

# Target Accounts Analysis & *Weighted Model* with *R & Excel*

# Obtain Weights for Model via Logistic Regression in R

```
46
47 # perform logistic regression
48 full_model <- glm(pipe_gen ~ ., data=dataset, family=binomial)
49 model_probs <- full_model %>% predict(dataset, type="response")
50 process_probs <- preProcess(as.data.frame(model_probs), method=c("range"))
51 norm_probs <- predict(process_probs, as.data.frame(model_probs))
52 summary1[[i]] = summary(full_model)
53 }
54
55 # RESULTS
56 coeffs <- (full_model$coefficients)[-1]
57 process <- preProcess(as.data.frame(coeffs), method=c("range"))
58 norm_coeffs <- predict(process, as.data.frame(coeffs))
59 weights <- (norm_coeffs/sum(norm_coeffs))*100
60 weights <- format(round(weights,2), nsmall=2)
61 df <- data.frame(coeffs, norm_coeffs, weights)
62 df <- df %>% rename("original coeffs"=coeffs,
63                   "normalized"=coeffs.1,
64                   "weights (%)"=coeffs.2)
65 df
```

	Original Coeffs	Normalized		Weights
Account Health Score	0.51640369	0.3738399	0.37/2.21 =	16.94%
6Sense Account Intent Score	0.80258252	0.5972739	0.6/2.21 =	27.06%
6Sense Account Profile Fit	1.31840252	1.0000000	1/2.21 =	45.30%
Global Ultimate Parent Employee Count	0.03758173	0.0000000	0/2.21 =	0.00%
Interesting Moments	0.34022550	0.2362889	0.24/2.21 =	10.70%
Sum		<b>2.2074027</b>		<b>100%</b>

# Input Weights into Excel Model & Identify Target Accounts

A	B	C	D	E	F	G	H	I	J	K
		Marketing Segment		Employee Segment		Health Score	6sense	6sense	#Interesting Moments	Has SQLs ?
2024	Weight in Scoring	15%		15%		5%	22%	18%	15%	10%
Ultimate Parent Comp	Account Owner	Segment	Sub Segment	Employee Segment	Focus Rating	Health Score	6sense intent Score	6sense profile score	Interesting Moments This Year	SQLs Created this year

		Percentile Normalization Ranking							
Value Score	Percentile of Score	Marketing Segment	Employee Segment	Health Score	6sense intent Score	6sense profile score	#Interesting Moments	SQLs	
0.99	100%	1	1	1	0.95	1.00	0.998	0.997	
0.98	100%	1	1	1	0.93	1.00	0.993	0.998	
0.98	100%	1	1	1	0.94	1.00	0.995	0.955	
0.98	100%	1	1	1	0.92	1.00	0.997	0.996	

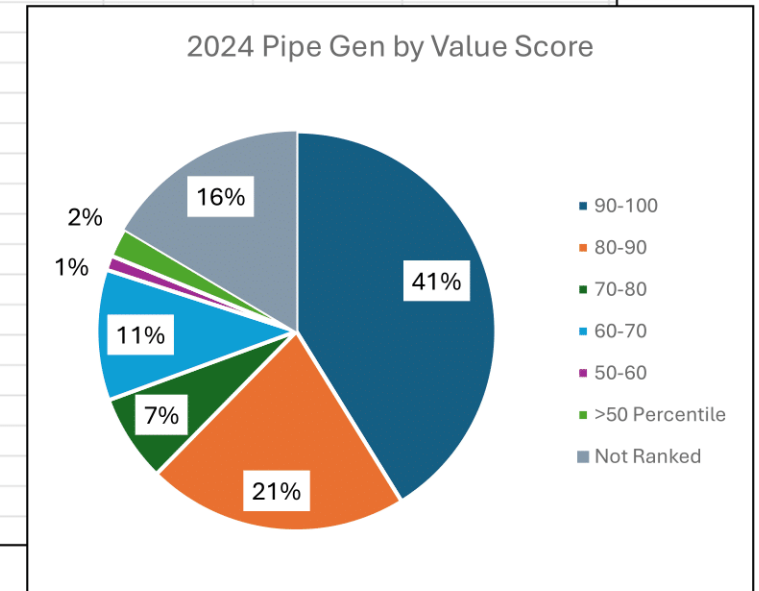
Weights were further adjusted based on conversations with Sales Managers and regional VPs. Above are screenshots of columns that are weighted and their normalized counterparts. After normalization, each account is assigned a Value Score intended to help identify target accounts.

*Check if model is properly capturing accounts*

F	G	H	I	J	K	L	M	N	O	P	Q
2023						2024					
Value Score	Percentile of Score	Interesting Moments	SQLs	Elite Pipe Gen	Closed Won	Interesting Moments	SQLs	SQLs (#)	Elite Pipe Gen	lite Pipe Gen	Closed Won
0.1	0%	0 \$	-	\$ -	\$ -	0 \$	-	-	\$ -	-	\$ -
0.1	0%	0 \$	-	\$ -	\$ -	0 \$	-	-	\$ -	-	\$ -
0.1	0%	0 \$	-	\$ -	\$ -	0 \$	-	-	\$ -	-	\$ -
0.1	0%	0 \$	-	\$ -	\$ -	0 \$	-	-	\$ -	-	\$ -
0.1	0%	0 \$	-	\$ -	\$ -	0 \$	-	-	\$ -	-	\$ -
0.1	0%	0 \$	-	\$ -	\$ -	0 \$	-	-	\$ -	-	\$ -
0.1	0%	0 \$	-	\$ -	\$ -	0 \$	-	-	\$ -	-	\$ -
0.1	0%	0 \$	-	\$ -	\$ -	0 \$	-	-	\$ -	-	\$ -
0.1	0%	0 \$	-	\$ -	\$ -	0 \$	-	-	\$ -	-	\$ -
0.1	0%	0 \$	-	\$ -	\$ -	0 \$	-	-	\$ -	-	\$ -
0.1	0%	0 \$	-	\$ -	\$ -	0 \$	-	-	\$ -	-	\$ -
0.1	0%	0 \$	-	\$ -	\$ -	0 \$	-	-	\$ -	-	\$ -
0.1	0%	0 \$	-	\$ -	\$ -	0 \$	-	-	\$ -	-	\$ -
0.1	0%	0 \$	-	\$ -	\$ -	0 \$	-	-	\$ -	-	\$ -
0.1	0%	0 \$	-	\$ -	\$ -	0 \$	-	-	\$ -	-	\$ -
0.1	0%	0 \$	-	\$ -	\$ -	0 \$	-	-	\$ -	-	\$ -
0.1	0%	0 \$	-	\$ -	\$ -	0 \$	-	-	\$ -	-	\$ -
0.1	0%	0 \$	-	\$ -	\$ -	0 \$	-	-	\$ -	-	\$ -
0.1	0%	0 \$	-	\$ -	\$ -	0 \$	-	-	\$ -	-	\$ -
0.1	0%	0 \$	-	\$ -	\$ -	0 \$	-	-	\$ -	-	\$ -
0.1	0%	0 \$	-	\$ -	\$ -	0 \$	-	-	\$ -	-	\$ -
0.1	0%	0 \$	-	\$ -	\$ -	0 \$	-	-	\$ -	-	\$ -

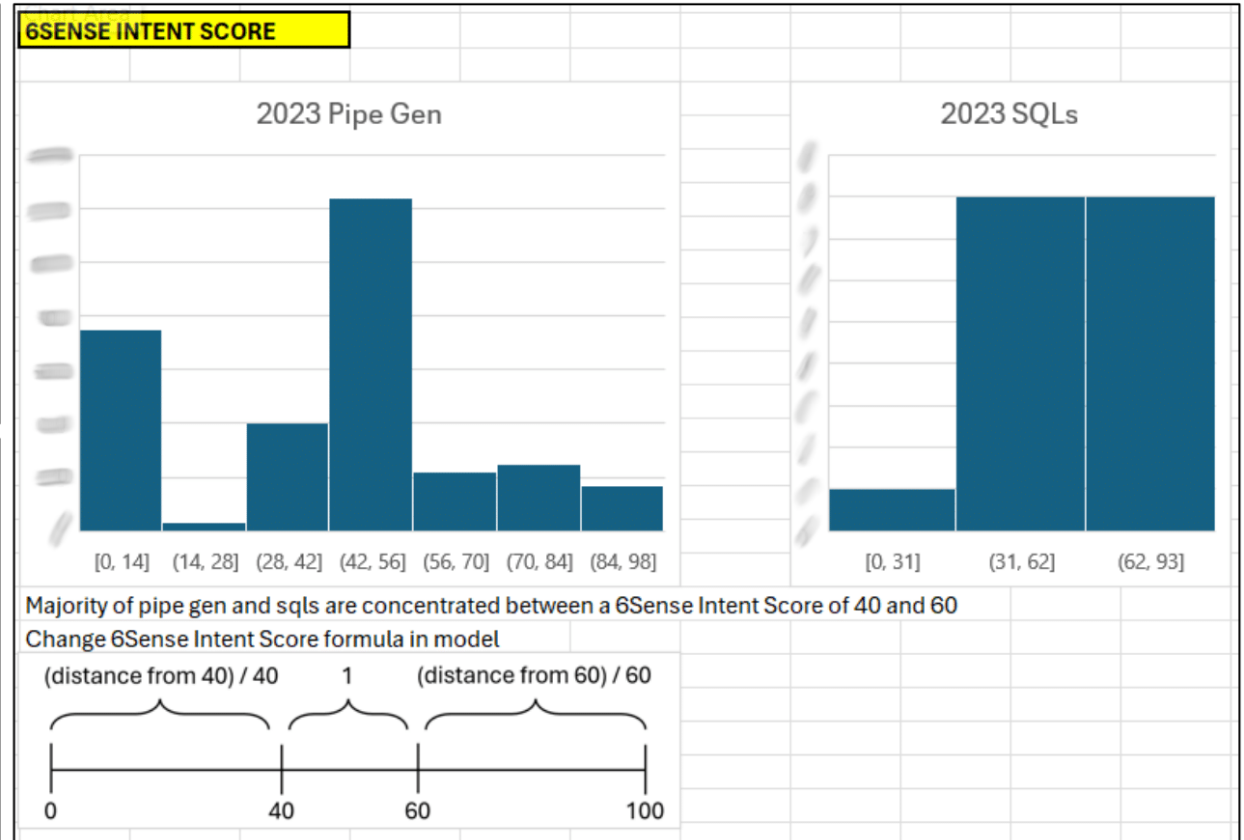
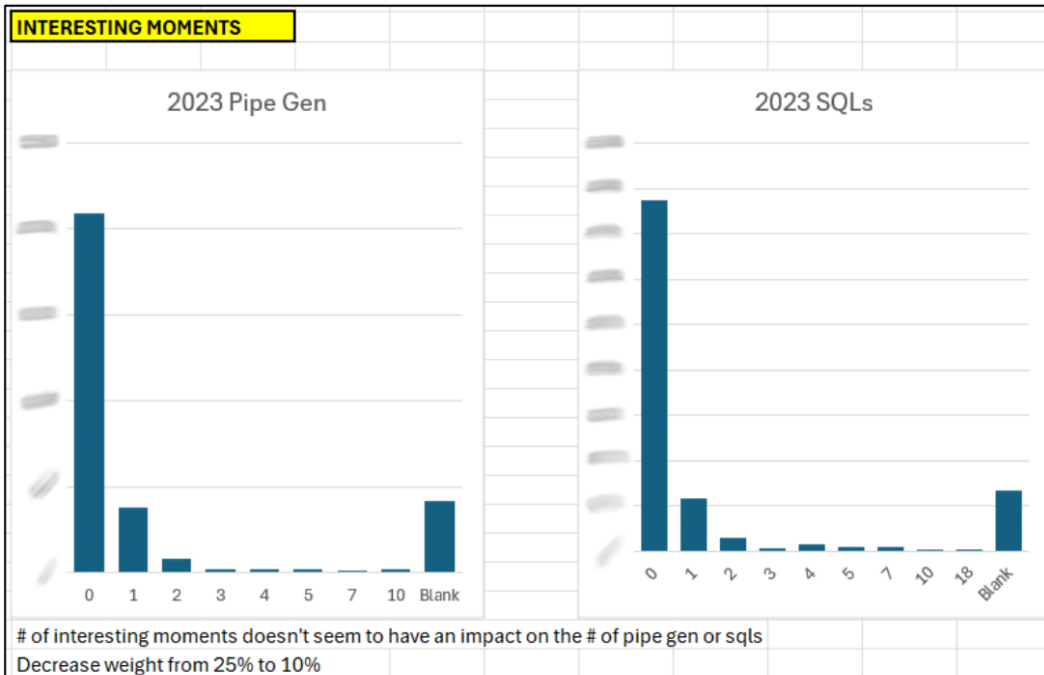
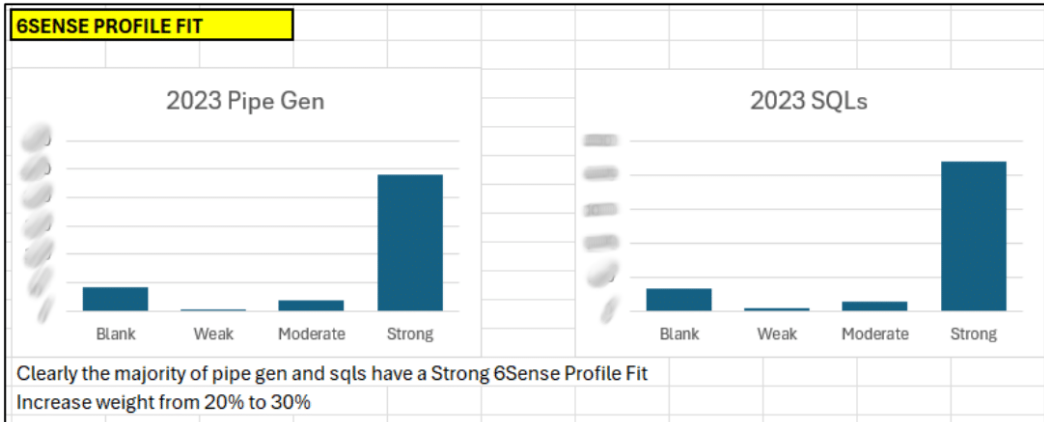
2024 Pipe Gen by Value Score

Value Score Range	Percentage
90-100%	41%
80-90%	21%
70-80%	7%
60-70%	11%
50-60%	2%
>50%	1%
No Data	16%



Since the model was based on 2023 data, we check 3 months into 2024 to see if the model is properly capturing the intended accounts (i.e., we want to see if the higher value scores are reflected in the pipeline generated). The high percentage makeup of '90-100' scored accounts indicates a good model but will be improved upon based on the current data collected.

# Check if model is properly capturing accounts (cont.)



In the case that the model is inaccurate for a particular team, we look at each factor individually and adjust accordingly.

# Check if model is properly capturing accounts (cont.)

