

TCP / ICP

Methodology
And
Execution Playbook

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Key Personas

Key business personas this playbook is intended for...

Responsible / Accountable



Sales Ops Leader

- **Model Development**-Manages the development of segmentation constructs, coordinating resources both internally and externally
- **Execution** - Responsible for deploying the outputs of segmentation across a variety of dependent workstreams
- **Sustainability**- Responsible for model stewardship and ensuring its sustainability and reliability



Sales Leader (CRO)

- **Revenue Growth** - Focuses on meeting growing revenue targets by optimizing finite resources across market opportunity
- **Productivity** - Proper customer segmentation aligns sales resources to market opportunity likely driving increased per rep productivity
- **Outperformance** - Targets outperforming competitors and growing faster than the industry standard

Consulted / Informed



CFO

- **Planning and Measurement**- Establishes defensible and attainable goals supported by bottoms-up segmentation outputs



Marketing Leader

- **Market Capture** - Seeks to maximize market share and outpace competitors' growth via superior market awareness and customer experience
- **Collaboration** - Aims to collaborate effectively with Sales team and leaders with particular investment in Ideal Customer Profile development

Introduction & Focus

Account Segmentation is the prioritization of accounts across geographies and segments

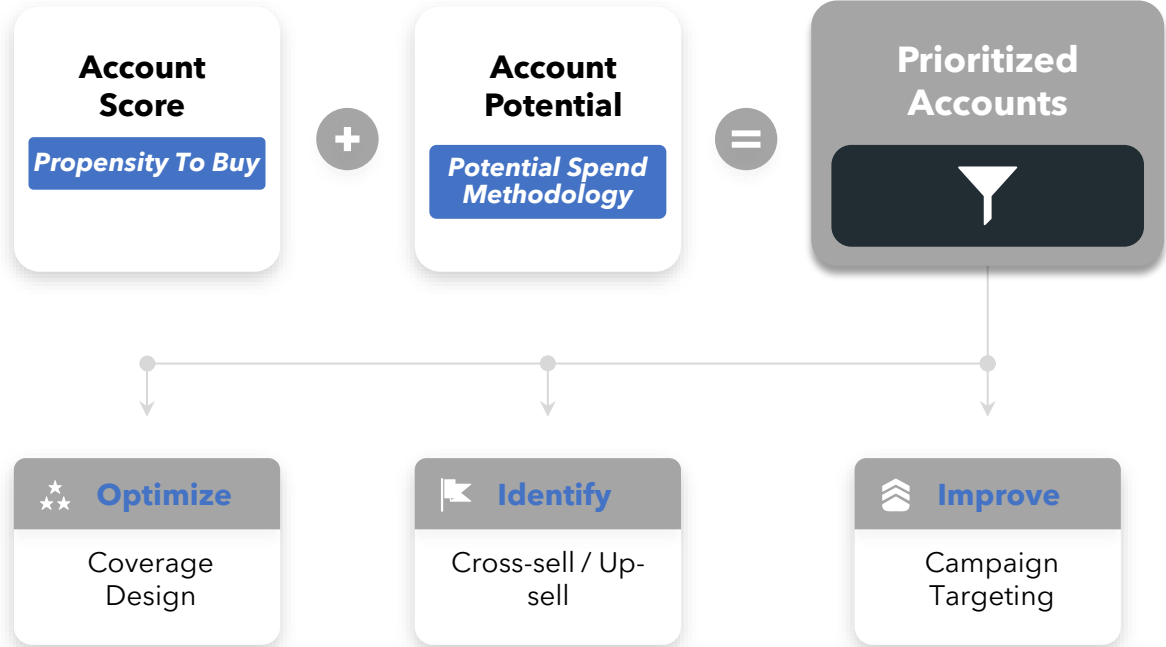


Account Segmentation

Prioritize accounts across geographies and segments based on propensity to buy and size of available opportunity within the account.

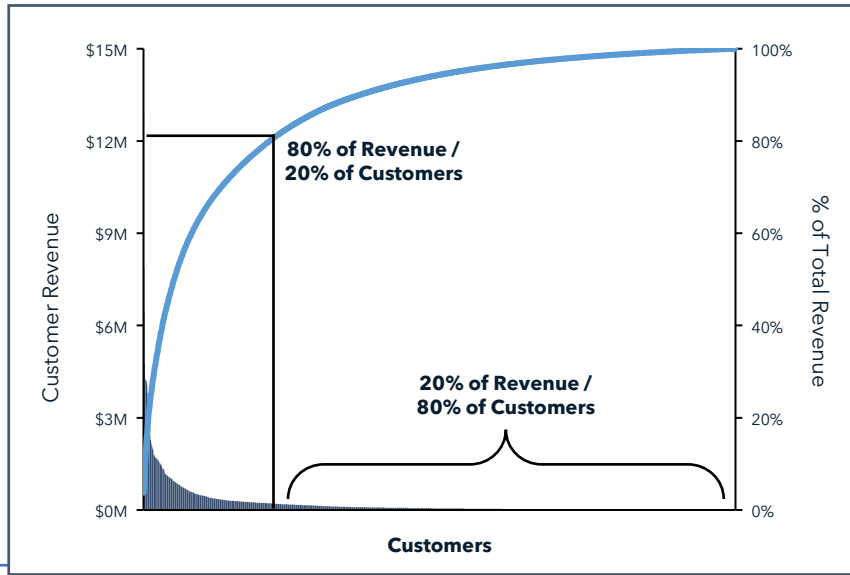
Benefit

Analyze, rank and select the customers that are **likely to spend the most money** in the **shortest period of time**. The output of Account Segmentation creates more clarity for strategic decisions in Coverage, Resourcing, Territory Design, Compensation/Quota Setting, and Hiring.



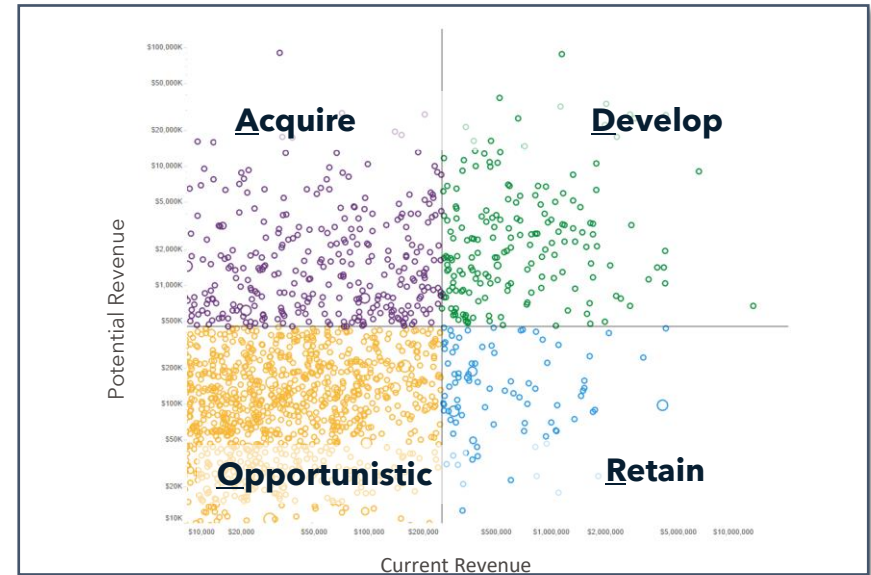
Account Segmentation informs how to divide the market and extract the highest ROI from each segment

Organizations tend to follow the Pareto Principle,
where most of a firm's revenue comes from a small subset of their customers



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Account Segmentation yields higher returns than a one-size-fits-all approach by dividing the market based on current spend, potential value, and ease of acquisition



Best in class Account Segmentation positively impacts selling time, sales productivity, and LTV:CAC Ratio



Why is Account Segmentation Important?

Account Segmentation is important so a company can allocate its resources most effectively to capture market opportunity.

There are three primary reasons to conduct an Account Segmentation exercise:

- 1. Strategy:** Understand who is buying your solution(s), so you can hone your GTM strategy to target the right market, accounts and buyers.
- 2. Prioritization:** Companies prioritize the customers and prospects with greatest potential spend
- 3. Allocation:** Companies allocate their sales, marketing, and customer success resources towards the accounts with the greatest



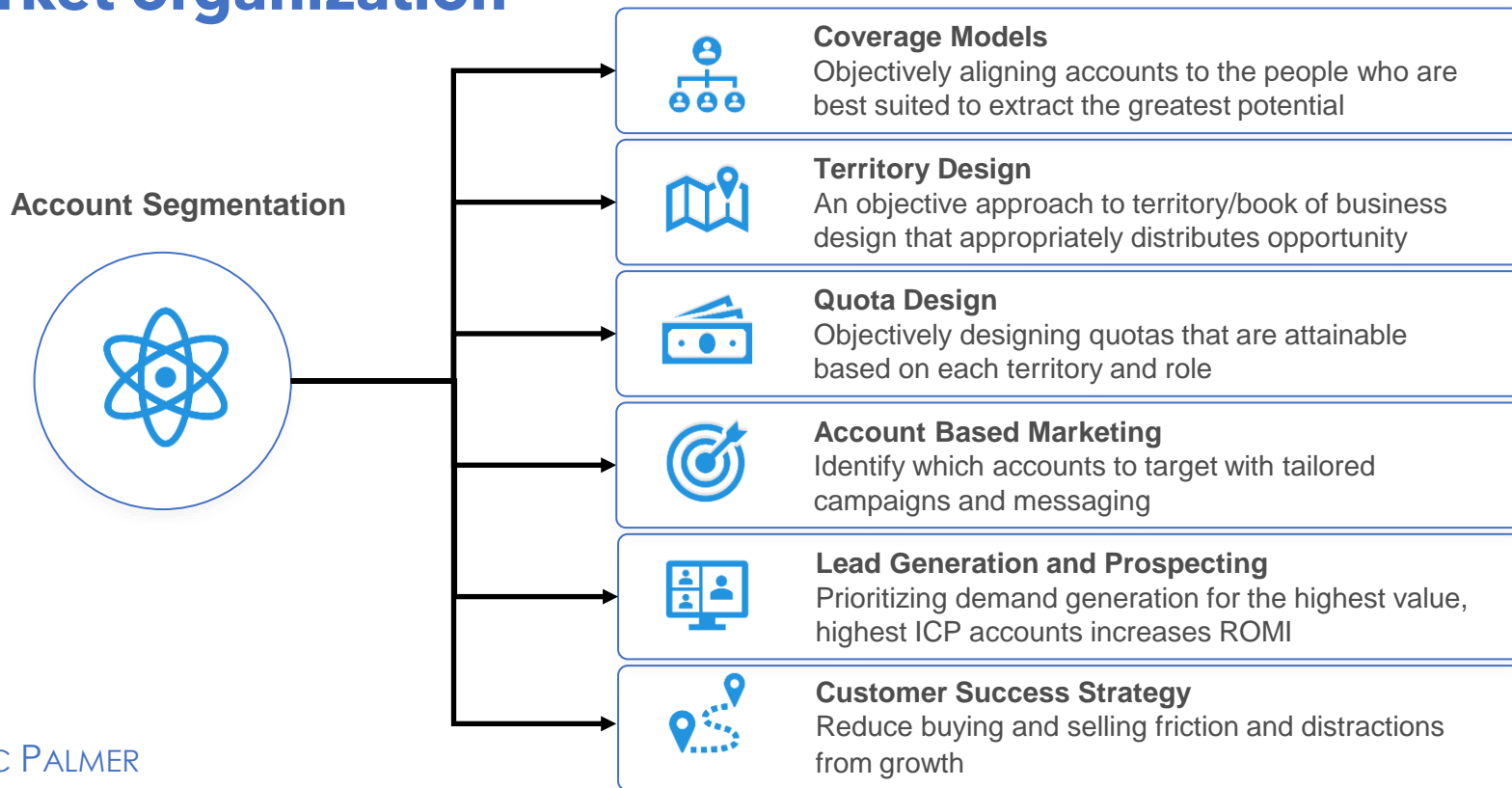
Key Metrics Impacted

Account Segmentation positively impacts three (3) core metrics. They are:

- 1. Selling Time:** Account Segmentation increase selling time by removing the burden of figuring out “which accounts to call on first”
- 2. Sales Productivity:** Account Segmentation improves the productivity of a sales reps and supports them in generating greater returns (faster sales cycle, larger deals)
- 3. LTV:CAC Ratio:** Segmentation ensures that the sales team is spending time on the accounts that have the greatest potential and propensity to buy

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Account Segmentation is the most foundational exercise a company can do, informing multiple workstreams in the Go-to-Market organization



Six common events/circumstances may signal a need to develop and/or revisit an organization's

| Circumstance / Event | Context | Rationale |
|--|---|---|
| Product Lifecycle Milestones (End of Life/ NPI) | New product(s) launched or planned for obsolescence | Shifts in product portfolio influence changes in account potential and subsequent prioritization |
| M&A / Divestiture | Acquisition and integration of a company with a different customer base and/or product set, or divestiture of products or business unit | Acquired accounts and potential associated with new/divested products alter the composition and profile of the account base |
| Unexpected Commercial Outcomes | "Top" growth accounts not providing revenue lift anticipated | Accounts categorized and prioritized incorrectly could lead to misallocation of resources and gap between expectations and outcomes |
| Varied prioritization frameworks deployed | Sellers deploy a variety of different approaches to prioritizing accounts | Standardized approach to segmentation and prioritization ensures consistency and alignment to GTM strategy and plan |
| Assumed Equality of Accounts | Lack of account prioritization framework or an approach where all accounts are equals | Optimization most often requires a differentiated sales motion and level of enterprise investment based on account characteristics |
| Singular Retrospective Lens | Prioritization is solely based on customers' historical revenue contribution | An account's upside potential is a critical dimension in segmentation |

Best in Class segmentation is characterized by a data-driven, scalable, repeatable process

Best in Class Account Segmentation

01

Developed cross functionally with stakeholder involvement in design, validation, and activation

02

Rooted in fact by leveraging account-level actuals in modeling and analysis

03

Purpose Built & Scalable: the models are activated across core GTM workstreams driving decisions and alignment

04

Continually improves and adapts to the business and use cases

05

Provides clarity in distinctions between accounts

06

Treated as a strategic asset with governance and master data management considerations

07

Granularity and accuracy thresholds dictated by use cases and cost/benefit tradeoffs

Signs of Suboptimal Account Segmentation

01

Resources are not focused on the best accounts, impacting their yield and productivity

02

Sales and Marketing functions are misaligned on where to focus resources

03

Sales coverage plans are misaligned with where the market opportunity is

04

Sales team calling on customers and prospects with low-to-no propensity to buy

05

Marketing campaigns targeting the wrong segments, industries and account types

06

There is no single "source of truth" dataset to go for strategic GTM decisions

Leading Practices

Account Segmentation methodology looks to prioritize accounts by utilizing an Account Score and Account Potential



"Who should we pursue?"

Account Score



Account Scoring Methodology

1. Enrich company data with additional 3rd party data (e.g., firmographic, technographic)
2. Understand which factors are correlated to commercial outcomes
3. Utilize working sessions to validate Ideal Customer Profile(s) and add gain additional perspective
4. Score each account against the final Ideal Customer Profile(s)

&



"How much can they spend?"

Account Potential



Potential Spend Methodology

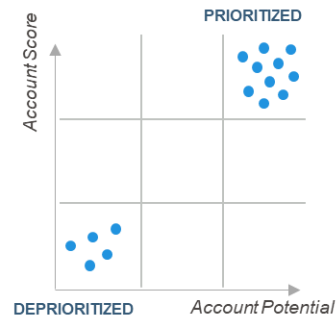
1. Determine approach to calculate account potential
2. Calculate account potential and whitespace for each account

=

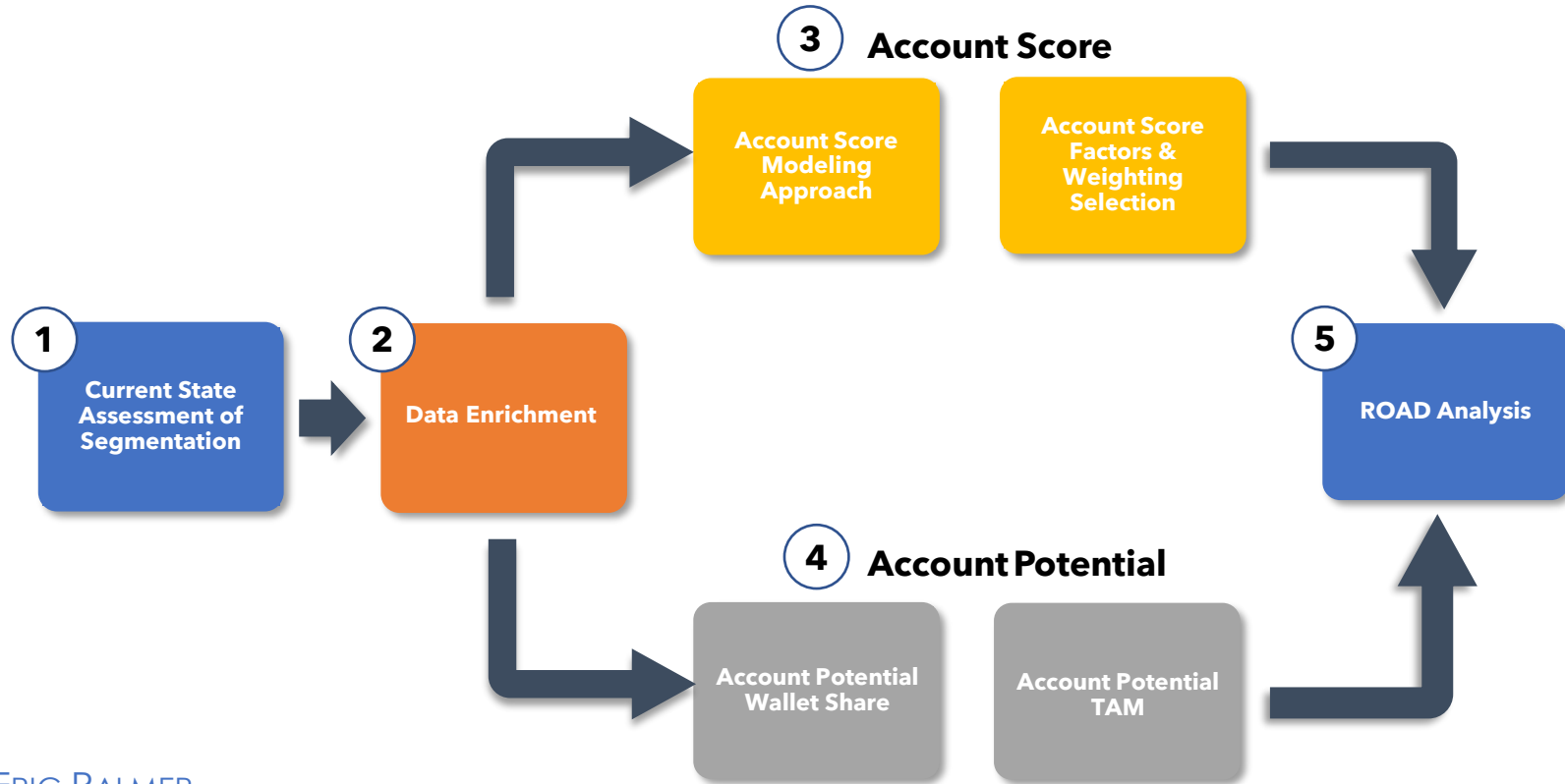


"Who is most attractive?"

Prioritized Accounts



To accomplish this prioritization of accounts, Account Segmentation follows a 5-step process



Establishing a perspective on the current state of Account Segmentation will drive solution design and surface barriers to activation

8 questions form the foundational fact base necessary for an effective Account Segmentation process:

1. How do you currently segment accounts?
2. How do you prioritize accounts?
3. How are Buying Centers defined (i.e., what's the definition of an 'Account')?
4. What are the top 2-3 things you're looking to use segmentation for?
5. Does the customer lifecycle change based on the customer and/or product/service?
6. What are the key dimensions you use to differentiate customers?
7. Are there any major differences in types of customers in terms of business models, product usage, etc that would necessitate a different type of coverage or influence prioritization?

8. What are the firmographic factors that you believe highlight the Ideal Customer Profile (ICP)?

| Current State Assessment | Data Enrichment | Account Score | Account Potential | ROAD Analysis |
|--------------------------|-----------------|---------------|-------------------|---------------|
|--------------------------|-----------------|---------------|-------------------|---------------|

Best in class Account Segmentation is data-driven and begins with five core inputs

Data-driven Account Segmentation relies on multiple inputs to set targets that drive maximum efficiency on the way to an annual sales objective

Ideal Customer Profile (ICP)



Identifies the characteristics of the best-fit customer for a product

Total Addressable Market (TAM)



A top-down view of the market opportunity by industry, product and/or geography

Historical Customer Base



3-year view of customer spend information with as much firmographic detail as possible

Prospect Base



View of who the prospects (non-current customers) are and who is targeting them

Opportunity Data & Win/Loss



Current pipeline, and win/loss history for A/B testing

| | | | | |
|--------------------------|-----------------|---------------|-------------------|---------------|
| Current State Assessment | Data Enrichment | Account Score | Account Potential | ROAD Analysis |
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Creating an ICP is the first crucial step of building an Account Score, as this simply applies a numeric value based on the ICP

What is an Ideal Customer Profile “ICP”?

- 1 A firmographic profile of an account using where the company typically wins or performs well
- 2 Companies may have multiple ICPs if they compete in different markets with different buyer profiles
- 3 ICPs should evolve over time to ensure they accurately represent the target buyer(s) of the company

What is an Account Score?

- 1 Numerical 1 - 100 score for each account which represents how closely the account looks to the ICP
- 2 Higher scoring accounts indicate that the account closely resembles accounts that the company typically wins with
- 3 Account scores are calculated by weighting all the firmographic elements that are part of the ICP

ICP/ Account Score Misconceptions

- 1 **Low Score = Will not Win** - A company may still win in accounts that are low scoring, however the model predicts these as less likely wins
- 2 **High Scoring Accounts will spend the most** - Account score is not a depiction of spend or ability to spend
- 3 **Do not cover low scoring accounts** - Low scores should not mean no coverage but rather varying the coverage type based on the score (i.e., strategic vs. inside)

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An effective scoring model follows balances complexity with reality to make a scalable & repeatable process

| # | Guiding Principles | Principle in Practice |
|---|---|---|
| 1 | Consistent Approach | Creation of a shared approach at the Plan-wide (lexicon, data, logic, etc.) while still enabling business segment customization where necessary |
| 2 | Simplicity Over Complexity | Creating a simple, segment-specific model prioritizing the top accounts will have more of an immediate impact on the company rather than over-indexing on long tail accounts |
| 3 | Balance Forward-Thinking & Historic Trends | Although many of the assumptions built into the segmentation model will incorporate historic trends, we can build a forward-thinking model that prioritizes where the company will want to be in 5 years from now |
| 4 | Think 'Downstream' | Conviction to use segmentation outputs as basis for investments such as coverage, territory, compensation, quota decisioning in the future |

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Account Scores have 4 - 6 components that are used to prioritize Accounts on a point scale of 0 to 100

- 1

Factors
Scores should have 4 to 6 factors
- 2

Factor Weights
No weight should be less than 10% of the score
- 3

Variable Values
No value should represent <5% of the population
- 4

Variable Weights
Band weights should represent band performance within factor

Illustrative Scoring Model:



| Annual Revenue | % of Factor Weight |
|-----------------|--------------------|
| >\$1B | 20% |
| \$1B - \$500M | 80% |
| \$100M - \$500M | 100% |
| \$0-\$100M | 60% |

| IT Spend | % of Factor Weight |
|--------------|--------------------|
| \$25M+ | 100% |
| \$5M - \$25M | 80% |
| \$1M - \$5M | 60% |
| \$0-\$1M | 20% |

| IT Employees | % of Factor Weight |
|--------------|--------------------|
| 100+ | 40% |
| 50 - 100 | 80% |
| 10 - 50 | 100% |
| 0 - 10 | 80% |

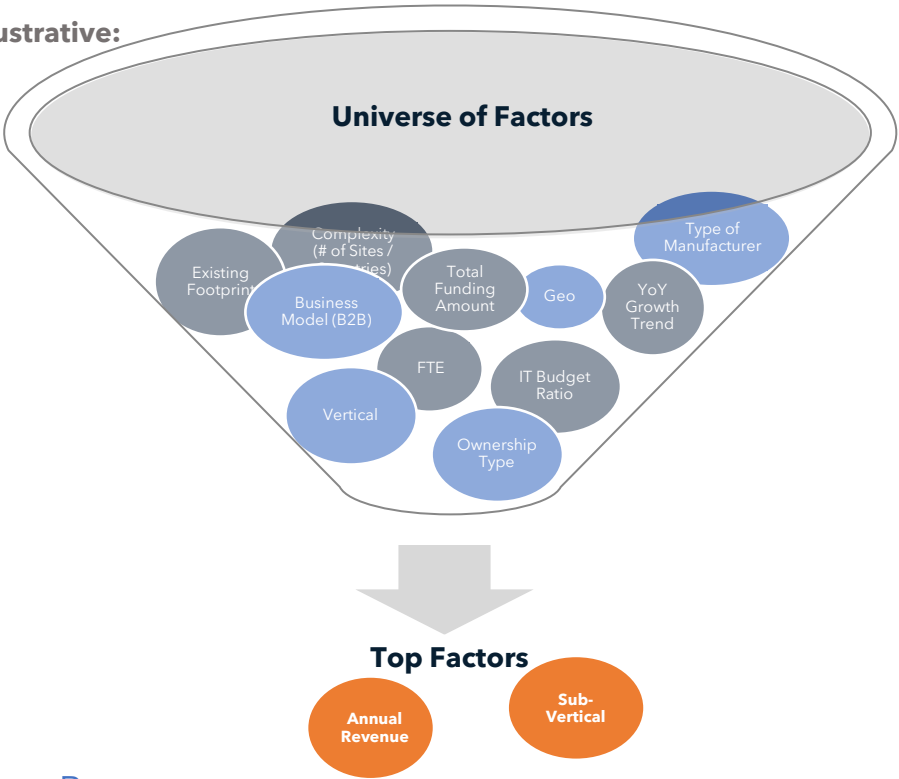
| Number of Locations | % of Factor Weight |
|---------------------|--------------------|
| 1000+ | 80% |
| 500 - 1000 | 100% |
| 100 - 500 | 80% |
| 0 - 100 | 10% |

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Correlation analysis on the universe of factors can produce a list of top factors

Illustrative:




Non-Selected Factors

- **Existing Footprint:** “Leading the witness”, and we would be unable to score prospects against this score
- **Total Funding Amount:** Unreliable data from 3rd parties
- **YoY Growth Trend:** Unreliable data from 3rd parties
- **IT Budget Ratio:** Inadequate ZoomInfo data
- **# of FTE:** Eliminated during expert panel discussion
- **Geo:** Accounted for within potential calculations (frontier analysis)
- **Vertical:** Does not reach the level of granularity required for the model
- **Ownership Type:** There is a slight correlation between public companies spending more, however it was excluded from the 5 factors.
- **Technographic Data:** Part of the account pursuit (not firmographic)
- **Business Model (B2B):** Eliminated during expert panel discussion

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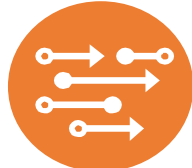
Three checks should be run when selecting key factors: variance, multicollinearity, and population rate



01

Variance

Factors with **different win rates** depending on the factor band



02

Multicollinearity

Factors which are **not related to one another** - or are "telling the same story"



03

Population Rate

Factors with **enough data** to be useful in practice

Illustrative:







| # of Claims | Win Rate |
|-------------|----------|
| 1M+ | 16% |
| 500K – 1M | 30% |
| 100K – 500K | 25% |
| <100K | 11% |

| Account | # of Claims | # of Providers | Correlation |
|---------|-------------|----------------|-------------|
| A | 100K | 10K | .98 |
| B | 1M | 40K | |
| C | 250K | 20K | |

| Factor | % of Accounts Populated |
|----------------------|-------------------------|
| # of Claims | 86% |
| # of Providers | 67% |
| # of Medicare Part D | 40% |

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After factor selection, one of three distinct approaches should be used to produce the scoring model

| Example Client Objective | | Approach | Description |
|---|---|---|--|
| <ul style="list-style-type: none">Prioritize customers based on what our business will look like next year, not last year |  |  Heuristic | Point allocation using qualitative / rule-of-thumb information |
| <ul style="list-style-type: none">Identify prospects we're most likely to win |  |  Classification | Point allocation based on likelihood to achieve a 'Yes / No' outcome |
| <ul style="list-style-type: none">Prioritize accounts that are likely to have the highest spend |  |  Regression | Point allocation based on linear outcome |

Approaches can be mixed and matched; good models have a mix of 'art' and 'science'.



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The Heuristic Scoring Model develops point allocation using qualitative / rule-of-thumb information

Notes

- Dynamic inputs enable experimentation to triangulate on what 'feels right' to the client in terms of....

- A** Factors included in the model
- B** Factor Weights
- C** Variable Values
- D** Variable Weights

- Including relevant data helps contextualize point values to the client, in this case....

- E** How scoring adjustments impact a picklist of relevant customers & prospects
- F** Conversion rates for the 37 sales cycles that have included the new product

Illustrative: Heuristic Scoring Model for a New Product (SD-WAN)

| SD-WAN Scoring Model | | | | | | | | | | | |
|----------------------|--------|--------|----------|---|------------------|------------------|---------------|--------------|-----------------|---------------------|---------------------|
| Calculation | | | | E | Customer | Annual Revenue | IT Spend | IT Employees | Employees | Age of the Business | Number of Locations |
| Manual Input | | | | | Circle K | \$21,000,000,000 | \$420,000,000 | 1700 | 100000 | 69 | 15000 |
| Not in Use | | | | | Total Score: 70 | 10 | 30 | 6 | 0 | 0 | 24 |
| | A | B | | | | | | F | D | | |
| Factor | In Use | Weight | Weight % | C | Annual Revenue | Opportunities | Wins | Win % | Variable Weight | % of Potential | Notes |
| Annual Revenue | Yes | 25 | 25% | | \$ 1,000,000,000 | 35 | 15 | 43% | 2 | 40% | |
| IT Spend | Yes | 30 | 30% | | \$ 500,000,000 | 1 | 0 | 0% | 4 | 80% | |
| IT Employees | Yes | 15 | 15% | | \$ 100,000,000 | 1 | 0 | 0% | 5 | 100% | |
| Employees | No | 0 | 0% | | \$ - | 1 | 1 | 100% | 3 | 60% | |
| Age of the Business | No | 0 | 0% | | | | | | | | |
| Number of Locations | Yes | 30 | 30% | | | | | | | | |
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| Current State Assessment | Data Enrichment | Account Score | Account Potential | ROAD Analysis |
|--------------------------|-----------------|---------------|-------------------|---------------|

For Classification and Regression Models, factors with the strongest relationship to an organization's objective will be selected...

Identify breakpoints within each of the factors where customers start to perform with the highest level of dissimilarity.
*Segmentation is about separating accounts into cohorts that **perform differently**.*

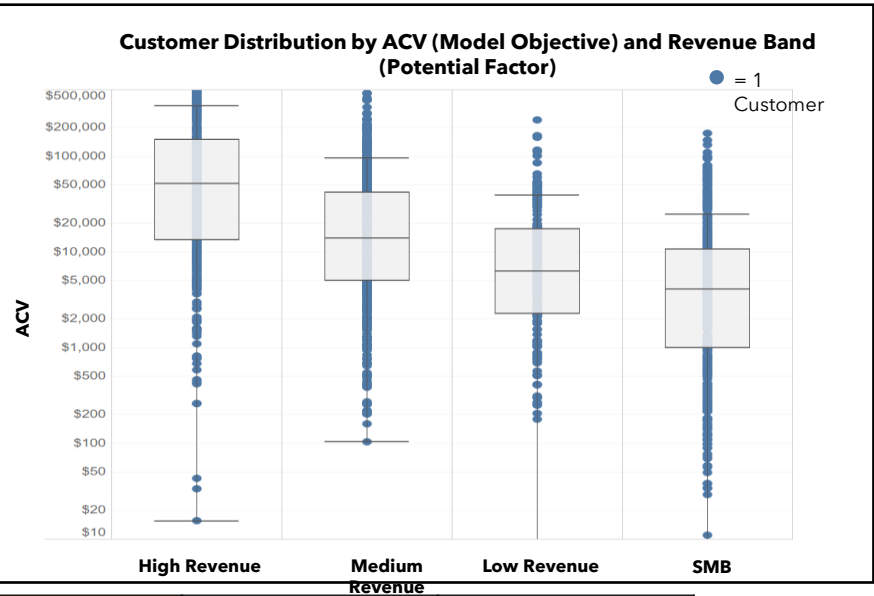
Example 1: Classification

Evaluating Win Rates by Industry we can see the client has a higher ease of acquisition in specific Verticals

| Win Rate (Model Objective) by Industry (Potential Factor) | |
|--|----------|
| Industry | Win Rate |
| Finance | 44% |
| Professional Services | 42% |
| Insurance | 37% |
| Chemicals and Petroleum | 31% |
| Retail | 24% |
| CPG | 22% |
| Healthcare | 21% |
| Manufacturing | 17% |
| Media and Entertainment | 15% |
| Government | 12% |

Example 2: Regression

Using Whisker Charts to show buying distribution by Revenue Band, we can see that Annual Revenue is correlated with Customer ACV and Customers in each Revenue Band buy differently than one another



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... before variable weights for relevant factors are determined using either Descriptive Analysis (Option 1)...

Example 1: Classification

Determine variable weights for each factor's cohorts by dividing the success rate (data points that have hit the client's objective) per cohort by the cohort with the max success rate

| Factor Cohort | Successes | Successes + Failures | Success % | % of Factor Weight |
|---------------|-----------|----------------------|-----------|--------------------|
| A | 50 | 120 | 42% | 100% |
| B | 189 | 610 | 31% | 74% |
| C | 65 | 271 | 24% | 57% |
| D | 85 | 386 | 22% | 52% |
| E | 120 | 801 | 15% | 36% |

15%

Success %

÷

42%

Max Success %

=

36%

% of Factor Weight

Example 2: Regression

Determine variable weights for each factor's cohorts based on the ratio of customers in the top 20th percentile (e.g. customer ACV) vs the bottom 20th percentile compared to the cohort with the highest ratio

| Factor Cohort | # in top 20 th Percentile | # in bottom 20 th Percentile | Ratio | % of Factor Weight |
|---------------|--------------------------------------|---|-------|--------------------|
| A | 137 | 29 | 4.7 | 100% |
| B | 42 | 20 | 2.1 | 44% |
| C | 83 | 63 | 1.3 | 28% |
| D | 19 | 34 | 0.6 | 12% |
| E | 83 | 219 | 0.4 | 8% |

0.4

Ratio

÷

4.7

Max Ratio

=

8%

% of Factor Weight

| | | | | |
|--------------------------|-----------------|---------------|-------------------|---------------|
| Current State Assessment | Data Enrichment | Account Score | Account Potential | ROAD Analysis |
|--------------------------|-----------------|---------------|-------------------|---------------|

... or Predictive Analytics (Option 2)

Alteryx makes predictive analytics accessible to the everyday business user. Using point-and-click tools in a guided interface, building a predictive classification or regression model is simple and easy

Just select which factors to use...

| <input checked="" type="checkbox"/> | Feature | Feature Info |
|-------------------------------------|----------------------|---|
| <input checked="" type="checkbox"/> | PSD Segment | <input checked="" type="checkbox"/> This feature is a good predictor. |
| <input checked="" type="checkbox"/> | IT Budget | <input checked="" type="checkbox"/> This feature is a good predictor. |
| <input checked="" type="checkbox"/> | Finance Budget | <input checked="" type="checkbox"/> This feature is a good predictor. |
| <input checked="" type="checkbox"/> | MKTG Budget | <input checked="" type="checkbox"/> This feature is a good predictor. |
| <input checked="" type="checkbox"/> | Revenue | <input checked="" type="checkbox"/> This feature is a good predictor. |
| <input checked="" type="checkbox"/> | Employees | <input checked="" type="checkbox"/> This feature is a good predictor. |
| <input checked="" type="checkbox"/> | 3 YR Employee Growth | <input checked="" type="checkbox"/> This feature is a good predictor. |
| <input checked="" type="checkbox"/> | IT Employees | <input checked="" type="checkbox"/> This feature is a good predictor. |

...which types of model to test...

| | |
|---|--|
| <input checked="" type="checkbox"/> Decision Tree | <input checked="" type="checkbox"/> Logistic Regression |
| <div>Pros<ul style="list-style-type: none">• Easy to interpret.• Built-in feature selection.Cons<ul style="list-style-type: none">• Favors stronger features, ignoring more subtle features.<i>Best use: simple classification problems.</i></div> | <div>Pros<ul style="list-style-type: none">• The linear equation is fairly easy to interpret.• Estimation time is relatively short.Cons<ul style="list-style-type: none">• Limited to only binary classification.• Linear nature of the model has limitations.</div> |
| <input checked="" type="checkbox"/> Random Forest | <input checked="" type="checkbox"/> XGBoost |
| <div>Pros<ul style="list-style-type: none">• Better than a single decision tree at handling imbalanced targets.• Better than a single decision tree at capturing the effects of subtle features.Cons<ul style="list-style-type: none">• Results are more difficult to interpret.</div> | <div>Pros<ul style="list-style-type: none">• Models nonlinear associations.• Is less subject to overfitting and underfitting (even compared to random forest).Cons<ul style="list-style-type: none">• Approximates linear associations.<i>Is an expensive computation (even compared to random forest).</i></div> |

...and add the most performant model to the workflow

| Model Overview | |
|-------------------|------------|
| Model | Accuracy ↓ |
| ● Random Forest 1 | 68.9% |
| ● Random Forest 2 | 68.9% |
| ● XGBoost 1 | 67.9% |
| ● XGBoost 2 | 67.9% |
| ● Decision Tree 1 | 64.2% |

| | | | | |
|--------------------------|-----------------|---------------|-------------------|---------------|
| Current State Assessment | Data Enrichment | Account Score | Account Potential | ROAD Analysis |
|--------------------------|-----------------|---------------|-------------------|---------------|

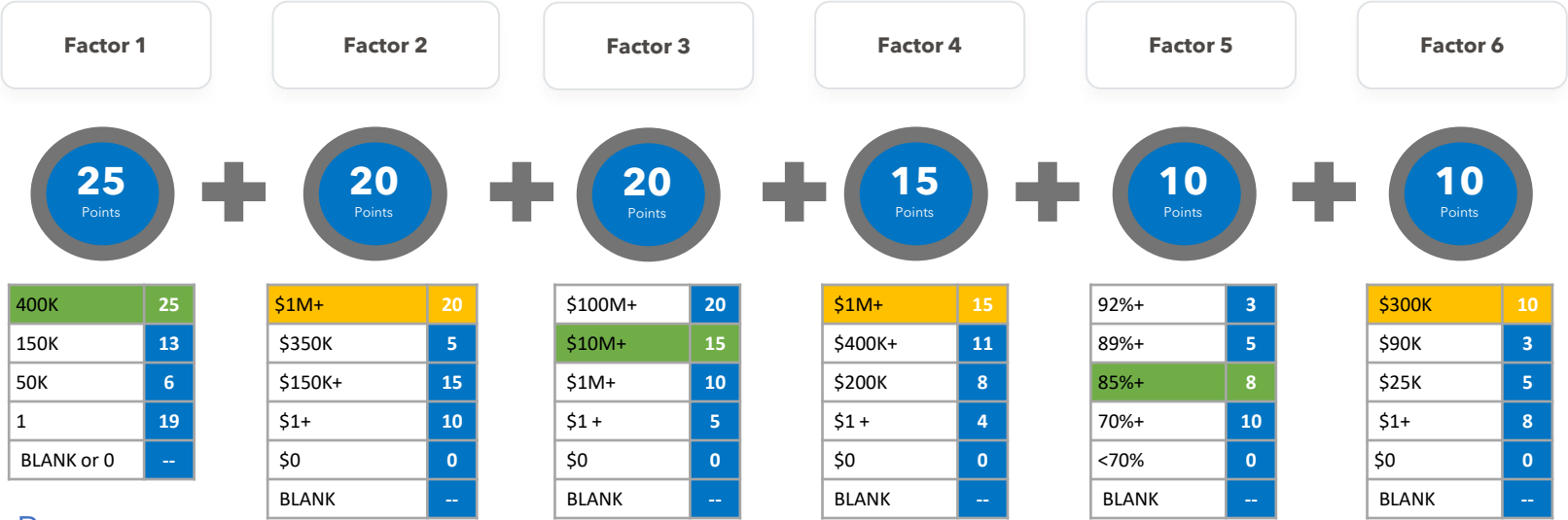
The key output of Account Scoring runs all accounts through the model and assigns numeric scores

Illustrative:

Account A

- Factor 1: 1.7M
- Factor 2: \$3M
- Factor 3: \$39M
- Factor 4: \$4M
- Factor 5: 85%
- Factor 6: \$700K

= 25 + 20 + 15 + 15 + 8 + 10 = 93



| | | | | |
|--------------------------|-----------------|---------------|-------------------|---------------|
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Before scores are finalized, select accounts should be reviewed to ensure scores are optimized for Accuracy, Precision, and Recall

Using metrics to describe the model's performance along with a comparison to the client's current business process (e.g. is the score better at prioritizing an account than sellers are) adds confidence in the efficacy of the model

Apply Scores against historical data and use a confusion matrix to determine how predictive your scoring model is against actual results

- A Score greater than ≥ 50 can be used to identify a 'Positive' in the Score
- For regression models, identify a cutoff (e.g. \$50K in ACV) for 'Positives' in the historical data

| | |
|--|--|
| True Negative: Low Score / Bad Outcome  | False Positive: High Score / Bad Outcome  |
| False Negative: Low Score / Good Outcome  | True Positive: High Score / Good Outcome  |

Balancing Scores based on Accuracy, Precision, and Recall ensures the scoring model prioritizes Accounts with the right quality and quantity

| Metric | Calculation | What it tells us |
|-----------|--|--|
| Accuracy | $\frac{\text{True Positives} + \text{True Negatives}}{\text{All Results}}$ | How often does the score get the right outcome <i>How often do we correctly identify "Dogs" and "Not Dogs"?</i> |
| | | |
| Precision | $\frac{\text{True Positives}}{\text{True Positives} + \text{False Positives}}$ | Are prioritized accounts the right quality <i>Are we excluding as many cats as possible?</i> |
| Recall | $\frac{\text{True Positives}}{\text{True Positives} + \text{False Negatives}}$ | Are we prioritizing enough of the high value accounts <i>Are we getting as many dogs as possible?</i> |

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|--------------------------|-----------------|---------------|-------------------|---------------|
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Account scores should be reviewed and updated on an annual or biannual basis to account for business or account changes

| | Reason | Actions to Take |
|------------------------------|---|---|
| Refresh Factor Data | Factor values for accounts may change based on the frequency of source updates. Accounts that may have been null for a factor may receive a value during refresh. New data may be purchased by the company to fill certain factors. | <ul style="list-style-type: none">• Refresh source files of factors to pull in latest information• Rerun scoring and compare |
| Review Factor Correlation | Factor correlation to win rate or spend may change over time. New factors may also be come available. Factors used in scoring should be the most correlated factors that pass the Variance, Fill Rate and Multicollinearity test | <ul style="list-style-type: none">• Review that the factors used in scoring remain highly correlated to spend• Assess if any factors not used in scoring become highly correlated to spend |
| Assess New Business Strategy | Since scores are built on both the art and the science, there may be strategic changes to the business that affect the ideal customer being targeted. | <ul style="list-style-type: none">• Review any changes in business strategy and assess if a new ICP should be created or modified to remain aligned |
| Review Score Weightings | Since factors may change in correlation, and performance within factor bands may change, score weightings should be reviewed. | <ul style="list-style-type: none">• Review changes in correlation and band performance• Adjust factor weightings to align |

| | | | | |
|--------------------------|-----------------|---------------|-------------------|---------------|
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There are two effective approaches to calculating Account Potential:

Wallet Share Approach

Total Addressable Market (TAM) Approach

Wallet Share Approach
(i.e. Performance Frontier)

Account Potential is calculated based on historical top customer spend as a percentage of their financials (i.e. Wallet Share). This is the most common methodology for calculating potential and is revenue model agnostic.

- ✓

Easy to Calculate
- ✓

Based on Historical Performance
- ✓

Accounts for platform solutions / multi-product
- ✗

Doesn't account for underperformance in top customers



To calculate Account Potential, TIBCO simply multiplies the annual revenue of their other customers by 0.005% to estimate the total amount they could spend on TIBCO products.

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| | | | | |
|--------------------------|-----------------|---------------|-------------------|---------------|
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Account potential is calculated based on the organization's pricing model and total potential product consumption at an account level. This approach is more common with user-based revenue models.

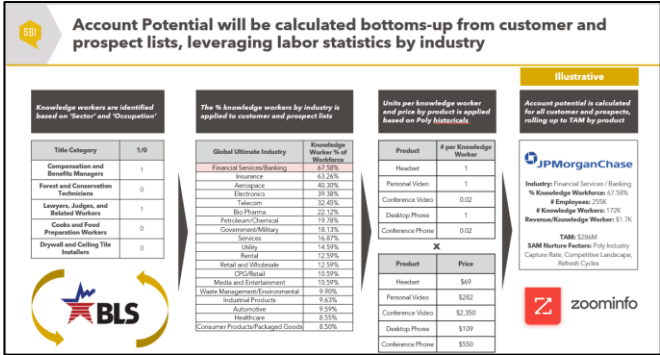
- ✓

Simple and Defendable
- ✓

Good for user-based pricing models
- ✗

Assumes current product adoption is nascent
- ✗

Difficult to create / time intensive



To calculate Account Potential, Poly estimated the number of Office Workers at each account and multiplied the Office Worker count by the price per desk phone

The Wallet Share Approach identifies the customer measure most directly correlated to spend to calculate the Performance Frontier...

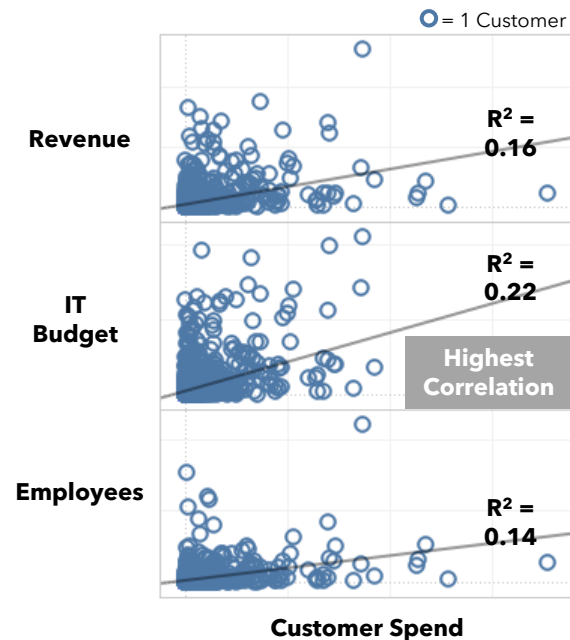
Identify the metrics with the greatest correlation to Customer Spend

Identify Customer Peer Groups

Identify 'like' peer groups of customers with similar attributes that would exhibit different buying behaviors

Calculate Performance Frontier

For each Peer Group, calculate the top 80th percentile of Customer Spend as a % of the selected measure



Potential Dimensions

Theatre

Customer Type

Industry

Revenue Band

Product Category

If peer groups have an N count below 50, consider limiting the number of dimensions in consideration or consolidate dimension values (e.g. by grouping industries)

Peer Group

80th Percentile of Customer Spend / IT Budget

AMER: CPG

2.5%

EMEA: CPG

10.1%

APAC: CPG

4.8%

AMER: Oil & Gas

3.5%

EMEA: Oil & Gas

1.2%

APAC: Oil & Gas

7.2%

AMER: Pharma

5.4%

EMEA: Pharma

3.4%

APAC: Pharma

2.7%

... and applies the Performance Frontier Benchmark per cohort to the account base to calculate potential Spend




| Customer | Performance Frontier | IT Budget | Potential Spend |
|--------------|----------------------|-----------------|-----------------|
| Customer #1 | 3.25% | \$180,000,000 | \$5,850,000 |
| Customer #2 | 3.25% | \$51,500,000 | \$1,673,750 |
| Customer #3 | 3.25% | \$54,500,000 | \$1,771,250 |
| Customer #4 | 3.25% | \$126,000,000 | \$4,095,000 |
| Customer #5 | 3.25% | \$20,000,000 | \$650,000 |
| Customer #6 | 3.25% | \$67,300,000 | \$2,187,250 |
| Customer #7 | 3.25% | \$1,067,000,000 | \$34,677,500 |
| Customer #8 | 3.25% | \$50,000,000 | \$1,625,000 |
| Customer #9 | 3.25% | \$50,000,000 | \$1,625,000 |
| Customer #10 | 3.25% | \$305,000,000 | \$9,912,500 |

| Cohort |
|---|
| Geo = AMER Industry = Consumer Goods Revenue Band = \$1B+ |

$$\begin{array}{ccccc} \text{Performance} & & & & \text{Potential} \\ \text{Frontier} & \times & \text{IT Budget} & = & \text{Spend} \end{array}$$

| | | | | |
|--------------------------|-----------------|---------------|-------------------|---------------|
| Current State Assessment | Data Enrichment | Account Score | Account Potential | ROAD Analysis |
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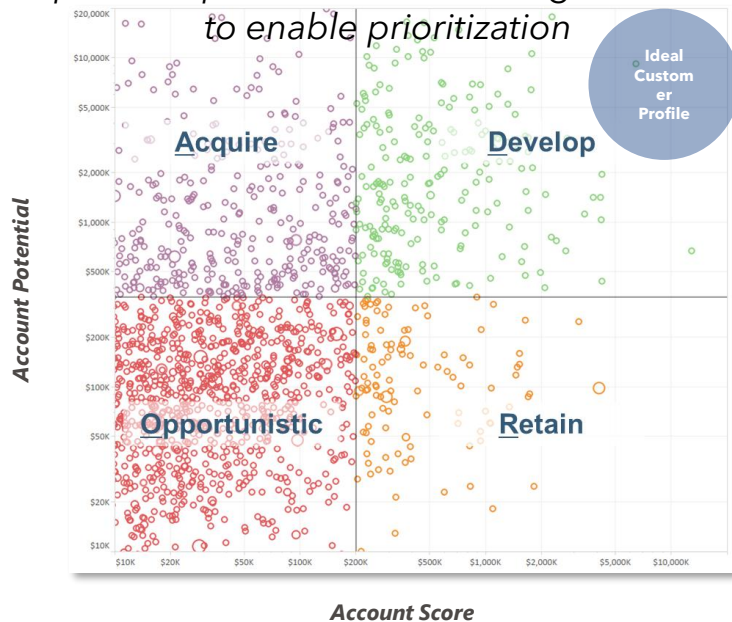
The TAM Approach begins by identifying a metric correlated to potential Users, then selecting a source of truth before calculating spend

| Product Example | User Metric | Data Source | TAM Calculation |
|------------------------|---------------------|--|--|
| Office Phones | # of Office Workers |  BLS | $\begin{aligned} &\text{Industry Office Worker \% of Workforce} \\ &\times \\ &\text{Employee \#} \\ &\times \\ &\text{Price per Phone} \\ &\times \\ &\text{Phone Refresh Cycle} \end{aligned}$ |
| Desktop BI Product | # of Analysts |  zoominfo | $\begin{aligned} &\# \text{ of Analysts at the account} \\ &\times \\ &\text{Price per User} \\ &\times \\ &\text{Discount Curve} \end{aligned}$ |
| eLearning for Colleges | # of Students |  IPEDS | $\begin{aligned} &\# \text{ of Students} \\ &\times \\ &\text{Units per Student} \\ &\times \\ &\text{Price per Unit} \end{aligned}$ |

| | | | | |
|--------------------------|-----------------|---------------|-------------------|---------------|
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The output of Account Segmentation informs the coverage motion and identifies which accounts sellers should target via the ROAD Model

Account Score can be paired with potential spend outputs from Account Segmentation to enable prioritization

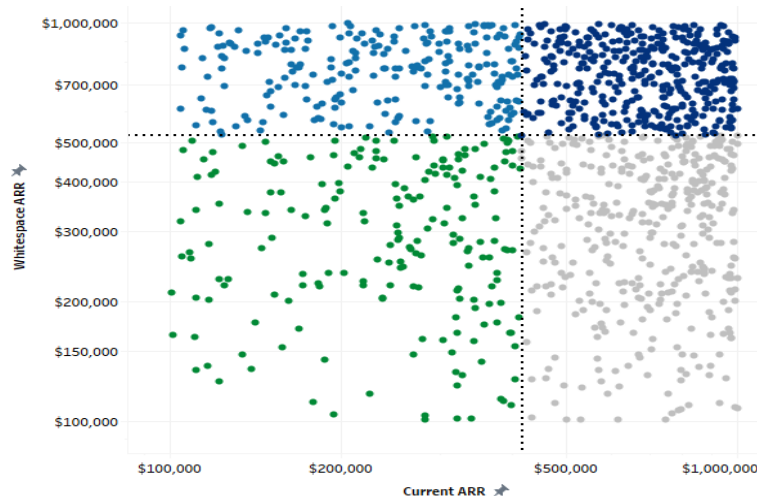


| ROAD Quadrant | General Definitions |
|------------------------|---|
| <u>R</u> etain | Highly penetrated customers with limited opportunity to expand / grow / develop (goal is for nurture & renew) |
| <u>O</u> ppportunistic | Low-touch, opportunistic conversion of low penetration, low value customers |
| <u>A</u> cquire | High value, no / low penetrated prospects that drives hunt / expand motions |
| <u>D</u> evelop | High value, less-than-optimal penetration customers with expansion opportunities |

| | | | | |
|--------------------------|-----------------|---------------|-------------------|---------------|
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Detailed, account-level data allows segments to be defined & aligned to strategic objectives and resource levels

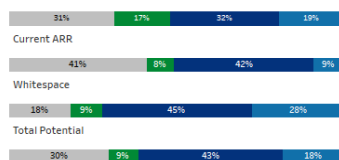
Illustrative:



Distributions

| | Distinct Accounts | Current ARR | Whitepace ARR | Total Potential |
|---------------|-------------------|-------------|---------------|-----------------|
| Develop | 389 | \$275M | \$290M | \$567M |
| Retain | 378 | \$271M | \$121M | \$391M |
| Acquire | 284 | \$60M | \$181M | \$241M |
| Opportunistic | 200 | \$52M | \$62M | \$114M |
| Total | 1,251 | \$657M | \$656M | \$1,312M |

Accounts



How do you use it?

- **Resource Planning & Forecasting:** Budgeting and targeting based on revenue potential and headcount coverage needs
- **Sales Strategy:** Determining where to play and how to win, both Direct and via the Channel
- **Territory Design & Management:** Assigning well-balanced territories based on quantified potential
- **Compensation & Quota Setting:** Assigning individual goals and incentives based on achievable targets
- **Marketing Strategy:** Determining, designing and activating the right set of activities based on where they are likely to generate the best results

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Current State
Assessment

Data Enrichment

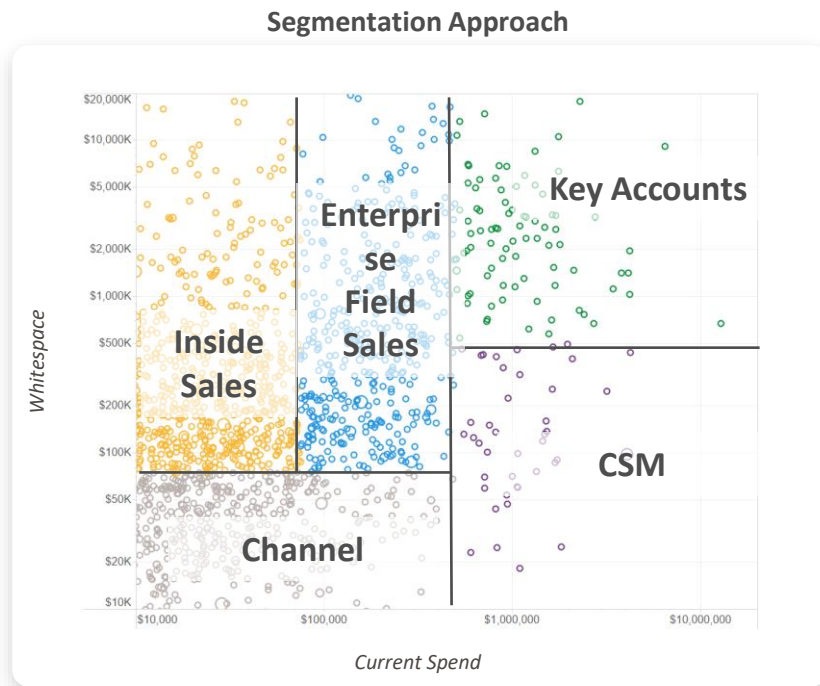
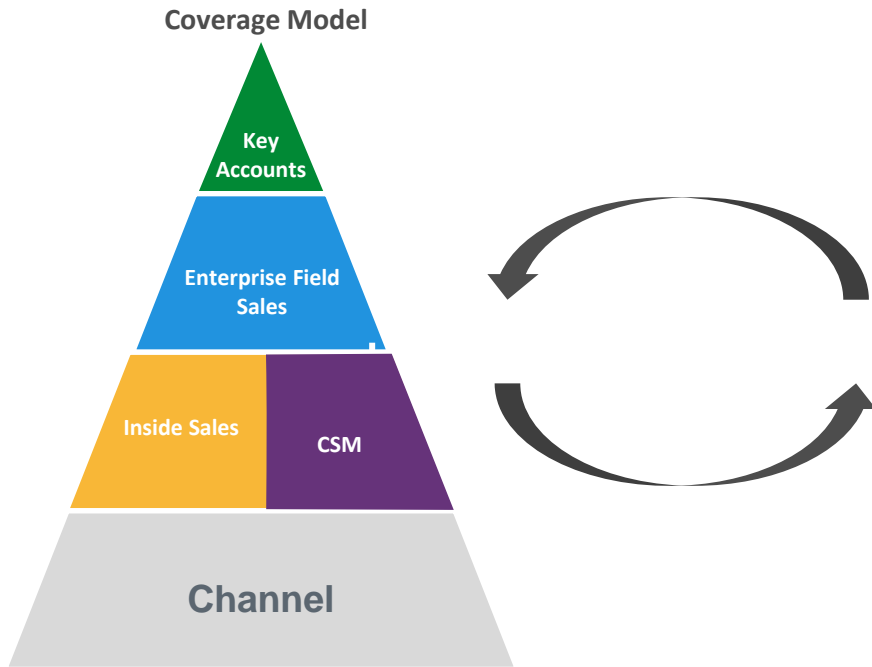
Account Score

Account Potential

ROAD Analysis

The ROAD frame allows for iterations of expansion and/or simplification based on future coverage models and resource allocations each fiscal year

Illustrative:



| | | | | |
|--------------------------|-----------------|---------------|-------------------|---------------|
| Current State Assessment | Data Enrichment | Account Score | Account Potential | ROAD Analysis |
|--------------------------|-----------------|---------------|-------------------|---------------|

Risk Mitigation

Mitigation tactics can prevent risks from impeding incentive structure planning

| | Risk | Severity | Mitigation |
|-------------|--|----------|--|
| Execution | Resources are unable or unwilling to prioritize development of segmentation model resulting in risk of delay | High | <ul style="list-style-type: none"> Develop a project charter and recruit executive sponsorship to facilitate the focus and alignment of the development team |
| | Outputs are not carried through to downstream GTM activities resulting in inefficiencies and misalignment | Medium | <ul style="list-style-type: none"> Socializing the vision and expectations of how and why activation of the model will occur coupled with tracking and monitoring deployment |
| Financial | Unsuccessful segmentation impairs ability to achieve growth expectations | High | <ul style="list-style-type: none"> Stakeholder involvement in model development serves to ensure exercise is built with quality in mind |
| | Segmentation is costly to develop and implement | Low | <ul style="list-style-type: none"> Embedding sustainability and repeatability in the design and production of modeling activities serves to reduce operational cost of model maintenance. |
| Talent | Sales talent is misaligned to market opportunity driving low attainment, morale, and attrition | Medium | <ul style="list-style-type: none"> Aligning talent to best opportunities surfaced during segmentation optimizes seller activity |
| | Lack of framework to prioritize accounts and efforts leads to wasted time, opportunity, and frustration | Medium | <ul style="list-style-type: none"> Enablement materials and proactive change management prepare and equip sellers to benefit as much as possible from segmentation |
| Operational | Downstream GTM planning activities rely on the completion of segmentation to fully execute | Medium | <ul style="list-style-type: none"> Careful planning prior to the annual revenue planning exercise should consider the sequencing and dependencies of related areas |
| | Front line managers do not understand how segmentation outputs can and should drive behavior and outcomes, resulting in lost opportunities | Medium | <ul style="list-style-type: none"> Training materials and expectation clarity support consistency in adoption and alignment across the sales org |

Execution Plan & KPIs

We follow a four-step process in segmentation model development



The following represent metrics & KPIs to track performance and progress

| | | Metric or KPI | Detail | Additional views |
|-----------------------|----|------------------------------|---|------------------|
| Lagging Metrics | 1 | Gross Revenue Retention Rate | Percentage of dollars recaptured (ARR less churn) based on prior years customer base and ARR | Region & Segment |
| | 2 | Net Revenue Retention Rate | Percentage of dollars recaptured and upsold (ARR less churn plus upsell) based on prior years customer base and ARR | Region & Segment |
| | 3 | Seller Attrition | Percentage of sellers departing voluntarily | Region & Segment |
| | 4 | Quota Attainment | % Quota achieved during time period | Region & Segment |
| | 5 | Bookings | Average annual bookings per account and seller | Region & Segment |
| Behavioral Indicators | 6 | Account Engagement | # times key accounts are engaged during the year | Region & Segment |
| | 7 | CSAT Score | Customer ranking of how satisfied they are (1-5 scale) | Region & Segment |
| | 8 | Focused Accounts | % selling time aligned to top accounts in territory | Region & Segment |
| | 9 | Adoption | % sellers leveraging segmentation data in territory planning | Region & Segment |
| Leading Insights | 11 | Territory Equity | Percentage of territories with total whitespace or potential within 20% of segment median | Region & Segment |
| | 12 | Territory Balance | % of top Accounts within territory that account for 50% of territory potential | Region & Segment |
| | 13 | Enablement/Training | % sellers that know how to leverage segmentation data in territory and account planning | Region & Segment |