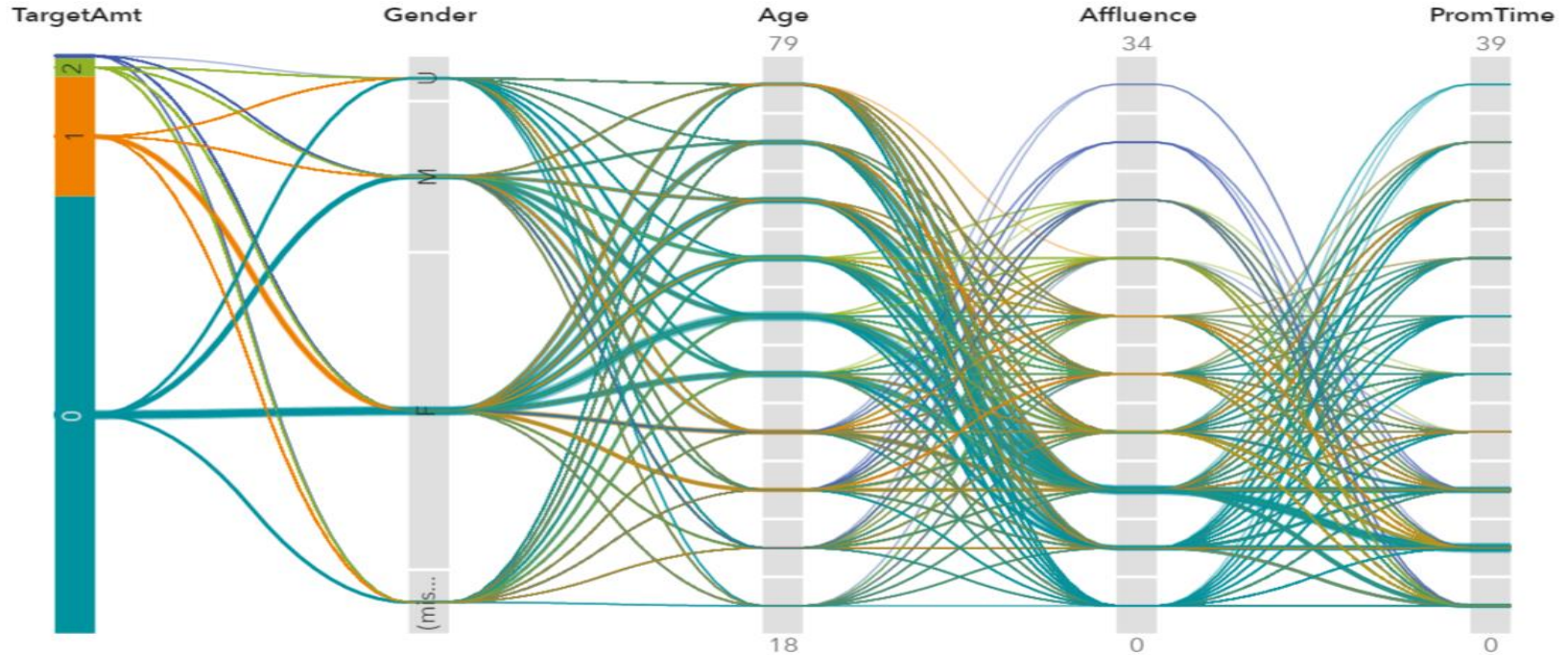
- Create a crosstab in SAS Visual Analytics
- Add barcharts to illustrate the values



Clust erID ▼	Frequency	Frequency Percent	Age	Female	Affluence	PromSpend	PromTime	Response%
5	19,463	19.42%	63	0.55	11	6071	6	21.60%
4	29,902	29.84%	42	0.56	8	2475	6	28.52%
3	9,459	9.44%	42	0.67	15	2588	6	66.48%
2	24,192	24.14%	63	0.50	6	5798	6	7.71%
1	4,727	4.72%	67	0.51	8	7525	21	14.13%

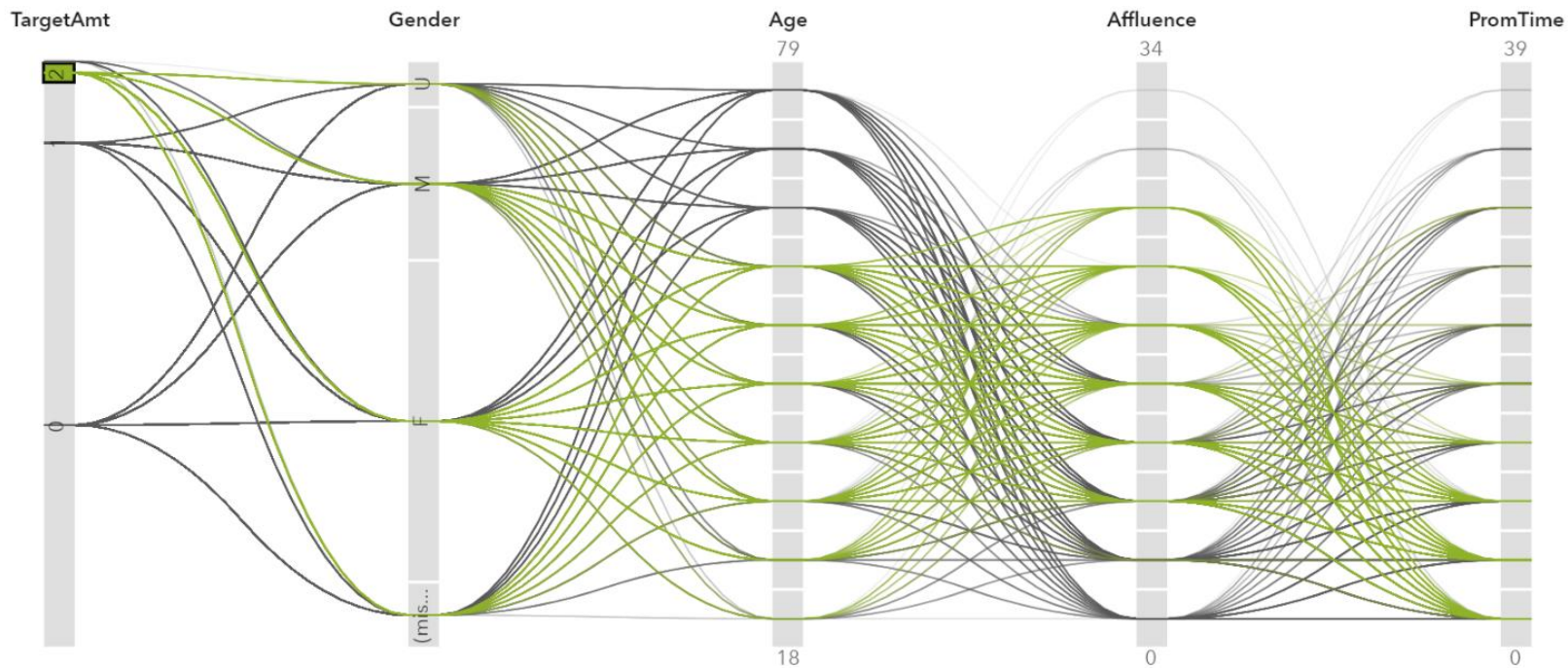
# Use a Parallel Coordinate plot to illustrate cluster features

Parallel Coordinates of Selected Variables



# More affluent customers buy more often

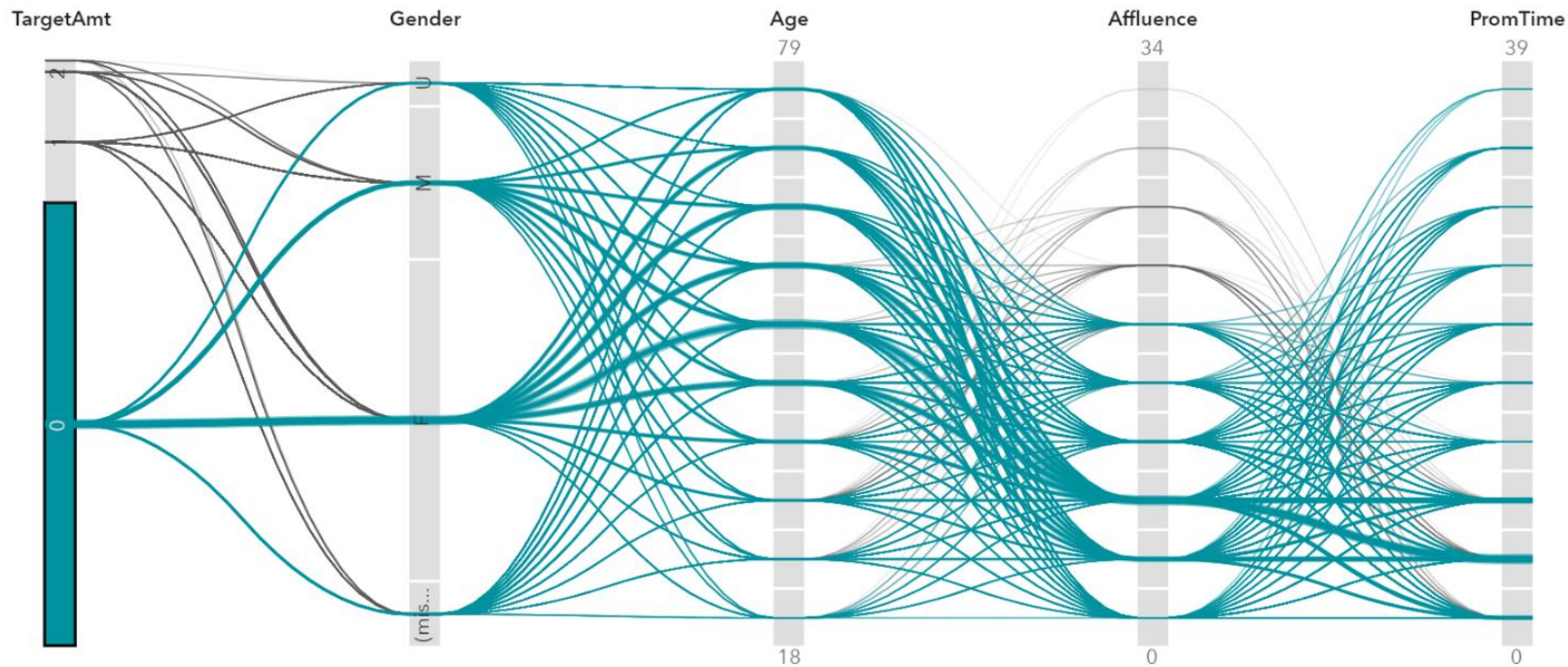
Parallel Coordinates of Selected Variables





# Non-Buyers are older and less affluent

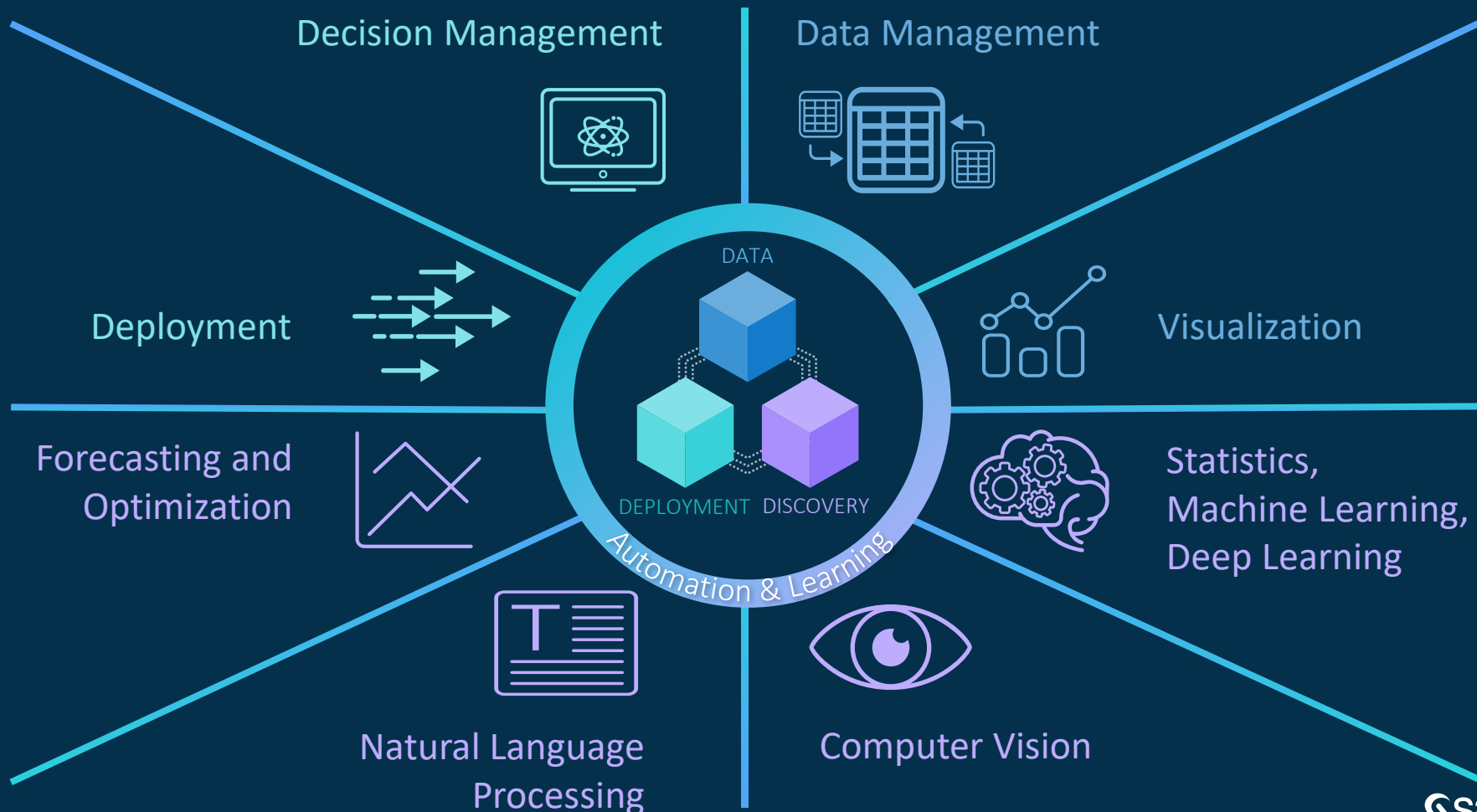
Parallel Coordinates of Selected Variables





# Conclusion

- Communicating model results has many dimensions
- SAS Viya offers you a broad range of tools and methods to illustrate your findings
- Machine learning models that are understood are likely to have a higher business impact



# Agenda

15:30 - 16:00 Uhr	Registrierung
16:00 - 16:15 Uhr	<b>Begrüßung &amp; Einleitung</b> Gerhard Svolba, SAS
16:15 - 16:45 Uhr	<b>Die Kommunikation von analytischen Ergebnissen und die Interpretation von ML-Modellen mit SAS Viya - 5 Tipps und Tricks, die Ihr Leben als Data Scientist vereinfachen</b> Gerhard Svolba, SAS
16:45 - 17:10 Uhr	<b>SAS in der Cloud</b> Phillip Manschek, SAS
17:10 - 17:40 Uhr	<b>Simplification of Routine Data Analyses &amp; Tasks Using SAS</b> Christoph Gruber, Novartis / EBEWE Pharma Ges.m.b.H.
17:40 - 18:00 Uhr	<b>Die Neuerungen im SAS Enterprise Guide 8.1</b> Christian Url, mayato AT GmbH
ab 18:00 Uhr	Gemütliches Get2Gether

- Downloads: <https://github.com/gerhard1050/DACH-SASUserGroups>