

# ANALYTICS EXPERIENCE 2018



Sept. 17-19 | San Diego



# Customer Lifetime Value via SAS<sup>®</sup> Viya<sup>®</sup>

David C. Ogden, Advisory Analytical Consultant, SAS Institute



SAS and all other SAS Institute Inc. product or service names are registered trademarks or trademarks of SAS Institute Inc. in the USA and other countries. \* Indicates USA registration. Other brand and product names are trademarks of their respective companies.



# A Bit of Terminology

Wikipedia, as of September 2018:

In marketing, **customer lifetime value** (CLV or often CLTV), lifetime customer value (LCV), or life-time value (LTV) is a prediction of the net profit attributed to the entire future relationship with a customer. The prediction model can have varying levels of sophistication and accuracy, ranging from a crude heuristic to the use of complex predictive analytics techniques.

For the most part, I like this definition. Though, I do wish it mentioned the notion of historical value. And I personally would not preface this with “in marketing” alone.



# The Gist

- Customer Lifetime Value (CLV) implementations can be complex, leading to “analysis paralysis”.
- This presentation provides a few “best practices” for how to structure your CLV project.
- SAS<sup>®</sup> Viya<sup>®</sup> makes CLV modeling and analysis easier.
- The end result gets you closer to *managing* customer value.



# Outline of Today's Talk

- Motivating the need for CLV.
- Implementing CLV.
- Using CLV output.
- Demo of CLV implemented via SAS<sup>®</sup> Viya<sup>®</sup>.



# Outline of Today's Talk

- **Motivating the need for CLV.**
- Implementing CLV.
- Using CLV output.
- Demo of CLV implemented via SAS<sup>®</sup> Viya<sup>®</sup>.

If you aren't making data-driven decisions, then you aren't keeping up with your competition.



# Quote-worthy

“In God we trust. All others bring data.”

*W. Edwards Deming*



# The Impetus

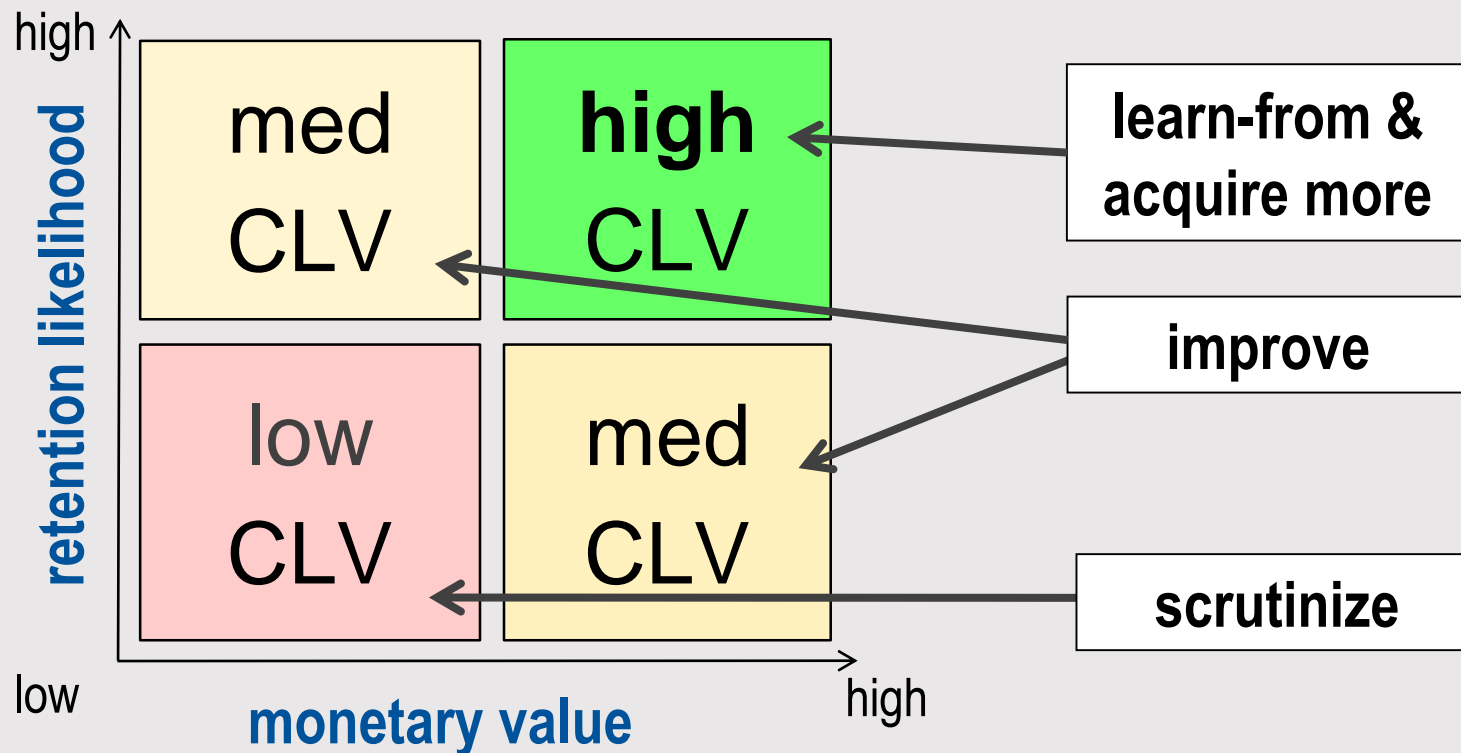
- Who are my *best* customers?
- How do I keep my *best* customers?
- How do I acquire more of the *best* customers?
- How do I make my existing customers *better*?

How do I use an understanding of customer value to make economically sound decisions...  
in marketing, sales, financial planning, operations,  
customer service, process improvements, etc.





## CLV 101



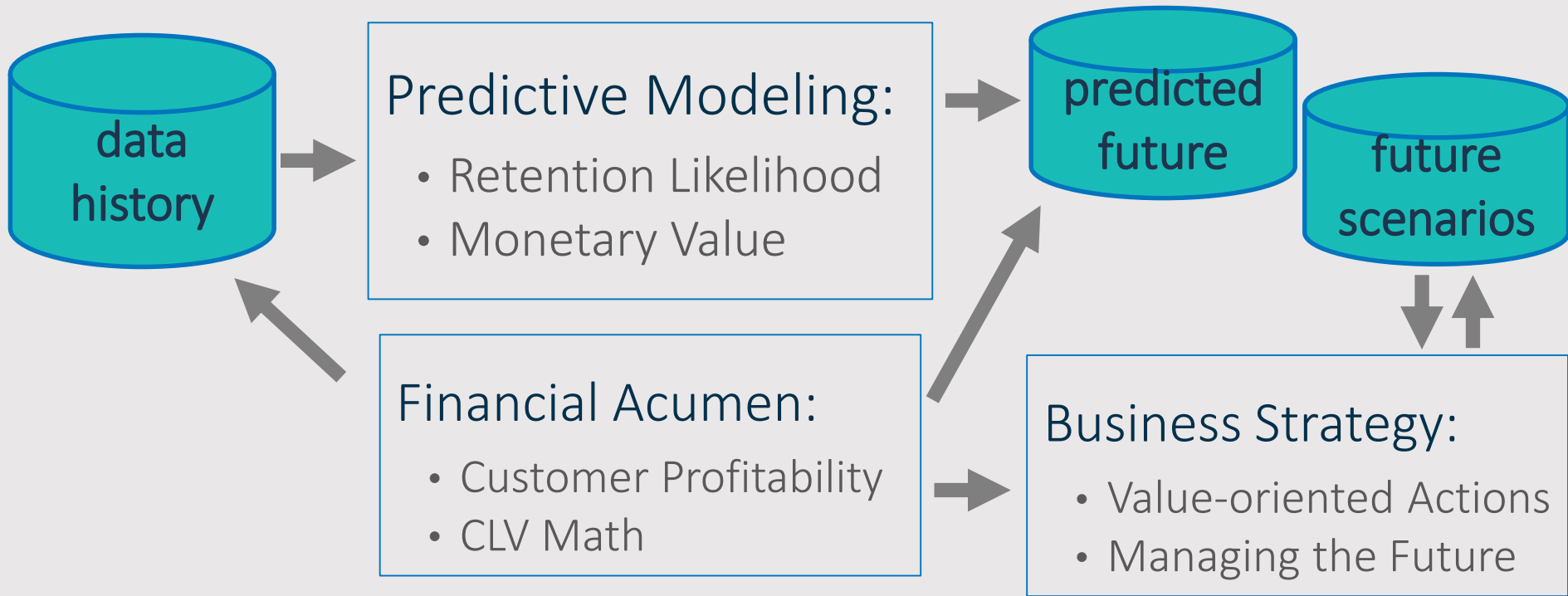
# Outline of Today's Talk

- Motivating the need for CLV.
- **Implementing CLV.**
- Using CLV output.
- Demo of CLV implemented via SAS<sup>®</sup> Viya<sup>®</sup>.

Collecting all the requisite data may be the most time-consuming part of any CLV project. This presentation does not cover the data requirements for a successful CLV project.



# CLV 201



# CLV Step-by-Step

1. **Pick an approach for calculating future value.**
  - Financial acumen and mathematical formulas. (braving your fears)
2. **Calculate historical customer profitability.**
  - You can't predict it if you can't measure it. (accepting imperfection)
3. **Build the predictive models that provide inputs to future value calcs.**
  - All models are wrong; yet some happen to be useful. (gaining insights)
4. **Use the CLV output to make value-oriented business decisions.**
  - Believing in what you built, and actually using it. (walking the walk)



# Pick Your Formula

$$CLV = CHV + CFV$$

$$CHV_j = A_j * (1 + r)^M + \sum_{t=0}^M m_{jt} * (1 + r)^t$$

$$CFV_j \rightarrow m_j \left( \frac{1 - c_j}{d + c_j} \right)$$

$$CFV_j = \sum_{t=1}^{T_j} m_{jt} * \frac{r_{jt}}{(1 + d)^t} + V_j$$

$$CFV_j = \sum_{t=1}^k \frac{m_{jt}}{(1 + d)^t}$$

$$V_j = M_j \left( \frac{1 - C_j}{d + C_j} \right) \left( \frac{1 - C_j}{1 + d} \right)^{T_j}$$

Your presenter welcomes a deep dive discussion into the various options for calculating CLV. That's just how geeky he is.

$$CFV_j \approx \sum_{t=1}^T m_{jt} * \frac{r_{jt}}{(1 + d)^t}$$

$$CLV^k = \sum_{t=0}^T \frac{E_t^k - A_t^k}{(1+i)^t} = \underbrace{(E_0^k - A_0^k)}_{\text{present}} + \underbrace{\frac{E_1^k - A_1^k}{(1+i)^1} + \dots + \frac{E_T^k - A_T^k}{(1+i)^T}}_{\text{future}}$$

The formula you pick is based on the modeling approach you use, and vice versa.



# Quote-worthy

“Beware of geeks bearing formulas.”

*Warren Buffett*



# Calculate Historical Customer Profitability

- An accurate history of customer-specific margins is the first critical component to an accurate projection of future value.
- The degree of accuracy and sophistication is a critical consideration in your CLV project, and depends primarily on two key factors:
  1. How much time do you have to work on this project?
  2. How will the output be used? Basic ranking for marketing lists? Strategic investment planning?
- Whatever you do, please don't simply use revenue as your measure of monetary value. *Please.*

The accuracy of customer-specific margins is as complicated as you want to make it. In order to get a CLV project off-the-ground, *some* “corners must be cut”.



# Build Predictive Models

- Option 1: The Game Theory Approach
  - Churn Model predicts future probabilities (time series) of retention.
  - Monetary Model predicts future cash, given retention.
- Option 2: The Survival Approach
  - Survival Model predicts remaining “life”.
  - Monetary Model predicts customer value over remaining “life”.
- Option 3: The Transactional Approach
  - Predict future transaction counts based on transactional history.
  - Monetary Model predicts value of future transactions.

Within each of these options, there are a variety of modeling techniques.

Each of these options requires history of customer data.

Each option leads to a different math for calculating future value.





And now for the hard part...



# Outline of Today's Talk

- Motivating the need for CLV.

- Implementing CLV.

- **Using CLV output.**

- Demo of CLV implemented via SAS<sup>®</sup> Viya<sup>®</sup>.

Believing in what you built.

Knowing how to use it.

Knowing how *not* to use it.

Convincing others to actually use it.



# Quote-worthy

"Back off, man! I'm a scientist!"

*Peter Venkman*  
*"Ghostbusters" (1984)*



# Two Cardinal Rules for Using CLV Output

1. First and foremost, *understand the mechanics* of what drives future value.
2. Focus on increasing the future value of your customers. It's not about labeling customers...it's about *managing the future*.



# Use Case: Rental Car Company

## Key challenge:

- Over 50% of customers had only one transaction within the last 24 months.

## Approach:

- Build CLV model based on frequent and infrequent transactors.
- Build look-alike model to determine which infrequent transactors *look like* they are really high-CLV (but are probably doing a majority of their business elsewhere).
- Target infrequent transactors with a strategy to gain share-of-wallet.

## Outcome:

- New strategy to target not only the frequent transactors for loyalty membership conversion, but to also target infrequent transactors based on a “CLV look-alike” segmentation.



# Use Case: Video Entertainment (stores vs online)

## Key challenge:

- Video chain launches DVD distribution via web & postal mail; pressure from online competitor.
- Executives disagree on the best investments: build more brick-and-mortar locations? ramp-up in-store coupons? more distribution centers? increase online inventory? more advertising?

## Approach:

- Build macro CLV models, by geography by time (metrics for both online & conventional business).
- Quantify incremental customer equity as a result of various strategic investment scenarios.

## Outcome (recommendations to executives):

- completely cancel the in-store coupon program;
- invest heavily in online inventory;
- fund the online inventory by closing 10% of brick-and-mortar stores.

Recommendations presented in 2007.  
Executives opted to ignore.  
Company filed bankruptcy in 2010.



## Third Cardinal Rule for Using CLV Output

### 3. Avoid the “death spiral”:

Consider this scenario:

- $\text{churn} = f(\text{trend in purchases, customer service calls, etc})$
- $\text{margin} = f(\text{product portfolio, customer service calls, etc})$
- CLV calculated from predicted churn & margin; converted to a simple four-star rating system.
- 3- and 4-star customers are given priority customer service. 1- and 2-star customers are given less than stellar customer service (higher wait times, routing to junior reps, etc).
- New cable box model sent out; one that (unfortunately for the customers) has a lot of technical issues...resulting in many calls to customer service to report issues.
- Over the next few months, customers are calling and complaining. Those calls result in a higher predicted churn, and a lower predicted margin. CLV is decreasing for anyone calling customer service repeatedly. As their CLV decreases, 3- and 4-star customers become 1- and 2-star customers. Which causes their customer service experience to get worse.



# Outline of Today's Talk

- Motivating the need for CLV.

- Implementing CLV.

- Using CLV output.

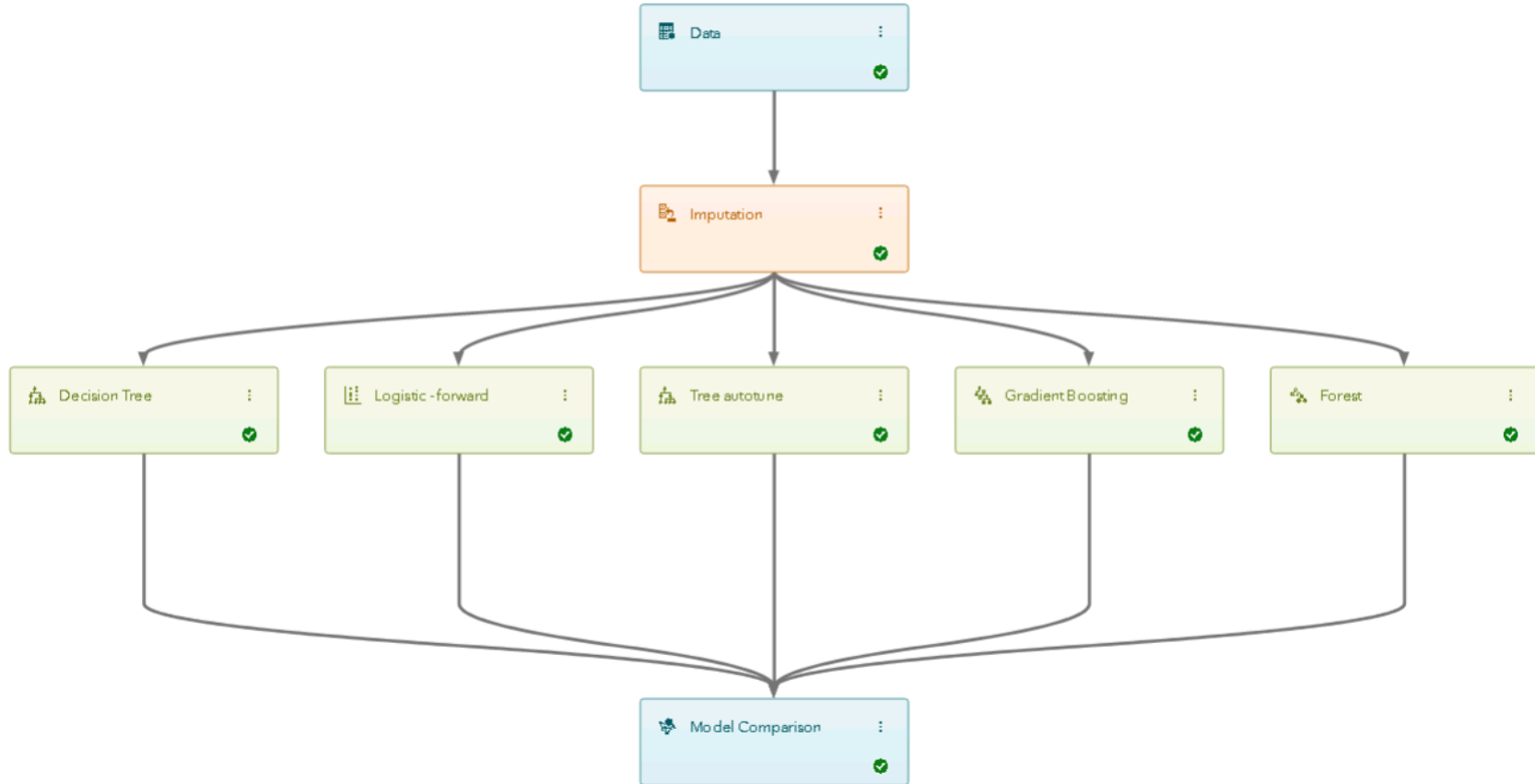
This short demo is just a taste of the capabilities in SAS® Viya®. If you'd like more information, please come visit the Innovation Hub.

- Demo of CLV implemented via SAS® Viya®.





# VDMML – Model Studio pipeline

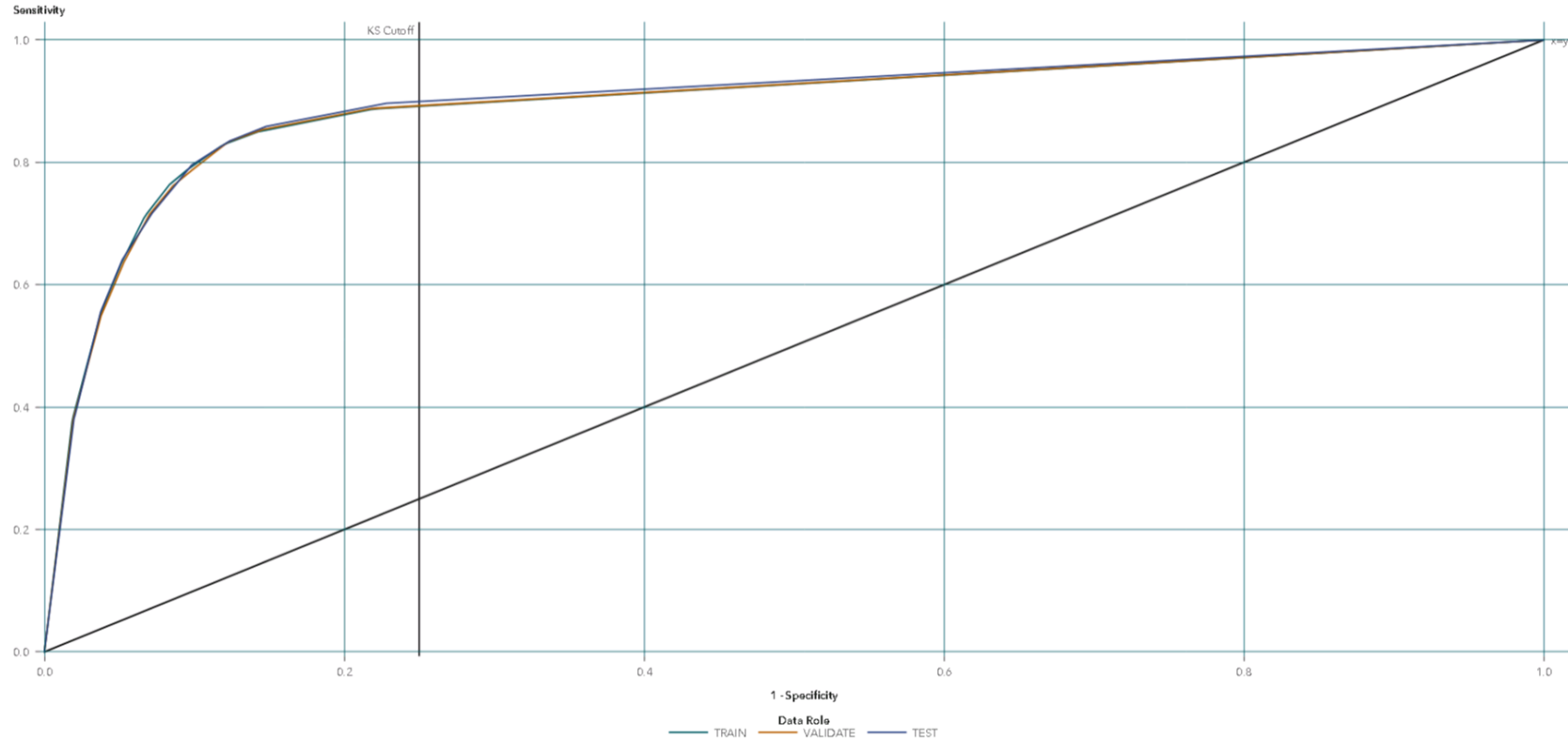


# VDMML – Model Results



ROC Reports

ROC



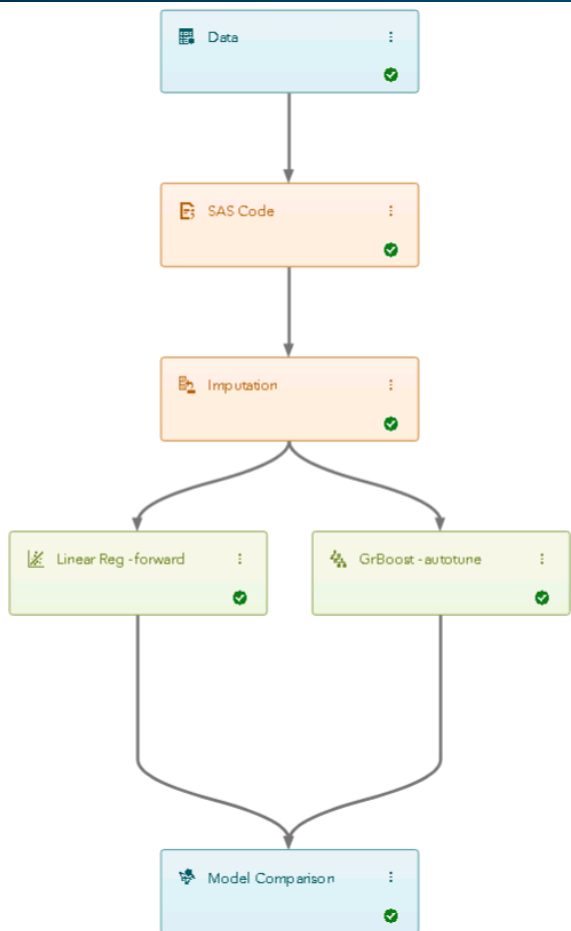
# VDMML – Model Comparison



## Model Comparison

Champion	Name	Algorithm Name	KS{Youden}	Misclassification Rate
A	Tree autotune	Decision Tree	0.7108	0.1260
	Gradient Boosting	Gradient Boosting	0.7106	0.1273
	Forest	Forest	0.7094	0.1296
	Decision Tree	Decision Tree	0.6985	0.1296
	Logistic - forward	Logistic Regression	0.6752	0.1377

# VDMML – Model Studio pipeline



Parameter	t Value	Sign	Estimate
purch_amt_36mo	145.0513	+	0.3463
margin_Y_neg_2	52.4841	-	-0.3683
marg_expctd_ann_v1	36.1631	+	0.5193
purch_amt_life	27.3699	-	-0.0089
purch_cnt_tt_36mo	25.0602	-	-10.0535
purch_avg_36mo	21.6906	-	-0.2052
tier_prev 5-Diamond	16.7459	+	1,018.6807
call_out_mo_36mo	16.4595	-	-31.1185
scheduled_purchase_flg 0	16.2271	-	-179.2938
IMP_demog_inc_fine \$ 250000 or more	15.4262	+	1,157.2779
Intercept	13.8637	+	661.7029
tier_prev 3-Gold	13.3006	+	358.9913
purch_cnt_ck_36mo	11.1002	+	4.6746
call_in_mo_36mo	10.7695	-	-0.8586
tier 4-Platinum	10.4303	-	-470.0420
margin_Y_neg_1	10.3733	+	0.0657
tier_prev 4-Platinum	8.0992	+	354.3811
tier 1-Bronze	7.5375	-	-336.4402
tier 3-Gold	7.3222	-	-329.0016
tier 2-Silver	7.3114	-	-330.5600
tier_prev 2-Silver	5.9123	+	119.9580
tier_prev 1-Bronze	4.4527	+	60.6102
IMP_demog_inc_fine \$ 150000 - \$174999	1.6537	+	79.6605
IMP_demog_inc_fine \$ 200000 - \$249999	1.2222	+	98.3359
IMP_demog_inc_fine \$ 125000 - \$149999	1.0886	+	30.0934
IMP_demog_inc_fine \$ 15000 - \$ 19999	0.8976	-	-17.6690
IMP_demog_inc_fine \$ 40000 - \$ 49999	0.7736	+	13.7450

# Develop SAS Code – SAS Studio



```
34  Ⓣ data public.CLV_SCORED;
35      merge work.original
36            work.churn_scored
37            work.margin_scored;
38      by acct_ID;
39
40      format CFV dollar12.2;
41
42      array P_churn_(3) ;
43      array P_margin_(3);
44
45      do i = 1 to 3;
46          P_churn_(i) = 1- (1-P_churn_year_plus1_ind1)**i;
47
48          *** since the margin model is based only on future non-churners ***;
49          *** the predicted values need to be adjusted per the risk of churn ***;
50          *** Note: we are using the same margin in each future year, but a cumulative retention risk ***;
51
52          P_margin_(i) = (1-P_churn_(i))*(P_margin_Y_pos_1);
53
54      end;
55
56      CFV = P_margin_1/((1+annual_discount)**1) + P_margin_2/((1+annual_discount)**2) + P_margin_3/((1+annual_discount)**3);
57  run;
58
59  Ⓣ proc casutil;
60      promote casdata="CLV_SCORED"
61      Incaslib="public" Outcaslib="public" CASOUT="CLV_SCORED";
62  run;
63
```

# VA – Explore and Visualize Data



Frequency of CFV

Frequency

30,000

20,000

10,000

0

\$0

\$1,000

\$2,000

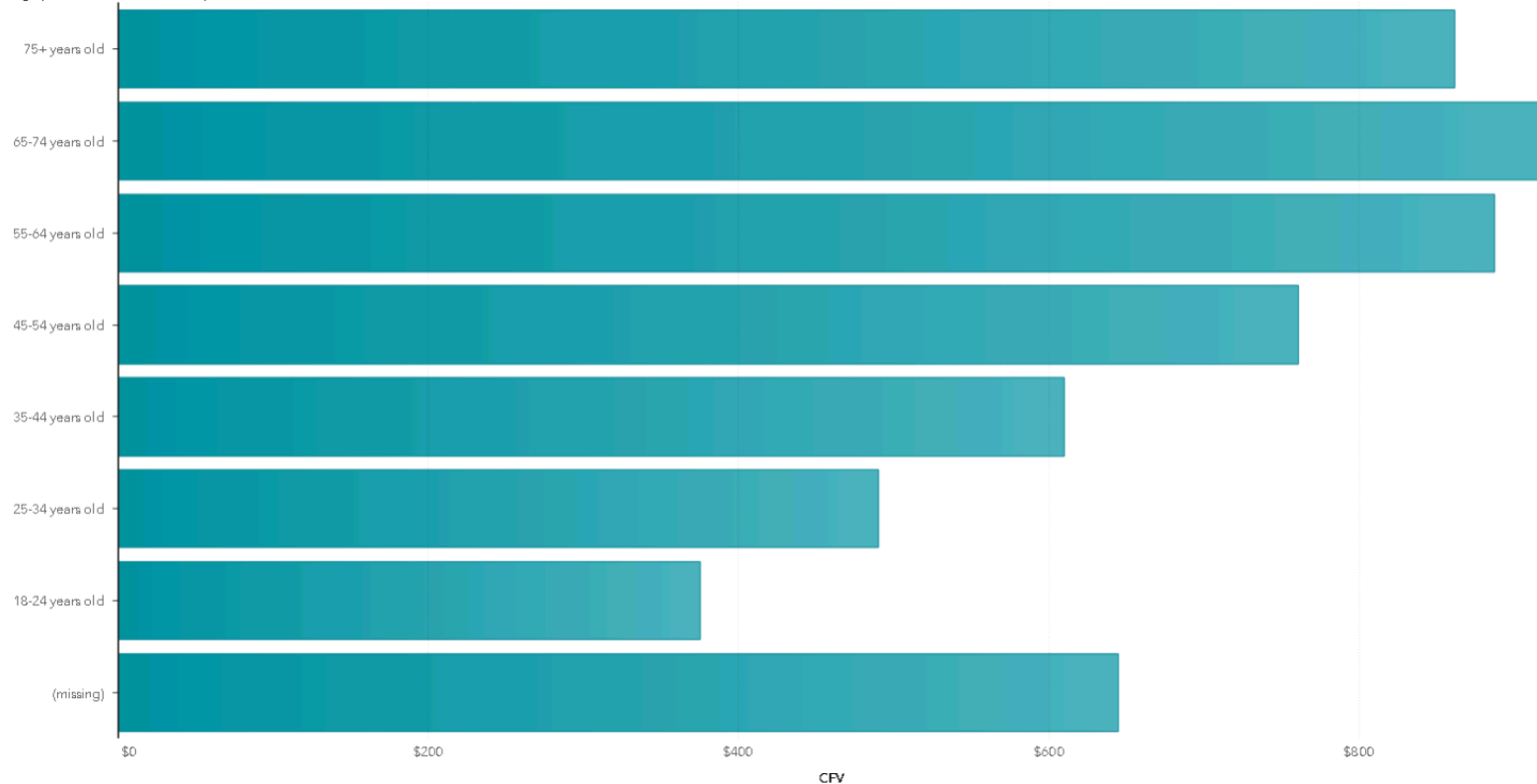
CFV

# VA – Explore and Visualize Data



Average CFV by Age (Head of Household)

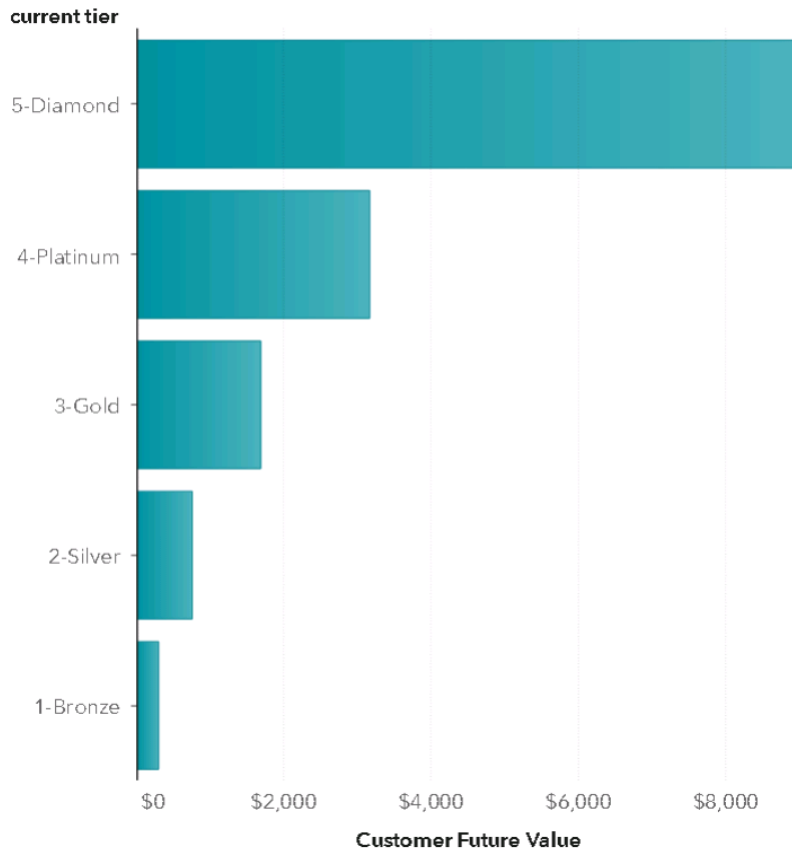
Age (of the Head of Household)



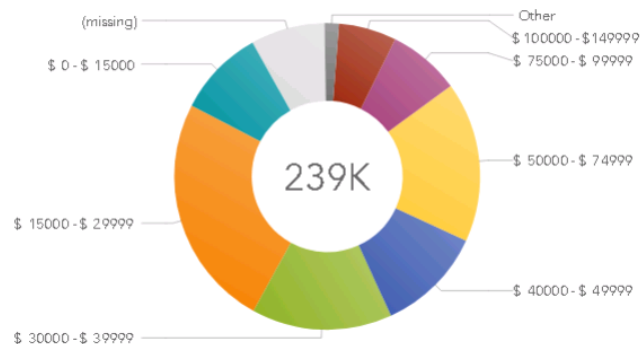
# VA – Explore and Visualize Data



Customer Future Value by current tier



Frequency of Total Household Income, course values



Own or Rent the Residence	Annual Churn Probability	Future Annual Margin
	32%	\$314
Definite Owner (any type dwelling unit)	23%	\$470
Definite Renter	28%	\$280
Probable Owner (any type dwelling unit)	29%	\$388
Probable Renter	34%	\$245



# Food for Thought

A client once asked me, “In testing CLV, what is the percentage correct for the dollar-values assigned to customers? How often are the numbers right?”

The answer: “Zero percent. But that’s OK. You can still use it to make better decisions.”

Is CLV the end-all-be-all measure?

Think beyond customer-specific cashflow.

Example: Social Media Influencers.



# Questions?

Don't forget to complete the session survey  
in your conference mobile app.

1. Go to the Agenda icon in the conference app.
2. Find this session title and select it.
3. On the Sessions page, scroll down to Surveys and select the name of the survey.
4. Complete the survey and click Finish.

