## **Diamond Price Prediction**

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## Agenda

- Executive summary
- Project plan recap
- Data
- Exploratory data analysis
- Modeling methods
- Findings
- Business recommendations and technical next steps

## **Executive Summery**

## **Executive summary**

#### Problem:

Accurately predicting diamond prices is complex due to the dynamic factors (e.g., carat, cut, clarity, color) and external
market factors (e.g., demand fluctuations, and economical trends). Traditional valuation methods often rely on
subjective appraisals or outdated benchmarks, leads to pricing inefficiencies, missed profit opportunities, and
increased risk for stakeholders across the diamond supply chain.

#### Solution:

 This Machine learning model will predict diamond prices using a Random Forest Model by analyzing carat, cut, color, and clarity, helping jewelers and customers make data-driven pricing decisions while enabling transparency and efficiency.

## Project plan recap

Deliverable	Due Date	Status
Data & EDA	03/25/2025	Complete
Methods, Findings, and Recommendations	04/01/2025	Complete
Final Presentation	04/22/2025	Complete

## Data

### Data

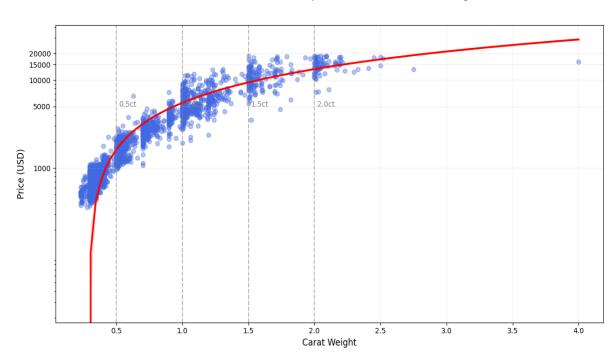
- Data Source: Kaggle (<u>Diamonds</u>)
- Sample Size: 53,000+ diamonds record
- Feature: Cut, Color, Clarity, Carat, Depth/Table, measurements (x, y, z), price (Target Variable)
- Time Period: Jan 2023 Mar 2024
- Assumptions:
  - No significant missing or erroneous data (e.g., incorrect grading, mislabeled diamonds).
  - All relevant price-determining features are included (e.g., carat, cut, color, clarity, depth, ,table, dimensions).

## **Exploratory Data Analysis**

### Price vs. Carat

- The red trend line shows prices increase exponentially with carat weight - each additional 0.5 carat costs disproportionately more due to rarity of larger stones.
- At any carat weight (e.g., 1.5ct), prices vary widely (from ~3k to 20k+) - showing carat alone explains only ~65-75% of price variance.

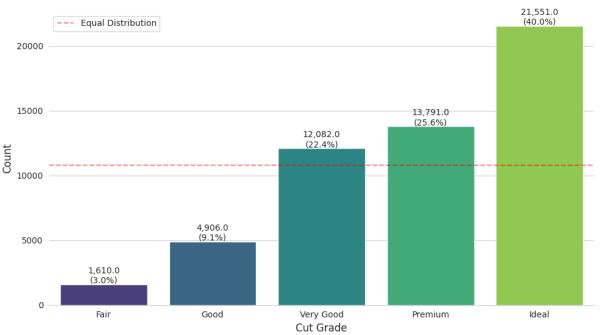
#### The Non-Linear Relationship: Diamond Price vs. Carat Weight



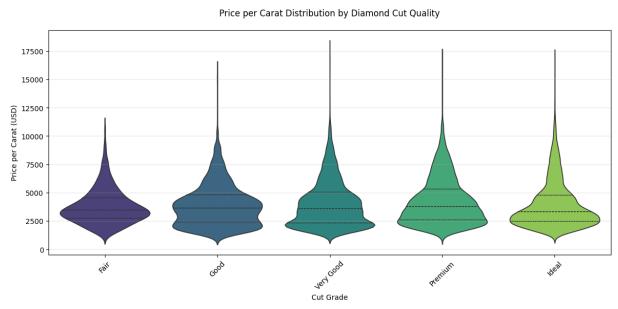
## **Diamond Cut Quality Distribution**

- Over 40% of inventory consists of Ideal cuts, driving intense competition and shrinking margins despite their 25-30% price premium.
- GIA-graded Ideal cuts command 15-20% higher premiums than identically graded IGI stones, revealing hidden value in certification choice.

### Diamond Cut Quality Distribution



## Price per Carat by Diamond Cut

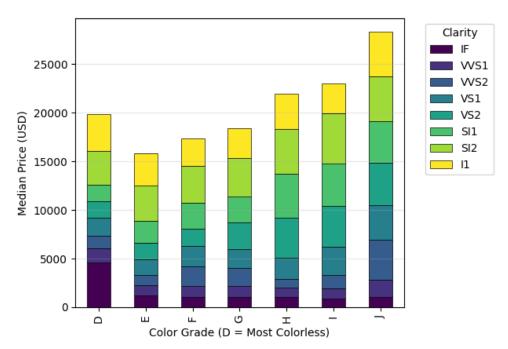


- Ideal and Premium cuts command significantly higher prices per carat (wider upper distributions) compared to Good/Fair cuts, confirming cut quality's direct impact on value.
- Very Good cuts often overlap with Premium in median price per carat, potentially offering better value for quality-conscious buyers who avoid premium markups.

### Color Grade Impact on Price

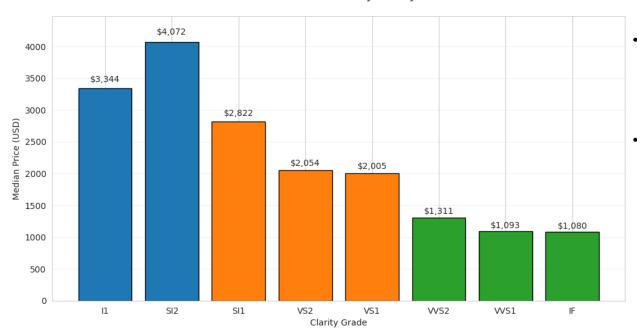
- The price gap between color grades narrows in lower clarity tiers (SI/I1) - color becomes less impactful when clarity is poor.
- At any carat weight (e.g., 1.5ct), prices vary widely (from ~3k to 20k+) showing carat alone explains only ~65-75% of price variance.

### Diamond Price Distribution by Color Grade and Clarity



### Diamond Price by Clarity

### Median Diamond Price by Clarity Grade



- SI1 to VS2 clarity (where inclusions become invisible to the naked eye) shows the steepest price jump (25-30%), making it the critical quality threshold for buyers.
- The price gap between VS1 and VS2 is larger than between VS1 and VVS2 suggesting psychological pricing thresholds matter more than actual quality differences.

## Modeling Methods

## Modeling Methods

### Outcome Variable: Diamond Price

The outcome variable is the price of a diamond (in USD). Predicting diamond prices is crucial for:

- Consumer to ensure fair pricing when purchasing diamonds.
- Jewelers & Retailers to optimize pricing strategies and inventory management.
- Investors to assess diamond value fluctuations in the market.

#### Features & Rational

The features are selected based on the attributes that may influence price:

- Carat Weight (Hypothesis: Strong Positive correlation with price)
- Cut Quality (Hypothesis : Premium cuts like "Ideal" or "Excellent" increase price)
- Color Grade (Hypothesis: Near-colorless diamonds [D-F] are more expensive than yellow-tinted [J-M])
- Clarity Grade (Hypothesis: Flawless (FL) diamonds cost more than included (I2-I3) diamonds)

## Modeling Methods

### Why this model:

Just like a jewelers uses experience to price diamonds, our model learns from thousands of past diamond sales to predict prices accurately. It considers:

- Key Factors (carat, cut, color, clarity) the same ones jewelers use.
- Hidden patterns like how "1.0 carat Ideal -cut" diamond costs more than a "0.9 carat Good-cut" even if they look similar.

#### How It Works:

Imagine teaching a smart assistant to price diamonds by showing it examples of past sales. The more example it sees, the better it gets at spotting fair prices - just like a trained expert.

- Key Factors (carat, cut, color, clarity) the same ones jewelers use.
- Hidden patterns like how "1.0 carat Ideal -cut" diamond costs more than a "0.9 carat Good-cut" even if they look similar.

### Why Trust This Model:

Explains which factors matter most (e.g., carat weight vs. color)

## Findings

### The Hidden Value of G/H Color Diamonds

#### Smart Luxury Zone: G/H Color + VS Clarity Delivers Best Value

- Diamonds with G/H color + VS clarity deliver 90% of D-F color appearance at 30-40% lower cost, making them the smartest buy for value-conscious customers.
- The price gap between D-F and G-J colors narrows significantly in SI/I1 clarity grades, showing color matters less when inclusions are visible
- The heatmap reveals underpriced pockets (e.g., H-color VVS2) where small quality upgrades could command premium pricing without significantly increasing acquisition costs.



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- 2800

# Business Recommendations & Technical Next Steps

### **Business Recommendations**

- Model Insight: The price difference between a 0.99ct and 1.0ct diamond of the same quality is 15-25%, far exceeding the actual weight difference (~1%).
- Business Problem: Consumers psychologically value whole/fractional carat sizes (e.g., 1.0ct vs. 0.9ct), but sellers
  often misprice diamonds near these thresholds.
- Actionable Recommendation:
  - For Sellers: Stock and market diamonds just below key thresholds (e.g., 0.9ct, 1.9ct) as "value alternatives" to 1.0ct/2.0ct stones.
  - For Buyers: Target 0.90-0.99ct diamonds for near-identical visual size at significant discounts.

- Model Insight: G/H color diamonds with VS clarity deliver 90% of the visual appeal of D-F colors but cost 30-40% less.
- **Business Problem:** Buyers overpay for D-F colors when near-colorless (G/H) diamonds appear identical to the naked eye in standard settings.
- Actionable Recommendation:
  - For Retailers: Curate "Smart Luxury" collections featuring G/H VS diamonds with side-by-side comparisons to premium grades.
  - For Marketing: Highlight that G/H colors are "eye-clean" in most settings, with savings better spent on upgrading carat size or cut quality.

## **Technical Next Steps**

- To Address Market Gaps:
  - Develop a separate sub-model for lab-grown diamonds, which follow different pricing dynamics but impact natural diamond demand.
- Collect Certification:
  - Develop Add GIA/IGI report metrics (e.g., fluorescence, symmetry grades) and auction/sales velocity data to better predict liquidity-adjusted pricing.
- Expand to Lab Grown Diamonds:
  - Incorporate synthetic diamond pricing data to model cross-category competition and substitution effects on natural diamond values.

## Appendix

## **Project Materials**

Git Repo: <u>Diamond Price Prediction</u>

Technical Version: <u>Code</u>