

# Price Homogeneity and Goods Classification: Technical Methodology

## Overview

This document describes a data-driven methodology for measuring price homogeneity in international trade and classifying goods into Rauch (1999) categories. The approach addresses three interconnected questions:

1. **Homogeneity:** How dispersed are prices across exporters within a product category?
  2. **Divergence:** Do large and small exporters exhibit systematically different pricing?
  3. **Classification:** Can we empirically identify homogeneous, reference-priced, and differentiated goods?
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## 1. Price Dispersion Measurement

### 1.1 Dominant Unit-of-Measure Restriction

Trade data often report quantities in multiple units of measure (UOM) within the same HS6 code (e.g., kilograms, items, litres). Comparing unit values across incompatible UOMs is meaningless, so we restrict analysis to the **dominant UOM**—the unit capturing the largest share of trade value within each HS6-year cell.

For each HS6 code  $h$  and year  $t$ :

$$\text{UOM}_{ht}^* = \arg \max_u \sum_e V_{ehtu}$$

where  $V_{ehtu}$  is trade value for exporter  $e$  in UOM  $u$ . We compute **coverage** as:

$$\text{Coverage}_{ht} = \frac{\sum_e V_{eht, \text{UOM}^*}}{\sum_{e,u} V_{ehtu}}$$

Observations with coverage below a threshold (default: 85%) are flagged as potentially unreliable.

### 1.2 Exporter-Level Unit Values

Within the dominant UOM, we compute exporter-level unit values by aggregating across importers:

$$P_{eht} = \frac{\sum_i V_{eih t}}{\sum_i Q_{eih t}}$$

where  $i$  indexes importers. This yields a single price observation per exporter-HS6-year.

### 1.3 Dispersion Metrics

We measure price dispersion using the **log interquartile ratio** between the 90th and 10th percentiles across exporters:

$$\text{LogGap}_{ht} = \log \left( \frac{P_{ht}^{(90)}}{P_{ht}^{(10)}} \right)$$

The **homogeneity ratio** is defined as:

$$H_{ht} = \frac{P_{ht}^{(10)}}{P_{ht}^{(90)}} = \exp(-\text{LogGap}_{ht})$$

where  $H \in (0, 1]$  and higher values indicate tighter price clustering.

We compute two variants:

Metric	Weighting	Interpretation
$H^{EQ}$	Equal-weight across exporters	Each exporter is an independent price signal
$H^{VW}$	Value-weighted by exporter trade	Reflects price distribution in actual trade flows

### 1.4 Aggregation to HS6 Level

Year-level estimates are aggregated to HS6-level using value-coverage weighted medians:

$$\tilde{H}_h = \text{wtd.median}(\{H_{ht}\}, \{V_{ht} \cdot \text{Coverage}_{ht}\})$$

This downweights years with poor UOM representation or low trade volume.

## 2. Divergence Testing

### 2.1 Motivation

If  $H^{EQ} \neq H^{VW}$ , large exporters occupy systematically different positions in the price distribution than small exporters. This divergence carries economic meaning:

- $\delta > 0$  ( $H^{VW} < H^{EQ}$ ): High-value exporters in the tails (premium or discount positioning)
- $\delta < 0$  ( $H^{VW} > H^{EQ}$ ): High-value exporters cluster near the median; small exporters are outliers

### 2.2 Test Statistic

Define:

$$\delta_{ht} = \text{LogGap}_{ht}^{VW} - \text{LogGap}_{ht}^{EQ}$$

Under the null hypothesis of no systematic price-value relationship, permuting exporter values  $V$  while holding prices  $P$  fixed should produce  $\delta$  values centered at zero.

## 2.3 Permutation Inference

For each HS6-year with  $n \geq 5$  exporters:

1. Compute observed  $\delta_{ht}^{obs}$
2. Generate null distribution: for  $r = 1, \dots, R$ , permute  $\{V_e\}$  and compute  $\delta^{(r)}$
3. Two-sided p-value:  $p = \frac{1}{R} \sum_r \mathbf{1} [|\delta^{(r)}| \geq |\delta^{obs}|]$

Bootstrap resampling (with replacement) provides confidence intervals for  $\delta^{obs}$ .

## 2.4 HS6-Level Inference

Year-level p-values are combined using **Fisher's method**:

$$\chi^2 = -2 \sum_t \log(p_{ht})$$

which follows a  $\chi^2_{2T}$  distribution under the null. Benjamini-Hochberg correction controls the false discovery rate across HS6 codes.

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## 3. Rauch Classification

### 3.1 Conceptual Framework

Following Rauch (1999), goods are classified as:

Category	Economic Meaning	Price Behavior
<b>Homogeneous</b>	Exchange-traded commodities	Tight clustering, law of one price
<b>Reference-priced</b>	Published benchmark prices	Moderate dispersion
<b>Differentiated</b>	Bilateral negotiation, quality variation	Wide price spread

### 3.2 Classification Features

We classify HS6 codes using two features:

1. **Dispersion level** —  $\text{Log}\hat{\text{Gap}}_h$  (median across years)
2. **Temporal stability** —  $\text{CV}_h = \frac{\text{sd}(\text{LogGap}_{ht})}{\text{mean}(\text{LogGap}_{ht})}$

The stability dimension captures the insight that truly homogeneous goods should exhibit not just low dispersion but *stable* dispersion over time.

### 3.3 Gaussian Mixture Model

We fit a GMM to the joint distribution of (LogGap, CV) across HS6 codes:

$$f(\mathbf{x}) = \sum_{k=1}^K \pi_k \mathcal{N}(\mathbf{x} \mid \boldsymbol{\mu}_k, \boldsymbol{\Sigma}_k)$$

Model selection ( $K = 2$  or  $3$  components) uses BIC. Components are ordered by a homogeneity score:

$$S_k = 0.7 \cdot \mu_k^{\text{LogGap}} + 0.3 \cdot \mu_k^{\text{CV}}$$

The component with lowest  $S_k$  is labeled “homogeneous,” highest is “differentiated.”

### 3.4 Stability Override

As a robustness check, HS6 codes classified as “homogeneous” but with CV above the 90th percentile are demoted to “reference-priced.” This prevents volatile goods from receiving the homogeneous label.

### 3.5 Fallback Hierarchy

For small samples where GMM estimation is unreliable:

Priority	Condition	Method
1	$n \geq 10$	2D GMM (dispersion + stability)
2	$n < 10$	1D GMM (dispersion only)
3	GMM fails	Quantile-based (33rd/67th percentiles)

## 4. Output Metrics

### 4.1 HS6-Level Summary

Variable	Definition
med_H_eq	Median equal-weight homogeneity ratio
med_H_vw	Median value-weight homogeneity ratio
cv_log_gap	Coefficient of variation of LogGap across years
rauch_category	Classification: homogeneous / reference / differentiated
posterior_prob	GMM posterior probability for assigned category
med_delta	Median divergence between VW and EQ measures
flag_divergence	TRUE if divergence is statistically significant (FDR-controlled)

### 4.2 Interpretation Guidelines

med_H_eq	Interpretation
0.8 – 1.0	Highly homogeneous (commodity-like)
0.5 – 0.8	Moderate dispersion (reference-priced)
0.2 – 0.5	High dispersion (differentiated)

flag_divergence	dominant_direction	Interpretation
TRUE	large_in_tails	Large exporters at price extremes (quality tiers?)
TRUE	large_in_center	Large exporters at market price; small exporters are outliers
FALSE	—	No systematic price-value relationship

## 5. Data Requirements

### 5.1 Input Fields

Field	Type	Description
<b>year</b>	integer	Observation year
<b>exporter</b>	string	Exporter country/entity
<b>importer</b>	string	Importer country/entity
<b>hs6</b>	string	6-digit HS code
<b>value</b>	numeric	Trade value
<b>quantity</b>	numeric	Trade quantity
<b>uom</b>	string	Unit of measure

### 5.2 Sample Size Considerations

Threshold	Default	Purpose
Minimum exporters per HS6-year	10	Reliable quantile estimation
Minimum coverage	85%	Dominant UOM representativeness
Minimum years for HS6-level	3	Temporal stability estimation
Minimum exporters for divergence test	5	Permutation test validity

## References

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