



# **Predicting and Preventing Churn with Customer Health Scoring**

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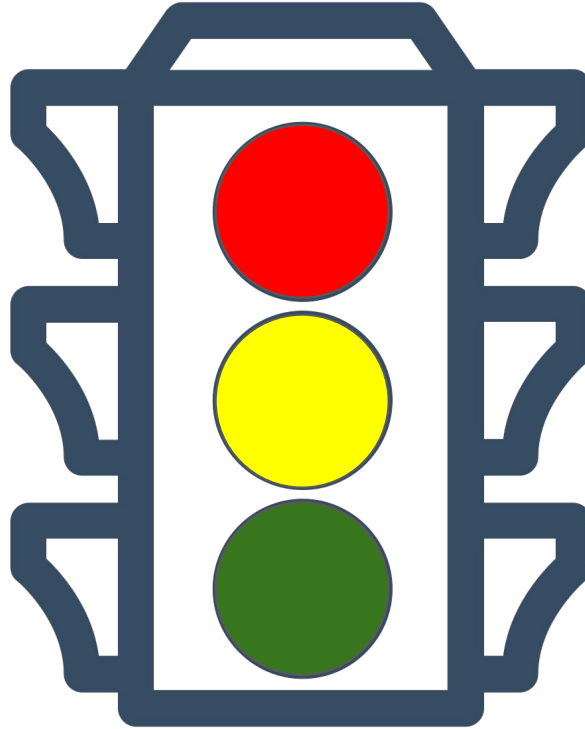
# My experience in Customer Health Scoring



Intelisecure



# What is a customer health score?



**A “red-yellow-green” assessment  
of the level of risk with the account**

# What is it used for?

## Focus

Helps your Customer Success Managers be more proactive about where they spend their time

## Automation

In tech-touch, product-led, fully automated environments, health scores can trigger actions to support, retain, and upsell customers

## Forecasting

CFOs at many companies associate probabilities to health score color codes and use weighted sums to estimate future revenue

## Improvement

health dashboards communicate account status, and when supported, they can show how improvements helps customer health and installed base revenue

# Where do teams typically start?

“What’s your gut say about this account?”

“I don’t know. I think I’d call them Yellow.”

<u>CustomerID</u>	Risk	ACV
1	Yellow	\$65,000
2	Green	\$80,000
3	Green	\$72,000
4	Red	\$140,000

# Why is predictive accuracy important?

**Bad accuracy = bad decisions**



***The watermelon problem***

# When do you need to get more sophisticated in your approach?

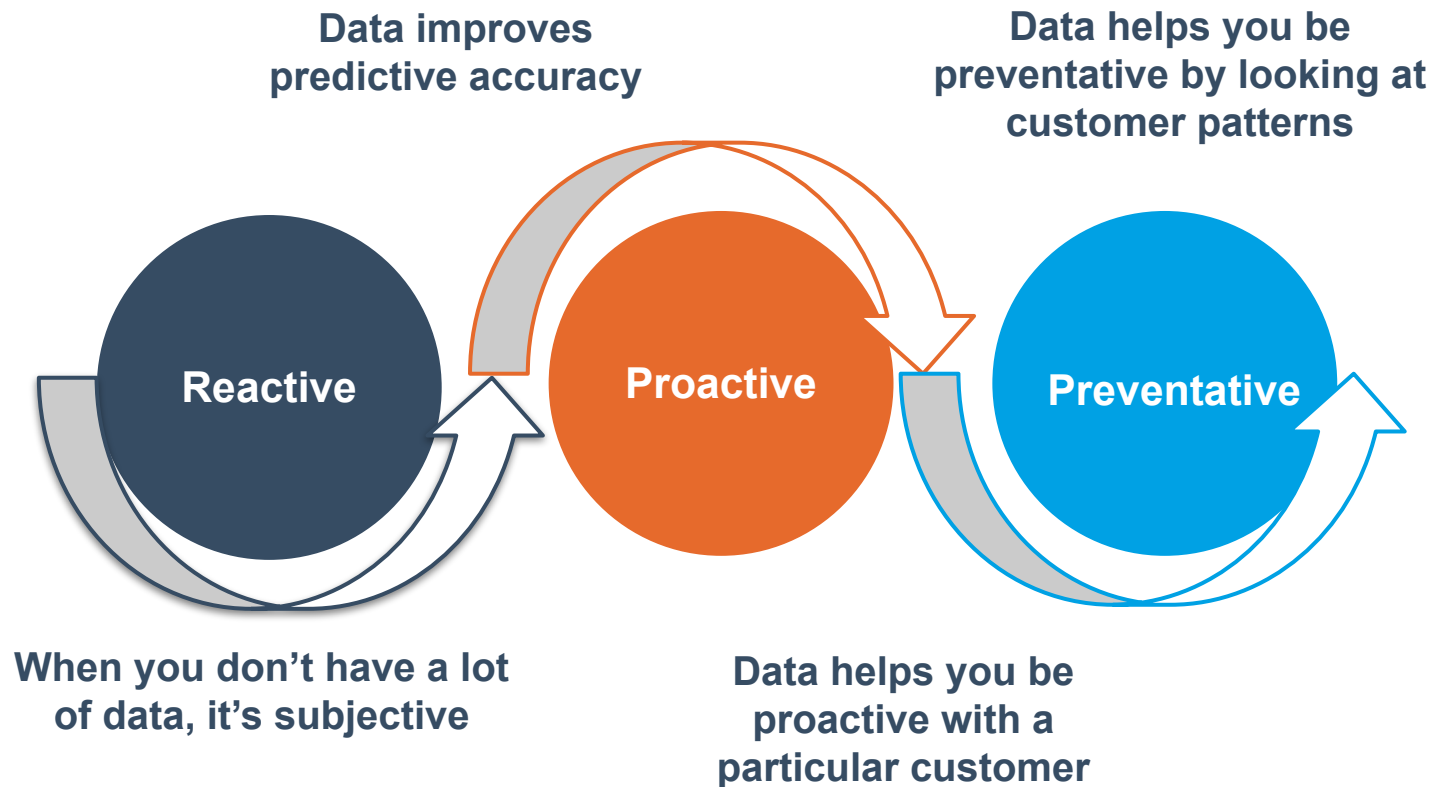
**In sales-led companies, you get more sophisticated as you add more accounts**



**In product-led (PLG) companies, you build models very early**



# How does adopting a more analytical approach help you move from reactive to proactive to preventative when it comes to churn?





**FOR DISCUSSION**

# **Which variables are you using in your health score?**

**Usage data  
(what kind?)**

**Adoption  
data  
(what kind?)**

**Sentiment  
data  
(what kind?)**

**Customer  
goals or fit  
attributes**

**Support  
tickets  
(what kind?)**

**Payment  
history**

**Something  
else?**

**COME OFF MUTE OR COMMENT IN THE CHAT**

# What types of data can go into health scores?



**Usage data**



**Onboarding and adoption**



**Sentiment and satisfaction**



**Customer goals and TTV**



**Support tickets (volume, criticality, TTR)**



**Payment history**



**Fit**

**How many variables should you include?**

**5-7**

# How do typical CS platform vendors compute customer health scores?

What factors are you using?

Well our other customers use NPS, usage, number of support tickets, ...

Let's start with NPS. So what's a "good" number for you?

OK. Now how should we weight NPS?

I don't know. What do you suggest?

OK. That sounds good.

I don't know. Maybe more than 7...?

I don't know. Maybe 20%...?

**Customer Health Score = subjective factor \* subjective weighting + subjective factor \* subjective weighting + ...**

# What would the steps be to build a more sophisticated, regression-based model?

1

**Learn why customers leave and why others stay & buy more**

2

**Form hypotheses**

3

**Collect and clean the right data (this part is hard)**

4

**Analyze factors**

5

**Develop and test the model**

6

**Deploy the model**

# What's an example?

**Y: Cancel/Renew**

**$X_1$ : Lifetime Bookings**

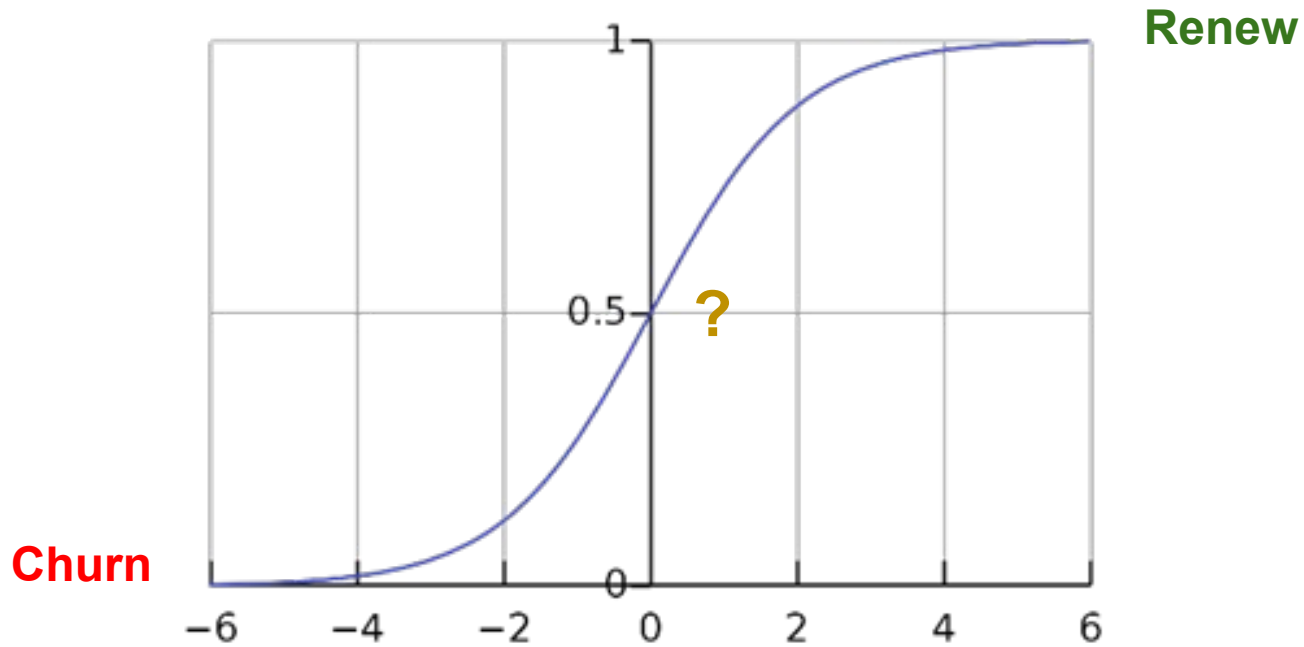
**$X_2$ : Weeks On**

**$X_3$ : No Shows**

**$X_4$ : Q Ratings**



# Model



$$\hat{y}(\theta) = \frac{1}{1 + e^{-\theta}}$$

$$\theta = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \dots + \beta_n x_n + \varepsilon$$

# Logistic Regression

$\beta$

	<i>coeff b</i>	<i>s.e.</i>	<i>Wald</i>	<i>p-value</i>	<i>exp(b)</i>	<i>lower</i>	<i>upper</i>
Intercept	-1.53395	0.22	50.3	1.3E-12	0.22		
X <sub>1</sub> Lifetime Bookings	0.10789	0.01	173	2E-39	1.11	1.1	1.13
X <sub>2</sub> Weeks On	-0.00029	0	0.01	0.94291	1	0.99	1.01
X <sub>3</sub> No Shows   All Time	-0.09058	0.02	33.8	6.2E-09	0.91	0.89	0.94
X <sub>4</sub> Q Ratings   All Time	-0.004	0.04	0.01	0.92772	1	0.91	1.09

$$\theta = -1.53 + 0.11 * \textit{Lifetime Bookings} - 0.09 * \textit{No Shows} + \varepsilon$$

$$\hat{y}(\theta) = \frac{1}{1 + e^{-\theta}}$$

**Classification Table**

	Suc-Obs	Fail-Obs	
Suc-Pred	279	67	346
Fail-Pred	87	352	439
	366	419	785
Accuracy	0.762295	0.840095	<b>0.803822</b>



# Deployment

AccountID	Lifetime Bookings	No Shows	Retention
373	0	0	18%
432	0	1	19%
300	7	2	36%
184	11	0	42%
393	9	4	46%
247	12	11	69%
160	22	4	78%
145	23	3	78%
173	15	13	78%
371	26	0	79%
386	26	2	82%
145	26	2	82%
324	35	7	95%
146	63	0	99%



# What tech tools do you need to run the analysis and operationalize the score?

Excel  
(and a plug-in)



Statistical  
Software Packages



Customer  
Success Platforms



CHURNZERO

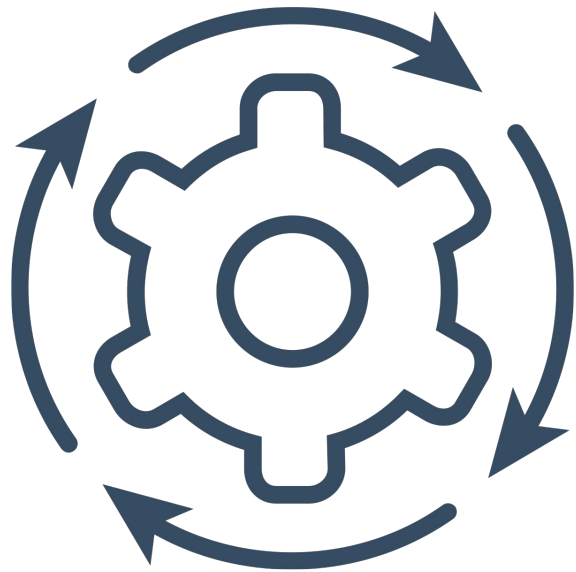
STURDYAI



ZAPSCALE

# Who's responsible for developing and updating the customer health score?

**CS Operations**



**A consultant  
(if you don't have CS Ops)**



# How often should you revisit or adjust your model?

**Check after a big external change  
(e.g. COVID)**



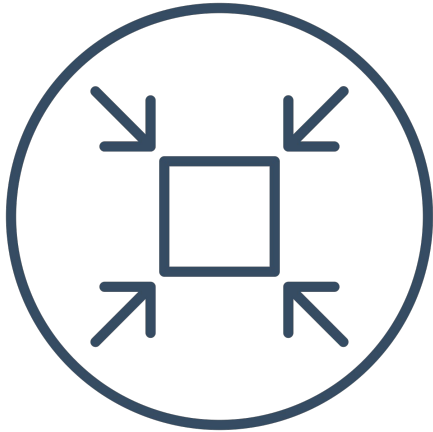
The majority of models are based on historical evidence, so when a dramatic change happens that affects a sector, you can expect the models to become less and less predictive.

**Test to see if the model is still  
performing well (and adjust when not)**

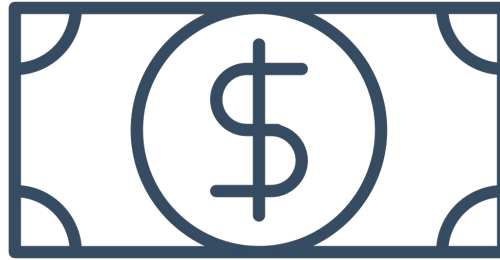


You want to measure predictive accuracy at periodic intervals and understand if it's still performing within a margin of error that you want. If it degrades, then it's time to update it.

# What are the most important pieces to get right?



**Start outside in  
(with the customer)**



**Connect  
it to money**



**Apply the  
scientific method**



# Questions