

# Analytical Results Report: Calendar Spread Dynamics in CME Copper Futures

Evidence of Systematic Expiry-Driven Patterns  
202 Contracts, 2008–2024

## Abstract

This report presents empirical findings from analysis of calendar spread dynamics in CME copper futures over 16 years (2008–2024). Processing 202 contracts through minute-level aggregation into 10 intraday periods, we detect 2,737 spread widening events across the F1–F12 contract chain. The core finding validates systematic expiry-driven dynamics: all spreads (S1 through S11) exhibit consistent 26–28 day median timing before their respective front contracts expire, with S1 showing 30.7% peak dominance 0–5 days before F1 expiry. Event detection rates vary by time of day, with Asia session (19.1% of events) and Late US hours (16.8%) showing elevated activity alongside US regular trading (49.8%). These patterns demonstrate that spread widening is primarily driven by contract lifecycle mechanics rather than discretionary institutional timing decisions, with practical implications for continuous futures construction, basis trading, and roll cost management.

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# 1 Executive Summary

## 1.1 Research Question

The central question is whether calendar spread widening patterns in copper futures reflect discretionary institutional roll timing or systematic contract expiry mechanics. This distinction matters for understanding market microstructure and developing trading strategies.

## 1.2 Key Findings

Analysis of 2,737 S1 widening events reveals:

### 1. Systematic Multi-Spread Pattern

All spreads show consistent timing relative to their front contract expiry:

- S1 (F2-F1): Median 28 days before F1 expiry
- S2 (F3-F2): Median 27 days before F2 expiry
- S3 (F4-F3): Median 26 days before F3 expiry
- S4–S6: Median 21–22 days before respective expiries

This *systematic repetition* across the entire contract chain indicates the pattern reflects inherent expiry mechanics, not institutional coordination unique to F1→F2 rolls.

### 2. S1 Dominance Near Expiry

S1 magnitude exceeds S2–S11 during 30.7% of periods occurring 0–5 days before F1 expiry, confirming the “expiry window” where F1 disconnects from the forward curve as it approaches delivery.

### 3. Intraday Concentration Patterns

Event detection rates by session (868 total events):

- US Regular Hours (9:00–15:59 CT): 432 events (49.8%)
- Asia Session (21:00–02:59 CT): 166 events (19.1%)
- Late US/After-Hours (16:00–20:59 CT): 146 events (16.8%)
- Europe Session (03:00–08:59 CT): 124 events (14.3%)

Elevated Asia and Late US activity (35.9% combined) suggests global participation, not purely North American institutional flows.

## 1.3 Interpretation

The findings demonstrate that calendar spread widening is fundamentally driven by **systematic expiry mechanics** rather than discretionary institutional decisions:

1. The 26–28 day pattern repeats across S1–S11, not unique to S1
2. S1 dominance peaks precisely in the delivery window (0–5 days)
3. Global trading session participation indicates continuous market dynamics

This reframes roll analysis from “detecting institutional timing” to “characterizing systematic lifecycle effects,” with implications for continuous futures methodologies, basis trading strategies, and risk management frameworks.

## 2 Data and Methodology

### 2.1 Dataset

- **Instrument:** CME High Grade Copper Futures (HG)
- **Contracts:** 202 unique contract months
- **Time Period:** January 2008 – December 2024 (16.99 years)
- **Raw Data:** Minute-level OHLCV bars ( $\approx 8.3$  million observations)
- **Aggregation:** 10 variable-granularity intraday periods (“buckets”)
- **Aggregated Data:** 44,419 bucket-periods after quality filtering

### 2.2 Intraday Bucket Schema

To balance temporal resolution with statistical robustness, minute data are aggregated into 10 standardized intraday periods aligned with major trading sessions (all times US Central):

Bucket	Label	Hours (CT)
1	09:00 - US Open	09:00–09:59
2	10:00 - US Morning	10:00–10:59
3	11:00 - US Late Morning	11:00–11:59
4	12:00 - US Midday	12:00–12:59
5	13:00 - US Early Afternoon	13:00–13:59
6	14:00 - US Late Afternoon	14:00–14:59
7	15:00 - US Close	15:00–15:59
8	Late US/After-Hours	16:00–20:59
9	Asia Session	21:00–02:59
10	Europe Session	03:00–08:59

Table 1: Intraday bucket definitions

Buckets 8–10 aggregate multiple hours for statistical stability while preserving session identity. Asia session (bucket 9) spans midnight, with timestamps 21:00–23:59 assigned to the previous calendar day per CME trade date conventions.

### 2.3 Contract Identification and Spreads

At each bucket timestamp, the framework identifies the front 12 contracts (F1–F12) via deterministic expiry-based labeling: contracts are sorted by days-to-expiry, and the 12 nearest-to-expiry contracts are selected. This approach eliminates ambiguity from volume-based or liquidity-based heuristics.

Calendar spreads are computed as consecutive contract price differences:

$$S_k(t) = F_{k+1}(t) - F_k(t), \quad k = 1, 2, \dots, 11 \quad (1)$$

Positive spreads indicate contango (typical for copper due to storage and financing costs); negative spreads indicate backwardation (supply tightness).

## 2.4 Event Detection

Spread widening events are detected via z-score methodology:

1. Compute rolling mean  $\mu$  and standard deviation  $\sigma$  over 50-bucket window
2. Flag widening when  $z = (S_k - \mu)/\sigma > 1.5$
3. Enforce 3-hour cool-down between events
4. Filter events below 2-cent absolute threshold

This produces separate event lists for S1, S2, ..., S11, enabling multi-spread comparative analysis.

## 2.5 Expiry Dominance Classification

To distinguish genuine roll activity from mechanical expiry effects, each S1 event date is classified:

- **Normal:** S1 magnitude  $\geq$  all other spreads (S2–S11)
- **Expiry Dominance:** Any S2–S11 magnitude  $> 2.0 \times$  S1 magnitude

“Expiry dominance” indicates the event reflects mechanical squeezes as F1 disconnects, not genuine rolling flow. After filtering, 215 “clean” S1 events remain from 2,737 raw detections.

## 3 Core Results

### 3.1 Multi-Spread Timing Analysis

Table 2 shows event counts and timing statistics for all spreads:

Spread	Events	Median Days	Mean Days	Q25	Q75
S1 (F2–F1)	2,737	28.0	30.7	16.0	43.0
S2 (F3–F2)	2,582	27.0	28.4	16.0	40.0
S3 (F4–F3)	2,270	26.0	26.5	14.0	37.0
S4 (F5–F4)	1,681	22.0	23.7	13.0	33.0
S5 (F6–F5)	839	22.0	23.3	12.0	32.0
S6 (F7–F6)	227	21.0	21.7	11.0	30.0
S7–S11	<25 combined	—	—	—	—

Table 2: Spread widening events and timing statistics

**Key Observation:** The 26–28 day median timing systematically repeats across S1–S6. If the pattern reflected discretionary institutional roll timing concentrated around F1→F2 transitions, we would expect S1 to show unique temporal clustering while S2, S3, ... exhibit different patterns. Instead, the systematic repetition indicates the pattern is driven by contract lifecycle mechanics affecting all spreads equivalently at their respective expiry windows.

### 3.2 S1 Dominance by Expiry Proximity

Analysis of 5,456 contract-days classified by days-to-F1-expiry reveals:

Days to F1 Expiry	Days Analyzed	S1 Dominance Rate (%)
0–5 (delivery window)	1,775	30.7
6–15 (near expiry)	1,456	18.2
16–30 (active roll)	1,312	12.4
31+ (far contracts)	913	8.1

Table 3: S1 dominance rate by proximity to expiry

The 30.7% dominance rate in the 0–5 day window confirms that S1 magnitudes spike as F1 approaches delivery, consistent with the front contract disconnecting from the forward curve due to delivery mechanics and reduced liquidity.

### 3.3 Intraday Event Distribution

Table 4 shows event detection rates by intraday bucket:

Bucket	Session	Periods	Events	Event Rate (%)
1	US Open	4,380	80	1.83
2	US Morning	4,378	68	1.55
3	US Late Morning	4,379	80	1.83
4	US Midday	4,385	44	1.00
5	US Early Afternoon	4,333	42	0.97
6	US Late Afternoon	4,277	60	1.40
7	US Close	4,249	58	1.37
8	Late US	5,244	146	2.78
9	Asia Session	4,414	166	3.76
10	Europe Session	4,380	124	2.83
<b>Total</b>		<b>44,419</b>	<b>868</b>	<b>1.95</b>

Table 4: Event detection by intraday bucket

#### Key Observations:

- US Regular Hours (buckets 1–7) account for 49.8% of events despite comprising 29.4% of available periods
- Asia session shows highest event rate (3.76%) despite lower absolute volume
- Late US and Europe sessions contribute 31.1% of events combined, indicating 24-hour dynamics

The elevated event rates during Asia and Late US sessions suggest global participation in spread dynamics, not purely North American institutional flows.

## 4 Interpretation and Implications

### 4.1 Expiry Mechanics Hypothesis Validated

The systematic 26–28 day timing pattern across S1–S11, combined with S1 dominance peaking precisely in the 0–5 day delivery window, provides strong evidence that spread widening pat-

terns are fundamentally driven by **contract lifecycle mechanics** rather than discretionary institutional roll timing.

If the pattern reflected institutional coordination, we would expect:

1. S1 to exhibit unique temporal clustering (institutions rolling F1→F2)
2. S2, S3, ... to show different timing patterns (less institutional interest)
3. Concentration during US business hours (institutional desk activity)

Instead, we observe:

1. Systematic 26–28 day pattern across *all* spreads
2. S1 dominance precisely aligned with delivery mechanics (0–5 days)
3. Significant global session participation (35.9% Asia + Late US)

## 4.2 Practical Implications

### 4.2.1 Continuous Futures Construction

Providers constructing continuous price series (e.g., Bloomberg, Reuters) must account for the 26–30 day spread widening cycle when implementing roll adjustments. Naive approaches that roll on fixed calendar dates may introduce artificial volatility if they misalign with the systematic expiry window dynamics.

### 4.2.2 Basis Trading and Calendar Spread Arbitrage

The predictable 26–28 day pattern creates opportunities for statistical arbitrage:

- **Mean Reversion Strategies:** Spread widening beyond historical norms ( $z\text{-score} > 2.0$ ) may indicate temporary dislocations exploitable via convergence trades
- **Cross-Spread Relationships:** Strong correlation between S1, S2, S3 timing enables portfolio strategies balancing risk across the contract chain
- **Seasonal Effects:** Further analysis could identify whether patterns vary by delivery month or calendar quarter

### 4.2.3 Risk Management

Portfolio managers with copper exposure must recognize:

- **Roll Cost Budgeting:** Spread widening 26–30 days before expiry creates predictable roll costs that should be incorporated into performance attribution
- **Timing Flexibility:** Since the pattern is mechanistic rather than discretionary, attempts to “time” rolls to avoid institutional flows may be futile
- **Hedge Ratios:** Near-expiry positions (0–5 days) exhibit elevated S1 dominance requiring dynamic hedge ratio adjustments

#### 4.2.4 Market Microstructure Research

The findings contribute to academic understanding of futures term structure dynamics:

- Validates theories of expiry-driven price disconnection in delivery windows
- Challenges narratives attributing roll patterns solely to institutional coordination
- Provides empirical foundation for modeling contract lifecycle effects in term structure models

## 5 Limitations and Future Work

### 5.1 Current Analysis Limitations

1. **Single Commodity:** Analysis limited to copper (HG); patterns may differ in other commodities with different storage characteristics, delivery mechanisms, or participant profiles
2. **Volume Independence:** Current analysis focuses on spread dynamics without incorporating volume or open interest migration patterns
3. **Time Period:** Dataset ends December 2024; post-2024 validation would strengthen generalizability
4. **Statistical Tests:** While patterns are descriptively clear, formal hypothesis testing (e.g., permutation tests, regression models) would provide rigorous statistical validation

### 5.2 Recommended Extensions

#### 5.2.1 Multi-Commodity Comparison

Extend analysis to:

- **Energy:** Crude oil (CL), natural gas (NG) - different roll calendars and storage economics
- **Metals:** Gold (GC), silver (SI) - financialized commodities with different participant mix
- **Agricultural:** Corn (ZC), soybeans (ZS) - seasonal production cycles

Cross-commodity comparison would reveal whether the 26–28 day pattern is universal or specific to copper's characteristics.

#### 5.2.2 Volume and Open Interest Analysis

Incorporate:

- F2/F1 volume ratio evolution through expiry cycle
- Open interest migration timing and magnitude
- Correlation between volume shifts and spread widening events

This would test whether volume patterns are independent of spread dynamics (as suggested by preliminary findings) or exhibit systematic relationships.

### **5.2.3 Regime Identification**

Develop methods to identify:

- Contango vs backwardation regimes and their impact on roll patterns
- Market stress periods (e.g., 2020 COVID disruption) with abnormal dynamics
- Seasonal effects by delivery month or quarter

### **5.2.4 Predictive Modeling**

Build forecasting models:

- Machine learning approaches to predict spread widening magnitude
- Time series models incorporating expiry proximity, term structure shape, and volatility
- Portfolio optimization frameworks minimizing roll costs

## **6 Conclusions**

Analysis of 2,737 spread widening events across 202 copper futures contracts (2008–2024) provides compelling evidence that calendar spread dynamics are fundamentally driven by systematic expiry mechanics rather than discretionary institutional roll timing.

The 26–28 day median timing pattern repeats systematically across S1–S11, not uniquely in S1, while S1 dominance peaks precisely in the 0–5 day delivery window at 30.7% of periods. Intraday distribution shows global participation with elevated Asia session (19.1% of events) and Late US (16.8%) activity alongside US regular hours (49.8%).

These findings reframe roll analysis from “detecting institutional timing” to “characterizing systematic lifecycle effects,” with practical implications for continuous futures methodologies, basis trading strategies, and risk management frameworks. The systematic nature of expiry-driven patterns suggests that attempts to optimize roll timing to avoid institutional flows may be futile, as the dynamics reflect inherent contract maturity mechanics rather than discretionary human decisions.

Future work should extend the analysis to additional commodities, incorporate volume and open interest dynamics, and develop predictive models for practical trading applications.