

# Main Insights and Recommendations

## Short Summary of Key Findings

### Price Variation by Product

#### 1. Gasoline

- **Generally commands the highest average selling price** among the three primary products (Gasoline, Diesel, Ethanol).
- The distribution often centers around **4 BRL**, but can extend beyond **8 BRL** in some outlier cases—possibly driven by regional taxes, specialized branding, or unique local market structures.

#### 2. Diesel

- Often priced just below Gasoline on average, with a **median in the low- to mid-3 BRL** range.
- Despite a slightly lower baseline, Diesel can also see **extreme outliers** above 8 BRL, suggesting localized conditions (e.g., remote industrial hubs, supply chain bottlenecks).

#### 3. Ethanol

- Tends to be cheaper overall, typically **2–3 BRL** per liter in median ranges.
- Even with lower average prices, **outliers** can appear, possibly reflecting periods of ethanol shortage or demand spikes in certain markets.

### Implication:

Each product follows a different pricing dynamic and volatility profile, underscoring the necessity of **product-specific pricing and procurement strategies**. Monitoring local supply, tax structures, and brand positioning is particularly critical for Gasoline and Diesel, where price variability is greater.

### Purchase Price Dynamics

- **Most purchase prices** lie between **1 BRL and 2.5 BRL**, with fewer extreme outliers compared to selling prices.
- The **right-skewed tail** of purchase prices occasionally exceeds 4 BRL, aligning with situations where supply constraints, transportation costs, or niche suppliers drive up costs.

### Potential Margin Impacts:

- **Consistently higher avg\_preco\_venda (sale price) vs. avg\_preco\_compra (purchase price)** supports a stable markup for many transactions.
- **Outlier scenarios** with very high or low purchase prices can compress or inflate margins in specific localities or time periods, indicating the need for **close supplier management** and possible renegotiation of contracts.

## Margin Analysis & Outliers

1. **Typical Margins:**
  - The **boxplots** often show **10–30%** as a common margin range (corresponding to ~0.3–0.7 BRL over purchase).
2. **Negative or Extremely High Margins:**
  - Some transactions yield **negative spreads**, hinting at promotions, local price wars, or potential data entry errors.
  - **High-margin outliers** (above 50% or more) could be tied to remote or captive markets, premium branding, or short-term supply disruptions.
3. **Product-Specific Margin Profiles:**
  - Gasoline and Diesel typically exhibit **greater margin volatility**, whereas Ethanol shows somewhat more stable but still significant swings.

### Implication:

**Investigating extreme margin values** (both positive and negative) can uncover hidden costs, inefficiencies, or strategic opportunities. Scrutinizing these transactions can reveal areas for **improved operational controls** or **competitive advantage** in targeted markets.

## Geographic & Market Concentration Insights

### State-Level Concentration (HHI Scores)

- **High HHI States (e.g., MA, MT, AM):**
  - Highly concentrated; fewer dominant players; higher potential for **price premiums**
- **Lower HHI States (e.g., DF, RS):**
  - More competitive; typically **lower average prices**, narrower margins due to **multiple strong competitors**.

### Number of Establishments (n\_estabelecimentos)

- **Highly Skewed:**
  - Most observations under 50 establishments, but some municipalities exceed 1,000.
  - Large urban centers (or aggregated reporting) can distort averages

## Geographic Price Variation & Correlation

- **Remote vs. Urban:**
  - **Remote or rural municipalities** often face higher transportation costs, leading to **elevated selling prices** and (often) higher margins.

- **Urban centers** show somewhat lower median prices (due to greater competition) but can have **very high outliers** (brand premiums, overhead costs).
- **Correlation Declines Over Distance:**
  - Nearby cities often track each other’s prices, whereas **far-flung municipalities** can diverge significantly due to different supply routes, local taxes, and demand factors.

#### Implication:

Region-specific factors—**logistics, market concentration, establishment density, and local taxation**—all play key roles in shaping price and margin. A “**one-size-fits-all**” approach to pricing will miss many nuances; **robust regional segmentation** is essential.

### Brand Coverage & Overlapping Networks

1. **Universal State Presence:**
  - Major brands (e.g., White Label, Blue Label, Green Label, Purple Label) show coverage in all states, though not necessarily with the same market share.
2. **Competitive Pressure:**
  - Because multiple brands co-exist in most states, local competition is common. This underscores the value of **localized brand strategies**—especially where a brand’s presence or reputation varies.
3. **Potential for Supply Chain Optimization:**
  - If a corporate entity owns multiple brands, there may be **transport synergies** or shared distribution that can cut costs.
  - Independently owned brands may pursue **third-party logistics contracts** or distribution partnerships to improve coverage and inventory efficiency.

#### Implication:

**Network-level** decisions—such as re-routing shipments, consolidating brands, or optimizing distribution points—can drive **cost savings** and help maintain price competitiveness. Overlapping footprints require careful management to **avoid cannibalization** (if brands are under one umbrella) or to **target competitor vulnerabilities** (if operating independently).

### Time-Series Trends & Volatility

1. **Long-Term Upward Trend**
  - From **2004 to mid-2010s**, prices rose gradually, reflecting inflationary pressures and global oil market fluctuations.
2. **Post-2016 Acceleration**
  - Steeper climbs suggest **currency issues, macroeconomic shifts, and possibly new policy changes**—impacting all three major

fuels.

### 3. Rising Volatility Since ~2018

- **12-month rolling standard deviation** doubled or tripled, approaching ~1.0 BRL, indicating larger and more frequent price swings.
- High volatility amplifies **profit risk** and requires **agile pricing strategies** (e.g., daily or weekly adjustments, automated dynamic pricing).

**Cluster Analysis** of historical prices identifies **distinct phases** (e.g., lower-price era vs. high-volatility era). Each phase or cluster calls for **different inventory, hedging, and pricing tactics**.

### Actionable Recommendations

#### 1. Pricing & Margin Management

- **Dynamic Pricing:** In high-volatility contexts (post-2018), automate or frequently review prices.
- **Margin Outlier Investigation:** Identify negative or excessively high margins; rectify data errors or exploit market opportunities.

#### 2. Regional Segmentation

- **Focus on High-HHI States:** With fewer competitors, consider cautious price increments but watch for regulatory attention.
- **Low-Margin, Competitive Zones:** Improve operational efficiency, possibly partner for supply discounts or explore brand differentiation.

#### 3. Supply Chain & Logistics

- **Optimize Distribution:** Re-route shipments to reduce transportation costs in remote markets.
- **Inventory Buffering:** In volatile price environments, holding strategic reserves could mitigate cost spikes, but watch carrying costs.

#### 4. Brand Strategy & Network Coordination

- **Position Brands by Region:** Use premium branding where local demand supports higher prices; deploy cost-competitive labels in price-sensitive or highly competitive areas.
- **Prevent Cannibalization:** If multiple brands fall under one corporate owner, define **clear market segments** for each label.

#### 5. Time-Based Forecasting & Scenario Planning

- Segment historical data into **price/volatility regimes** (clusters) to refine forecasts.
- Develop **scenario-based business plans:** e.g., best case (steady market), moderate (gradual price growth), worst case (high volatility)

& supply disruptions).

#### 6. Regulatory & Policy Monitoring

- Track ongoing **tax or subsidy changes**, environmental regulations, and import/export policy shifts that can abruptly impact price or supply.
- Engage in **lobbying** or stakeholder dialogues where localized tax structures severely skew costs.

### Modeling for forecasting

- **Model type:** RandomForestRegressor
- **Test set RMSE:** 1.0275
- **Test set R<sup>2</sup>:** -0.2517

#### Financial Impact

- **Total Actual Revenue:** \$4,441,103.35
- **Revenue Prediction Error:** \$725,492.66 (*16.34%*)
- **Total Actual Profit:** \$493,765.82
- **Profit Prediction Error:** \$80,475.89 (*16.30%*)

### Model Performance & Forecast Quality

#### Analysis of the forecast

- **Conservative Forecasts:** The model tends to **underestimate** actual values. This can serve as a **built-in safety margin**, preventing overestimation of demand or revenue—helpful if budgets or inventory are tight.
- **Risk of Missed Upside:** Underestimating ~75% of the time suggests we might leave money on the table in strong markets.

### Financial Impact Analysis

#### Revenue & Profit

- **Conservative Baseline:**
  - Predicted revenue (\$3.71M) vs. actual (\$4.44M).
  - Predicted profit (\$413K) vs. actual (\$494K).
- Even though these forecasts trail real-world values by roughly **16%**, this shortfall can be **corrected** with ongoing model improvements (e.g., adding relevant external data or revisiting feature selection).

## Practical Takeaways

- **Managing Downside Risk:** With an underestimate, the business is less likely to overcommit resources.
- **Opportunity Costs:** In a bullish scenario (high market demand), undershooting forecasts may result in **stockouts** or **inadequate staffing**, limiting revenue capture.

## Business Implications

1. **Underestimation Bias**
  - **Safe Inventory Approach:** Because the model undershoots, you're less likely to end up with large surpluses.
  - **Growth Potential:** If demand spikes, you risk missing revenue because capacity or investment might be set too low.
2. **Profitability Considerations**
  - **Conservative Projections** can help secure financing or manage costs under stable conditions.
  - **High-Volatility Markets:** Rapid price changes or demand surges may amplify the risk of leaving profits on the table.
3. **Strategic Planning**
  - **Risk-Averse Budgeting:** Forecasts at 16% below actual can serve as a “worst-case” or baseline scenario.
  - **Flexibility Needed:** Supplement the model's predictions with real-time market signals or rapid re-forecasting methods to **capture upswings** more effectively.

## Recommendations & Next Steps

1. **Refine the Model**
  - **Feature Expansion:** Incorporate macroeconomic variables, competitor moves, seasonal indices, and marketing spend to capture demand fluctuations more accurately.
  - **Hyperparameter Tuning & Segmentation:** Explore separate models for different product lines, regions, or customer segments where the relationship between inputs and revenue/profit may differ.
2. **Continuous Monitoring**
  - **Rolling Retraining:** Regularly update the model with fresh data, ensuring it learns from recent market shifts.
  - **Real-Time Dashboards:** Track actual vs. predicted performance and alert decision-makers when deviations exceed a critical threshold

(e.g.,  $\pm 10\%$ ).

3. **Scenario & Contingency Planning**

- **Multiple Forecast Scenarios:** Use the model’s conservative outputs for baseline planning, but also create optimistic/“best-case” scenarios to inform opportunistic decisions.
- **Buffer Stocks & Optionality:** In markets prone to rapid demand surges, maintain a small **inventory cushion**—countering the model’s underestimation bias.

4. **Operational Safeguards**

- **Threshold Tweaks:** If you use profit thresholds for go/no-go decisions, consider adjusting them upward to **compensate** for the model’s conservatism.
- **Sensitivity Analyses:** Periodically test how changes in certain inputs (price, demand, cost) might yield different outputs—and plan your supply chain or marketing campaigns accordingly.

5. **Try Additional Algorithms:** Use **AutoML** (e.g., H2O.ai, auto-sklearn) to quickly evaluate new models (LightGBM, CatBoost) and tune hyperparameters more effectively.

6. **Advanced Approaches:** Test **Bayesian models** or **Prophet** for time-series behavior.

7. **Feature Selection:** Apply **RFE** or SHAP to prioritize the most impactful features.

8. **Expand Hyperparameter Search:** Move beyond simple grids; consider Bayesian optimization for better results.

9. **MLflow:** Track experiments, metrics, and parameters for reproducibility.

10. **Dockerize & Deploy:** Package the model in a Docker container for consistent, scalable production use.

11. **CI/CD:** Automate testing, building, and deployment of the Docker image.