

# US Market Entry Strategy of Plant-Based High Protein Snack Bar

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**Abstract:** The U.S. plant-based high-protein bar category is expanding rapidly, creating a huge opportunity for a mid-sized FMCG firm seeking to enter this space. This report looks at the market entry decision as a structured choice among three alternatives: national launch, regional pilot, and online-first, and dives into the industry assessment with decision analytics. The different market segmentations indicate a positioning gap between extreme “fitness-first” plant-based protein bars and premium organic choices, supporting the concept of an organic, plant-based protein bar delivering 15–20g protein at  $\leq 200$  calories with an appealing taste and texture. Pricing analytics using K-means clustering and PCA suggest targeting the premium segment, with a recommended list price around \$2.63–\$2.75 per bar. Entry-strategy risk is evaluated using an EMV decision tree and a risk-averse MAUT model across six different risk factors (consumer acceptance, supply chain, regulatory compliance, brand positioning, sustainability claims, and price sensitivity). Both EMV and MAUT emphasize that a regional pilot is the preferred strategy, and testing under Dirichlet weight uncertainty confirms this choice. Finally, state-by-channel hotspot analysis supports an offline regional pilot together with targeted online activation in high-performing states, which will enable learning while limiting downside risk.

**Keywords:** Market basket analysis, price strategy, pricing analysis, hotspot analysis, decision tree,

## 1 INTRODUCTION

The U.S. market for plant-based protein bars has very significant growth projections over the next decade and beyond. At the same time, the market has many competitors and concentrated, with several major companies and a long list of smaller brands competing across premium, organic, mid-range, and budget tiers. For a mid-sized FMCG company to enter the market, the strategic opportunity would be to enter with clear product positioning and go-to-market design that can win early adoption while minimizing losses. In particular, competitive mapping suggests an opportunity between extreme fitness positioning and premium-organic positioning, where a bar can compete on both nutrition and consumer experience (such as taste and texture) while maintaining clean-label credibility.

### 1.1 Decision frame and decision alternatives

This report answers one central question: Which U.S. market-entry strategy should be selected for an organic, plant-based, high-protein bar? Three feasible entry alternatives are evaluated: A1 national launch, A2 regional pilot, and A3 online-first. The decision is framed to reflect a realistic managerial trade-off between (i) scaling quickly to capture growth and shelf-space and (ii) controlling the key uncertainties that drive early failure—most notably consumer acceptance (taste/texture) and supply reliability.

### 1.2 Analytical approach and how evidence supports the decision

To connect market context to a defensible entry recommendation, the report integrates four linked analyses. First, a market structure assessment summarizes competitor positioning and clarifies the differentiation logic for a new entrant. Second, a K-means clustering and PCA-based pricing analysis translates

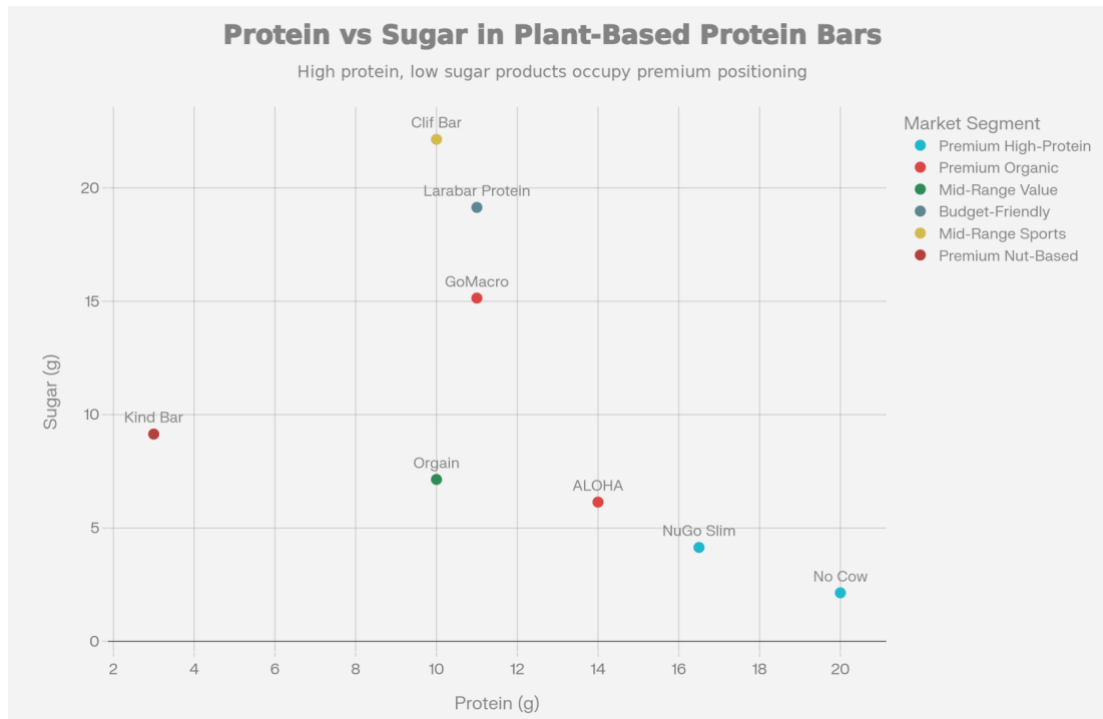
competitive features into interpretable price tiers and a premium positioning anchor, supporting a pricing strategy that signals quality rather than competing in cost-driven segments. Third, a risk-based decision analysis evaluates A1–A3 using an EMV decision-tree structure and a risk-averse MAUT model, ensuring that the recommendation is supported under both expected value and risk-focused preference frames. Finally, a hotspot analysis uses a synthetic transaction-level dataset to visualize state-by-strategy patterns in conversion, ROI, and pricing, providing directional guidance on where the chosen strategy is likely to perform best and where targeted digital activation may add value.

## 2 MARKET OPPORTUNITY & COMPETITIVE LANDSCAPE

The plant-based protein bars market has experienced exponential growth, valued at \$1.05 billion in 2025 and expected to reach \$5.43 billion by 2035. This is a CAGR (compounded annual growth rate) of 17.8%. The USA dominates the plant-based protein bar market due to the existence of a strong fitness culture and adoption of different diets.

The competition boils down to eight primary plant-based companies that hold about 60-70% of the market share. Clif Bar and Company has the highest market share with 18-22% market share, followed by RXBAR with 14-18%, KIND Snacks with 12-16%, LARABAR with 8-12% and GoMacro with 6-10%. There are also many smaller plant-based brands that have also captured a significant market share, some of these brands include No Cow, ALOHA and NuGo Slim. For a mid-sized FMCG firm entering the market, there is an opportunity to capture the market by prioritizing a well-rounded plant-based protein bar that tastes good and has a lower calorie to protein ratio.

Below is a breakdown of the competitors in the market and the sugar vs protein content in their plant-based protein bars.



**Figure 2.1** Protein vs Sugar in Plant-Based Protein Bars

There is a gap between No Cow's extreme fitness positioning and ALOHA's organic premium positioning. A 15-20g plant-based protein bar with 200 calories or less would fill this gap and capture significant market share quickly.

**Taste and Texture Differentiation:** the primary issue consumers face with high protein plant-based bars is the taste and texture. Third party reviews consistently identify taste and texture as the primary weakness in plant-based protein bars. A brand emphasizing great taste with plant-based ingredients and high protein can fill the gap in the market.

**Organic Certification and Sustainability:** Organic protein bars occupy 59.4% of market share. A brand entering the plant-based protein bar market should consider USDA Organic certification and transparent supply chain to align with the current trends in the market.

### **3 TARGET CONSUMER & VALUE PROPOSITION**

Below are the different segments in the plant-based protein bar market and the macro breakdown of competitors:

**Premium High Protein:** No Cow leads with 20g protein, 1g sugar and is priced at \$2.25. Their primary target market is fitness athletes and consumers on a keto diet who prioritize a lower calorie to protein ratio. NuGo Slim is next with 16.5g protein, 3g sugar and their focus is on taste, and this is seen in their product formulation

**Premium Organic Segment:** ALOHA with a \$2.50 price is USDA Organic certified and has Hawaii-sourced ingredients, they emphasize organic sprouted grains and clean labels which appeal to consumers who prioritize clean ingredients in their protein bars.

**Mid-Range Value Segment:** Orgain prioritizes value conscious consumers, they retail for \$2.00 and contain only 10g protein with 6g sugar. Clif Bar also falls in this segment with 10g protein and 21g sugar. Products in this category generally have a high Calorie to protein ratio.

**Budget Segment:** Larabar protein falls in this category where it retails for \$1.85 with very few ingredients and has 11g protein with 14g sugar. This segment appeals to consumers who want the cheapest protein bar in the market but don't prioritize a high calorie to protein ratio and clean ingredients.

The variation in nutrition amongst these segments is significant. Protein content ranges from 3g (KIND bar) to 20g (No Cow) and sugar ranges from 1g-21g. The plant-based protein sources also vary, brown rice and pea protein are used by premium brands such as No Cow and Orgain whereas sprouted grains and nut butters are used by brands focusing on taste focused positioning (GoMacro, ALOHA).

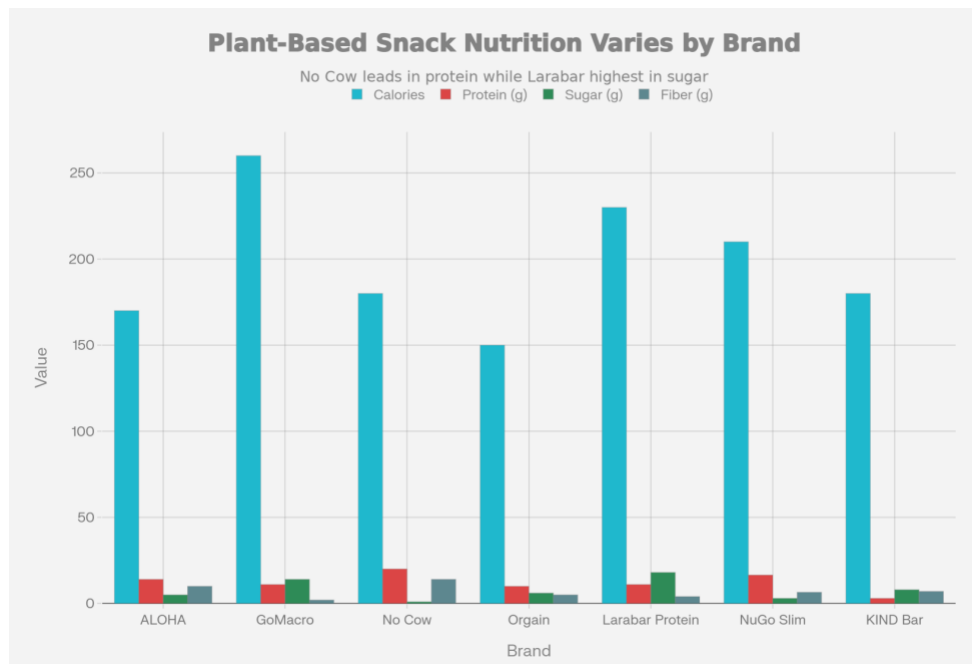


Figure 3.1 Plant-Based Protein Bar Nutrition by Brand

#### 4 K-CLUSTER ANALYSIS FOR PRICE STRATEGY

A combination of statistical techniques and business interpretation led us to choose  $K=3$  for the K-means clustering analysis, to guarantee that the pricing strategy's segmentation is solid and useful.

**4.1 Business Reasoning:** Three main competitive tiers are typically the most helpful for strategic pricing and positioning in the fast-moving consumer goods (FMCG) and competitive snack bar markets: They are A) Products in the premium tier - those with the best quality and the highest price (e.g., organic, high protein, unique ingredients). This is where the price of our plant-based snack should be rooted.

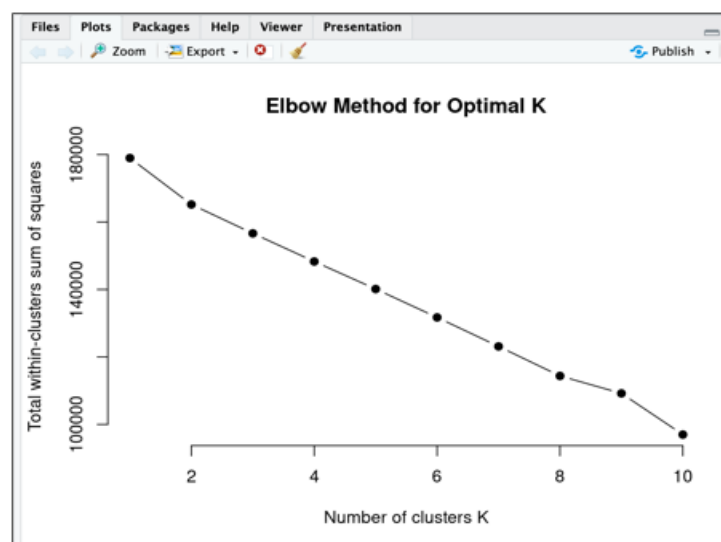


Figure 4.1. The elbow method was implemented to check the optimum value of K.

B) Mid-range tier: The core market where the majority of well-known companies compete is represented by the mid-range tier. Value and price are in balance here. This offers a discount area and a competitive boundary

C) Budget/Economy Tier: This is the lowest price tier where cost is the sole variable influencing competition. New products typically stay clear of this pricing since it doesn't reflect high-quality goods. By selecting  $k=3$ , one can directly translate the analytical results onto these obvious, practical strategic possibilities for product placement.

#### 4.2 Statistical Justification:

The Elbow Method is the main statistical method for determining the ideal  $k$  and its objective is to find the cluster efficiency point of decrease in returns. The Total Within-Cluster Sum of Squares (WSS) reduces (clusters get closer) as  $K$  rises. The "Elbow": The point on the WSS plot where the rate of decline sharply slows down and forms an "elbow" is known as the optimum  $K$ . After this, the model's fit is only slightly improved by including an additional cluster. In Practice: Choosing  $K=3$  offers a suitable balance if the Elbow Plot clearly bends about  $K=3$  (or  $K=4$ ). A higher  $K$  (such as  $K=5$  or  $K=6$ ) may be somewhat more statistically accurate, but it will not have enough size or differentiation for useful market targeting and may introduce confusingly small clusters.

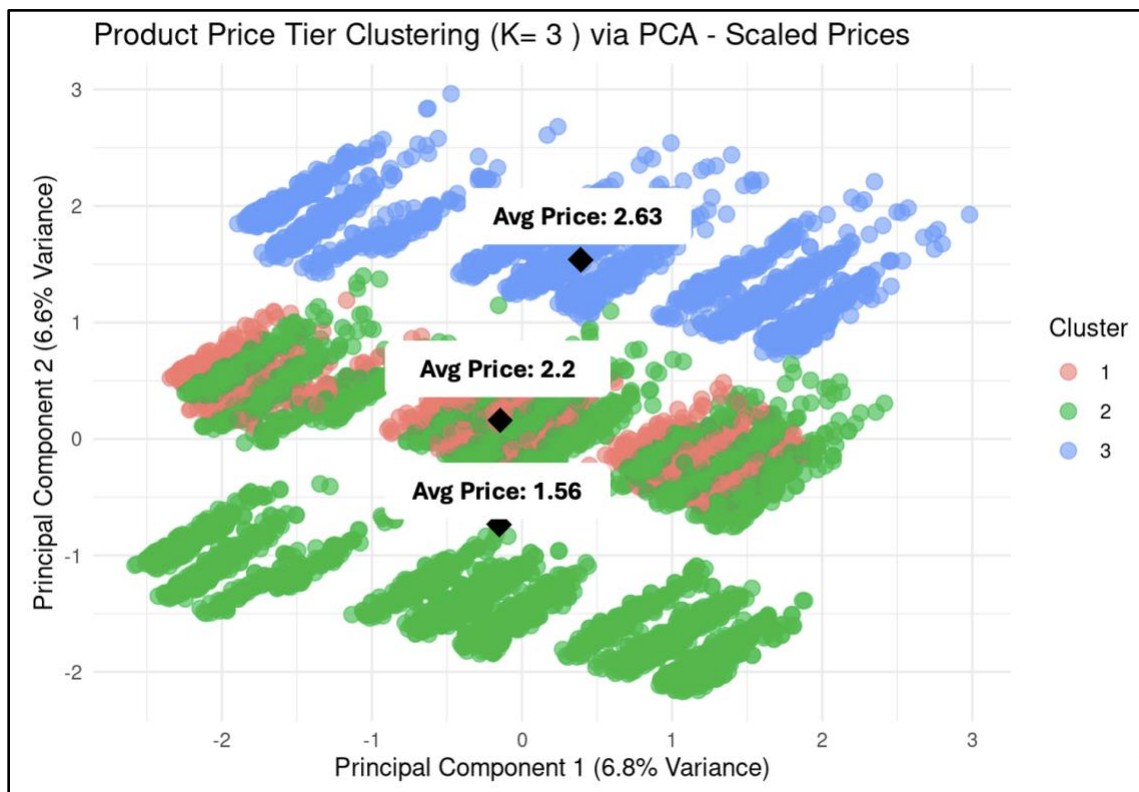


Figure 4. 2. K-cluster Analysis for  $K= 3$

#### K-Cluster Analysis Methodology

K-means clustering and Principal Component Analysis (PCA) on competition data were used to identify the pricing strategy for the plant-based snack bar, creating price tiers according to product attributes.

a)Information Inputs: Key product attributes, including Maximum Retail Price (Item\_MRP), Item\_Weight, and categorical features (Item\_Fat\_Content, Item\_Type), were included in the analysis.  
b)Preprocessing: To make sure that no single variable (like price) substantially affected the clustering distance, data was imputed (median for missing values), categorical variables were one-hot encoded, and all features were standardised (Z-scored) c) K Selection: K=3 was chosen to offer three unique, easily understood competing tiers: Budget, Mid-Range, and Premium.

The market was effectively divided into three price tiers by K-means clustering, and each group's competitive anchor was the Mean Scaled MRP.

A. Table 1. Table of Competitive Price Tiers

Cluster ID	Avg. Scaled MRP	Price Tier Interpretation	Strategic Conclusion
3	\$2.63	Competitors with the most alluring feature combinations (such as premium ingredients or reduced fat content) are represented by the Premium Tier, which has the highest average price.	Target Anchor: To indicate quality, the launch price needs to be stated here.
2	\$2.20	Mid-Range Tier: The average for the mass market; these rivals strike a balance between features and cost..	Secondary Price: The maximum allowable discount price is represented by the secondary price.
1	\$1.56	Budget Tier: The lowest price point with only cost competition.	Avoidance: This pricing significantly lowers the profit margin.

The PCA graphic displays this segmentation. The major axes (PC1 and PC2) indicate combinations of the underlying criteria (e.g., PC1 can represent "Premium Quality/Price" whereas PC2 represents "Visibility/Bulk").The visual separation of Cluster 3 from the others shows that a distinct group of items demands a clear price premium based on its feature set; the tight grouping of Clusters 1 and 2 implies less differentiation and more intense price competitiveness in the lower market segments.

#### 4.3 Conclusions and Recommendations for Pricing Strategy from K-cluster analysis

The final competitive pricing range based on the planned features of the product is provided by the K-cluster analysis.

Pricing is essentially a decision between increasing volume (Cluster 2) and maximizing margin (Cluster 3), according to the analysis. A plant-based snack bar must price itself out of the cost-driven budget tiers in order to strategically indicate quality and health value. By avoiding the fierce pricing wars of the Mid-Range and Budget categories, the suggested entry price of \$2.75 guarantees that the product is viewed as a Premium Tier contender.

Table 2. Table of Recommended Price Tiers

Decision Component	Final Value	Rationale
Target Cluster	Cluster 3 (Premium Tier)	The analysis demands a price anchor that aligns with the highest quality competitive set, maximizing the potential for profitability.
Final Price Recommendation	\$2.63 - \$2.75 per bar	The lower bound (\$2.63) matches the statistical mean of the Premium Cluster, while the upper bound (\$2.75) provides a small buffer to position the snack as superior without breaking from the established premium price structure.

## 5 ENTRY STRATEGY RISK ANALYSIS

Launching a new plant-based high-protein snack (protein bar) in the U.S. market offers big opportunities but also comes with several business risks. Majority of these risks are related to how U.S. consumers will respond, how reliably the ingredients can be supplied, and how the brand fits in the competitive food market (with organic plant based fast food or, snack category). Based on a literature review, six key risk factors have been identified. In this study, EMV and MAUT analyses are used to assess these risks and support the selection of an entry strategy for a successful product launch.

### 5.1.1 Risk Factors to EMV Based Strategy Selection

The six key risk factors are ranked by likelihood in a table to support subsequent expected monetary value (EMV) analysis and strategy selection.

Table 5.1.1: Risk register with likelihood

Risk Factors	Description	Likelihood
Consumer Acceptance / Taste-Texture	Product may not meet U.S. taste expectations	High
Ingredient / Supply Chain Uncertainty	Key ingredients may face shortage, unavailable, or cost changes	High
Regulatory & Labeling Compliance	Risk of breaking U.S. food label or claim rules	Medium
Brand Awareness / Market Positioning	New brand may struggle to compete	Medium
Sustainability / Greenwashing Risk	Unverified “eco-friendly” or “plant-based” claims can devalue reputation	Low
Pricing Pressure / Price Sensitivity	Consumers may feel price is too expensive	Low

According to the likelihood of risk factors, early commercial success of an organic high-protein snack in the U.S. market is most threatened by consumer acceptance and supply chain reliability, while regulatory, branding, sustainability, and pricing risks remain significant but are relatively less likely.

### 5.1.2 Qualitative Likelihoods to Quantitative probabilities

To enable EMV computation, likelihood ratings have been converted to midpoint probability bands for each risk factor's unfavorable outcome. This conversion provides a transparent bridge between qualitative judgment and quantitative decision modeling.

Table 5.1.2: Likelihood-to-probability mapping

Risk	Likelihood	Unfavorable p(U)	Favorable p(F) = 1 - p(U)
C1: Taste-Texture	High	0.50	0.50
C2: Supply Chain Uncertainty	High	0.50	0.50
C3: Regulatory/Labelling	Medium	0.30	0.70
C4: Brand Positioning	Medium	0.30	0.70
C5: Sustainability	Low	0.15	0.85
C6: Price Sensitivity	Low	0.15	0.85

### 5.1.3 Definition of market-entry: Top level decision alternatives

Three realistic entry strategies have been defined as decision alternatives at the top level of the decision tree. The goal is to select one of these entry strategies for U.S. plant-based (organic) protein bars.

Table 5.1.3: Table: Market-entry alternatives

No.	Alternative	Rationale (brief)
1	A1: National launch	Max scale and speed; highest exposure to scale-driven risks
2	A2: Regional pilot	Test in 3-5 key states; enable market learning and validation
3	A3: Online-first	Lower initial fixed commitments; DTC + marketplace

### 5.1.4 Risk Factor Aggregated Model structure

In this implementation, the decision tree uses a parallel risk architecture instead of modeling risks as sequential steps. Each strategy  $A_i$  is divided into six concurrent chance nodes (C1 - C6) that collectively feed into a single overall outcome node  $O_i$ . This structure reflects that taste - texture, supply chain uncertainty, regulatory compliance, branding – market positioning, sustainability, and price sensitivity can all emerge simultaneously during the product launch.

### 5.1.5 Mapping risk nodes to strategy-specific probabilities

Based on the latest research in fast-moving consumer goods (FMCG) strategy (including 'Real Options' valuation and 'Lean Launch' methodologies), the baseline probabilities explained in table (Likelihood-to-probability mapping) were further adjusted to reflect realistic operational outcome.

Table 5.1.4: Strategy-specific unfavorable probabilities

Strategy	Risk Performance Hypothesis	p(U) – Vector [C1, C2..., C6]
A1: National launch	Worse supply chain and broader market taste risk performance.	[0.52, 0.60, 0.30, 0.30, 0.15, 0.15]
A2: Regional pilot	Better customer acceptance, improved supply chain (due to local learning + staged scale up)	[0.38, 0.45, 0.30, 0.25, 0.15, 0.15]
A3: Online-first	Moderate supply chain issues	[0.50, 0.47, 0.30, 0.28, 0.15, 0.15]



### 5.1.6 EMV computation and payoff model

Total risk scenarios have been matched with four market performance outcomes, along with sample Net Present Values NPVs (to be replaced with the project's financial model if available).

Table 5.1.5: Terminal outcomes and payoffs

No.	Terminal node	Interpretation	NPV (M\$)
1	N1	Strong launch	60
2	N2	Moderate launch	30
3	N3	Underperformance	5
4	N4	Exit / Relaunch	-10

The EMV for each strategy is computed at the aggregate node:

$$EMV(A_i) = p_1 \cdot 60 + p_2 \cdot 30 + p_3 \cdot 5 + p_4 \cdot (-10)$$

where  $p_1$  to  $p_4$  are derived from the risk Aggregation rule applied to C1 – C6 (refer Appendix).

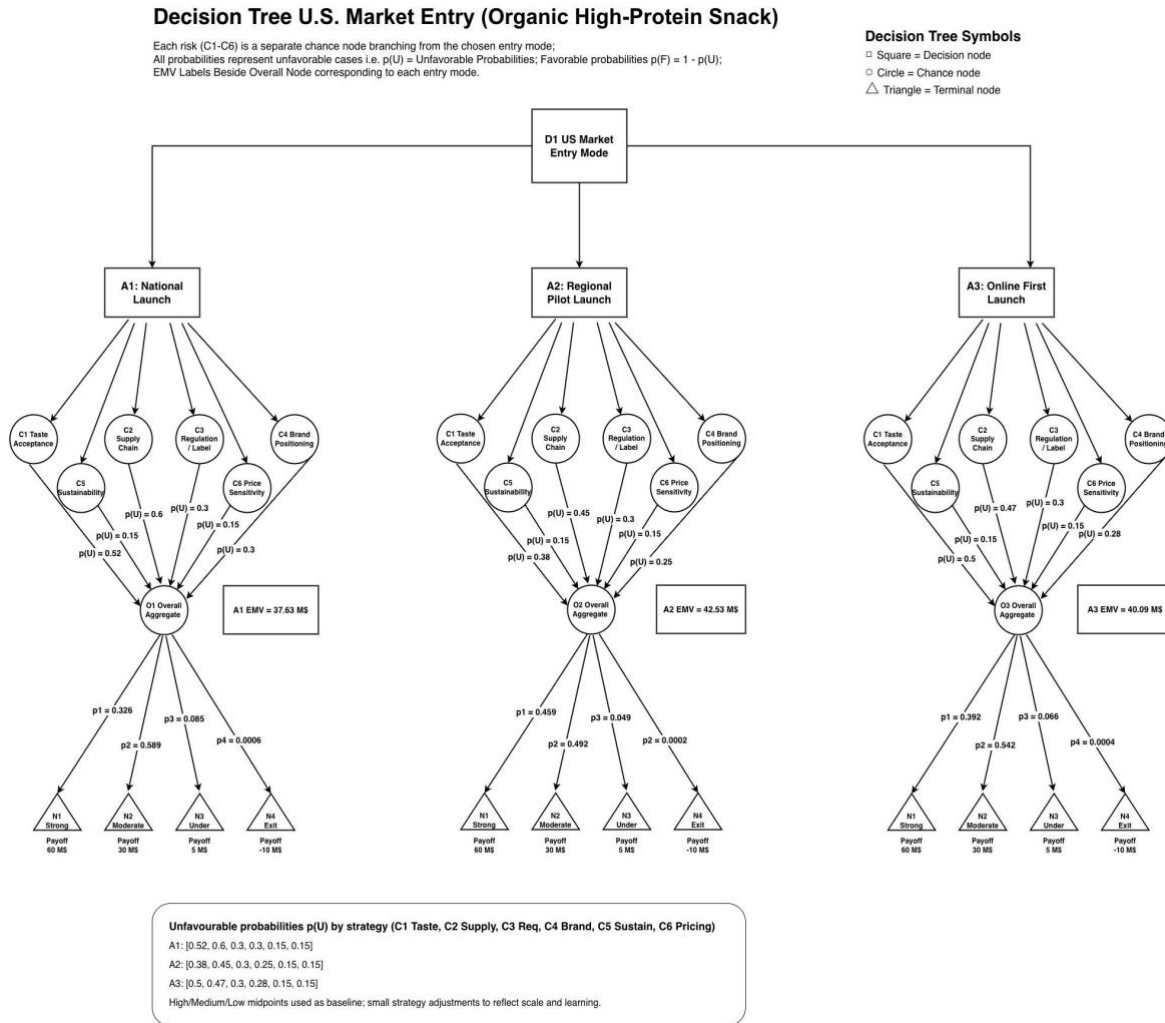


Figure 5.1.1: U.S. Market Entry with an Organic High Protein Snack (Bar)

Using probability of k unfavorable risks for C1 - C6, values of p1 – p4 are computed (refer Appendix).

Table 5.1.6: Value of p1 – p4 (highlighted in the decision tree)

Strategy	p1 = P(k ≤ 1)	p2 = P(k = 2 or 3)	p3 = P(k = 4 or 5)	p4 = P(k = 6)
A1 National	0.325822	0.588831	0.084715	0.000632
A2 Regional pilot	0.458650	0.492400	0.048661	0.000289
A3 Online-first	0.391507	0.542429	0.065619	0.000444

A1 National launch:

$$\begin{aligned} \text{EMV}(A1) &= 0.325822 * (60) + 0.588831 * (30) + 0.084715 * (5) + 0.000632 * (-10) \\ &= 19.549320 + 17.664930 + 0.423575 - 0.006320 = 37.631508 \text{ M\$} \end{aligned}$$

A2 Regional pilot:

$$\begin{aligned} \text{EMV}(A2) &= 0.458650 * (60) + 0.492400 * (30) + 0.048661 * (5) + 0.000289 * (-10) \\ &= 27.518999 + 14.772000 + 0.243305 - 0.002889 = 42.531438 \text{ M\$} \end{aligned}$$

A3 Online-first:

$$\begin{aligned} \text{EMV}(A3) &= 0.391507 * (60) + 0.542429 * (30) + 0.065619 * (5) + 0.000444 * (-10) \\ &= 23.490417 + 16.272870 + 0.328095 - 0.004444 = 40.086979 \text{ M\$} \end{aligned}$$

**5.1.7 EMV result table and ranking**

Below table shows EMVs for A1, A2, and A3 in order of ranking.

Table 5.1.7: EMVs for strategies A1 – A3 in ranking order

Strategy	EMV (M\$)	Rank
A2 Regional pilot	42.53	1
A3 Online-first	40.09	2
A1 National launch	37.63	3

**5.1.8 Recommended Strategy: A2 (Regional Pilot)**

The EMV ranking supports A2 (Regional pilot) as the most appropriate entry strategy. This result is structurally consistent with the data in the risk register with likelihood table. In essence, the regional pilot reduces the two highest probability uncertainties i.e. taste/texture acceptance & supply chain uncertainty.

**5.2.1 Risk Averse MAUT (Additive Utility) for U.S. Market Entry**

The decision to enter the US market with an organic protein snack is well suited for a risk-based multi-criteria decision analysis (MCDA), given the small number of distinct alternatives and a clearly defined risk register (table Risk register with likelihood). In this section, we applied a multi-attribute utility theory (MAUT) or weighted scoring model that employs the same six risk factors from the risk register as evaluation criteria. This ensures consistency across methods: the risk register identifies fundamental uncertainties, the decision tree converts these into expected monetary values (EMVs), and the risk-based MAUT compares how effectively each entry strategy mitigates these risks. The aggregate form provides a transparent first-order representation of risk-focused preferences and supports subsequent sensitivity analysis of weight uncertainty.

Alternatives:

The three market entry strategies i.e. A1 National launch, A2 Regional pilot, and A3 Online-first are selected as alternatives for evaluation.

Model:

A simple additive MAUT/weighted scoring model is considered for decision analysis.

$$U(A_i) = \sum_{j=1}^m w_j \cdot u_{ij}, \sum w_j = 1, w_j \geq 0$$

Where:

$w_j$  = importance weight of criterion  $j$  under a risk-averse frame  
 $u_{ij}$  = performance score of strategy  $i$  on criterion  $j$ , using a 1 - 5 scale (5 = best)

Aligned with the Risk factors described in section 1, the below criteria and weights are defined.

Table 5.2.1: Define criteria and weights for MAUT

Criterion	Justification	Weight
Consumer fit/learning	Mitigates taste - texture acceptance risk (High)	0.25
Supply chain robustness	Reduces ingredient shortage/cost risks (High)	0.25
Regulatory control	Reduces labeling/claim compliance risk (Medium)	0.15
Brand positioning strength	Addresses new brand positioning risk (Medium)	0.15
Sustainability	Mitigates greenwashing/claim credibility risk (Low)	0.10
Price competitiveness	Mitigates price sensitivity risk (Low)	0.10
<b>Total</b>		<b>1.00</b>

Table 5.2.2: Strategy performance scores consistent with research articles and market drivers (1 - 5)

Strategy	Consumer learning	Supply robustness	Regulatory control	Brand position	Sustainability	Price completion
A1 National launch	2	2	3	4	3	3
A2 Regional pilot	5	4	4	3	3	3
A3 Online-first	4	3	3	2	3	3

Utility Calculation:

$$U(A1) = 0.25(2) + 0.25(2) + 0.15(3) + 0.15(4) + 0.10(3) + 0.10(3) = 2.65$$

$$U(A2) = 0.25(5) + 0.25(4) + 0.15(4) + 0.15(3) + 0.10(3) + 0.10(3) = 3.90$$

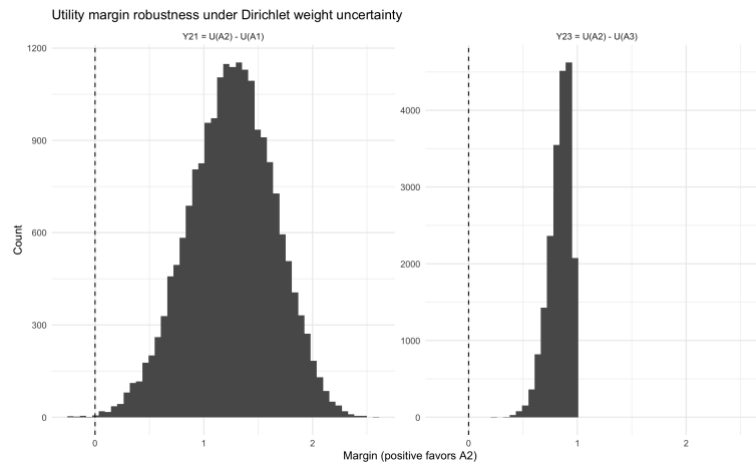
$$U(A3) = 0.25(4) + 0.25(3) + 0.15(3) + 0.15(2) + 0.10(3) + 0.10(3) = 3.10$$

Table 5.2.3: Final ranking based on Utility

Strategy	Risk-averse utility (U)	Rank
A2 Regional pilot	3.90	1
A3 Online-first	3.10	2
A1 National launch	2.65	3

## 5.2.2 Global Robustness Analysis of Risk-Averse MAUT

I apply a Dirichlet-based robustness analysis (Saint-Hilary et. al. 2017) to test whether the preferred entry option remains stable when there is variation in the weights of risk-averse decision-makers. This process ensures that the weights are non-negative, and their sum equals one.



**Figure 5.2.1: Utility Margin Robustness Under Dirichlet Weight Uncertainty**

The above figure demonstrates the robustness of A2's advantage under Dirichlet weight uncertainty using two utility margins. Left panel ( $Y_{21} = U(A_2) - U(A_1)$ ): the histogram is heavily concentrated well to the right of zero, with only a very small mass near the discrete zero line. This indicates that A2 almost always outperforms A1 under reasonable risk-averse weight variations. The spread shows that the strength of A2's advantage over A1 may vary, but the sign is overwhelmingly positive. Right panel ( $Y_{23} = U(A_2) - U(A_3)$ ): the distribution is entirely positive and tightly clustered, meaning that A2 consistently outperforms A3 in nearly all sampled weight profiles. The narrower spread reflects that A2 and A3 show a meaningful difference in a smaller risk dimension, making this margin more stable.

## 5.2.3 Conclusion from Risk Analysis

The EMV decision tree indicates that, considering high Consumer Packaged Goods (CPG) failure rates and fixed launch costs, A2 (approximately \$42.5 million) provides the highest expected value compared to the A1 national launch (approximately \$37.7 million) and the A3 online-first approach (approximately \$40.1 million). Although A1 offers higher theoretical potential, it has high risk of taste and supply failure. A3 faces challenges linked to online-only models like customer acquisition costs and logistical constraints.

According to risk-focused MAUT method: When higher weights are assigned to consumer acceptance and supply reliability, A2 achieves the highest risk averse utility ( $U=3.90$ ), followed by A3 (3.10) and A1 (2.65); The Dirichlet-based robustness analysis demonstrated that this ranking is practically stable: A2 ranks first in 99.94% of simulations;  $P(U(A_2) > U(A_1)) = 0.99$  and  $P(U(A_2) > U(A_3)) = 1.00$ .

These results align with the study in innovation and marketing literature where phased market entry through pilot programs (A2) is reported having small “application fee” that limits downside risks while preserving the right to scale up (e.g., Luehrman, 1998; McGrath, 1997; Roberts & Lattin, 2000). While A1 appears to be a large and irreversible bet, A3 behaves like a low-risk, low-potential niche strategy limited by customer acquisition cost (CAC) and logistical constraints (Avery et al., 2012; Teixeira & Brown, 2020). Therefore, our recommended market entry strategy is a A2 regional pilot launch that includes a gradual transition to nationwide rollout.

## 6 DISTRIBUTION & GEO “HOTSPOT” ANALYSIS

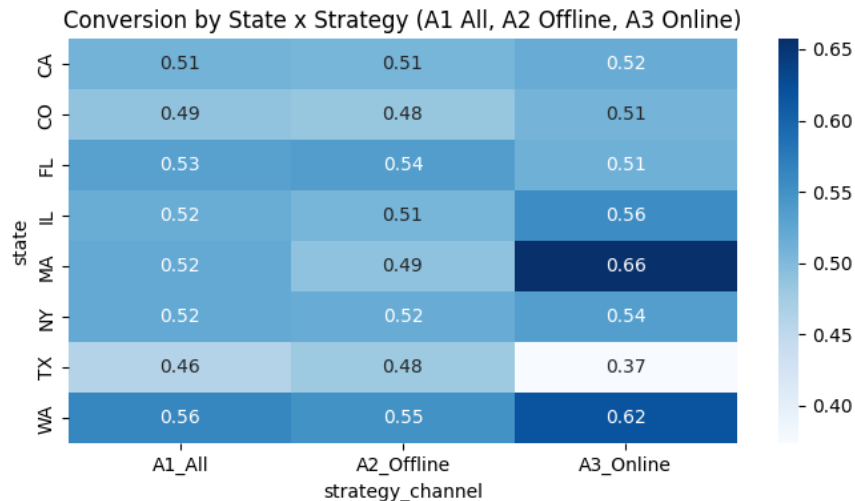
### 6.1 Data and Hotspot Analysis Approach

To support the market entry decision, a synthetic transaction level dataset was constructed for plant based high protein snack bars in the US. The data represents a potential customer interaction and includes state, retail channel (grocery, mass, club, online), diet segment, income, health orientation score, price per bar, purchase indicator (0/1) and a simple ROI proxy per observation. The dataset was then aggregated to the level of state  $\times$  strategy, where A1 corresponds to all channels combined, A2 represents offline channels only (grocery, mass, club), and A3 represents the online channel. For each state-strategy cell, three key metrics were calculated: average selling price, conversion rate (share of customers who purchased) and average marketing ROI. These metrics were visualized as heatmaps to identify performance hotspots and cold spots across the three strategies.

### 6.2 Conversion heatmap

This chart shows what fraction of customers buy the bar in each state for:

1. A1\_All shows the national view
2. A2\_Offline includes grocery, mass and club
3. A3\_Online
4. Darker blue = higher conversion.
5. MA and WA have very dark blue for A3\_Online, meaning online campaigns convert especially well there, while TX has light blue for A3\_Online, meaning online under-performs in Texas.
6. We can see that A2\_Offline is more even and its conversion is mid to high in most states, without extreme lows which supports a regional offline pilot.

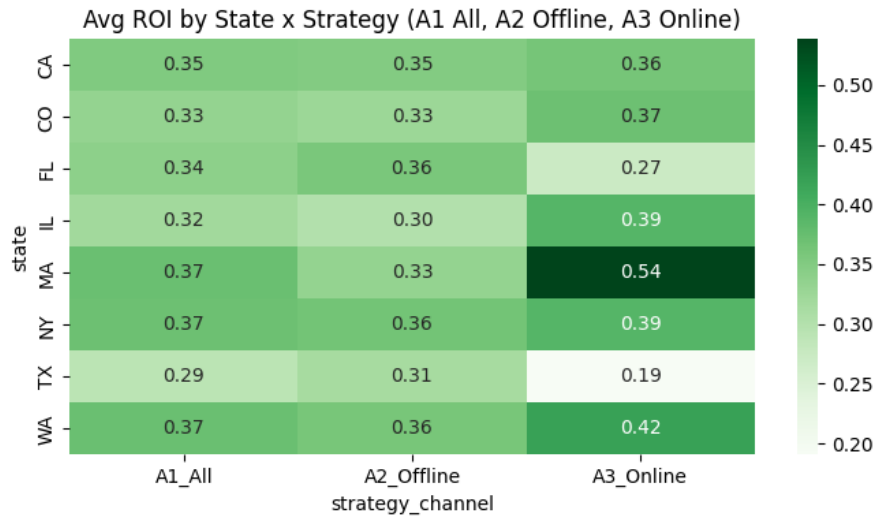


**Figure 6.1.** Conversion rate by state and strategy

### 6.3 ROI heatmap

1. This chart shows the average ROI of your marketing spend in each state-strategy combination.
2. Darker green = higher ROI
3. Again, MA and WA for A3\_Online are very dark, so online is a strong lever there but TX Online is pale, signaling poor ROI.

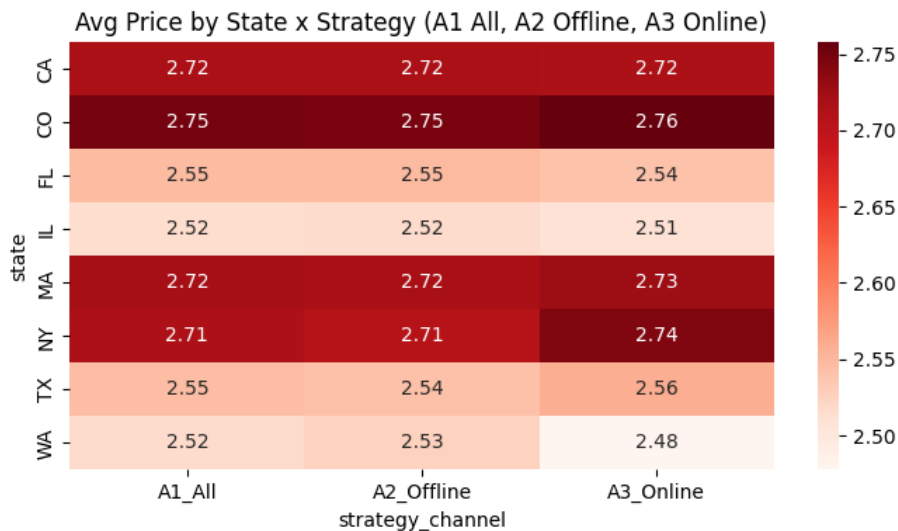
4. We can infer that A2\_Offline gives a steady mid green across many states, so it is a safer and a more balanced option than going online only everywhere.



**Figure 6.2.** Average ROI by state and strategy

#### 6.4 Price heatmap

1. This chart shows average selling price per bar in each state-strategy.
2. Colors are very similar across strategies: prices are around 2.5-2.75 dollars everywhere, with only a few cent differences.
3. That means your strategic choice is not about charging very different prices, but about where and how you sell to get better conversion and ROI.



**Figure 6.3.** Average price by state and strategy

## 6.5 Market Challenges

### **Consumer Taste and Texture Acceptance**

Challenge: Many consumers still perceive plant-based protein snacks as having chalky texture or earthy flavor due to pea or rice protein bases.

### **Competitive Saturation and Formulation Similarity**

Challenge: The U.S. protein snack market is highly competitive with overlapping product claims like “plant-based,” “clean label,” “high protein,” and “no added sugar.”

### **Price Sensitivity**

Challenge: Consumers are price-sensitive in the snack category and the Plant-based protein products often cost 20–40% more than traditional snacks due to their ingredients

### **Ingredient Supply Chain and Cost Volatility**

Challenge: Too much depended on specific ingredients such as pea protein and brown rice protein creates bottlenecks supply and cost fluctuations.

## 6.6 Conclusions and Recommendations from Hotspot Analysis

Hotspot analysis confirms that the A2 Regional pilot is the best entry strategy, aligning with EMV results (A2: \$42.53M vs. A3: \$40.09M and A1: \$37.63M). A national launch (A1) averages strong and weak regions, diluting performance, while an online-first approach (A3) excels in states like MA and WA but risks poor results in weaker regions such as TX. A regional pilot focuses on high-conversion, high-ROI states and avoids heavy investment in cold spots, offering the most balanced and profitable approach.

## 7 CONCLUSION

Market analysis indicates a clear positioning opportunity for an organic, plant-based, high-protein bar that balances nutrition with strong flavor and texture, positioned between performance-focused bars and premium organic products. This plant-based protein bar should ideally have between 15-20g of protein with a maximum of 200 calories.

K cluster pricing results support a premium list price corridor of 2.63 to 2.75 per bar. The 2.40 to 2.50 level should be treated as a trial reference point achieved through controlled promotions or multipack effective pricing during the pilot, without changing the premium anchor.

Market entry risk analysis supports a regional pilot as the preferred entry strategy because it maximizes value while limiting downside risk and enabling learning before scale.

Hotspot analysis reinforces this approach by showing performance varies by state and channel, so a focused pilot should prioritize the strongest geographies and use targeted online activation only where digital conversion and ROI are highest, rather than relying on a uniform national rollout.

Next steps include executing an offline regional pilot program with targeted online support, validating taste and texture as primary risks, ensuring supply and compliance readiness, and establishing clear scaling thresholds based on repeat purchase, sales velocity, profit margin, and service levels.

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