Financialisation of Commodity Markets Co-movement Behind-the-Scenes

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Abstract

In the early 2000s institutional investors entered the commodity futures markets en masse with passive, long only, index type positions in sharp contrast with those typically assumed by traditional expert participants. A heated public debate soon erupted over the perceived consequences of the phenomenon–commonly referred to as "financialisation"—and, in response to immediate policy concerns, the matter was thrust onto the academic sphere as a burning issue. With the benefit of hindsight it now seems that the academic debate was framed rather narrowly, used contentious research methods, and eventually led to regulatory changes that were therefore perhaps unwarranted. In contrast, we take a broader approach where we consider a large cross-section of liquid commodities as suggested by the nature of financialisation that comprehends commodity futures as an asset class. Using a bespoke asset pricing factors based framework we study the issue through the lens of co-movement and find that the phenomenon had ontological consequences for the commodity complex with its impact extending beyond the mechanical effects induced by indexation. The onset of the financial crisis and the monetary policy regimes that followed, on the other hand, seem to have set off a motion of reversion to legacy pre-financialisation fundamentals.

Keywords: Commodity markets, Financialisation, Asset pricing

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1. Introduction

The last twenty years have seen major upheavals in the commodity markets with the onset of financialisation as well as the period of the 2008 financial crisis and its aftermath. Financialisation appears to have affected multiple economic sectors, including agriculture and energy, and its impact has been hotly debated in the policy, legislative, regulatory, and academic spheres. A vigorous legislative debate accompanied by an equally intense policy and academic debate across a number of disciplines about the perceived effects on commodity prices of the unprecedented inflow of institutional funds brought into commodity futures markets by new index type investors² eventually led to regulatory changes in several individual markets.

In response to immediate policy concerns, the academic debate was initially framed rather narrowly around adequacy of speculation—the burning issue of the day—at the individual commodity market level and the ensuing academic analysis focused on the more mechanical effects of financialisation. With the benefit of hindsight, this approach seems to lack a fundamental understanding of the phenomenon it endeavoured to study and which consequences it attempted to address.

Financialisation was a phenomenon global in nature; through the channel of index investment it affected the whole cross-section of liquid commodities at once, eventually spurring its transition away from a collection of idiosyncratic markets to its emergence as an asset class. Focusing on individual markets in this context thus seems limited. Most of the studies underpinning the regulatory response to financialisation took the form of pairwise correlation, causation and to some extent co-integration analysis, all of which are idiosyncratic in nature for they study single pairs of individual assets. Besides, some of these techniques since proved controversial and seem to produce results the quality of which may fall short of the standards that may reasonably be expected to lay the ground for legislative action. Although appealing, for it comes with the promise of prompt "significant" results, this approach misses a critical corollary of the substantial advent of a new category of players in a (set of) market(s): the alteration of market fundamentals it potentially induces, best observed in this context through the lens of commercial hedging pressure (Keynes, 1930; Hicks, 1939).

While essential to the understanding of the phenomenon, the global nature of financialisation and its impact on market fundamentals therefore appear to have been originally overlooked and another approach, perhaps more refined, seems warranted.

We strive to provide one in this study by taking a fundamentally different empirical perspective that is broader in scope. We consider the entire cross-section of actively traded commodities on futures markets and use a futures-based asset pricing framework that includes factors constructed using both theoretical fundamentals and liquidity considerations.

The global nature of financialisation makes cross-sectional co-movement a cornerstone issue that the direct impact of increased market participation alone might not be able to fully explain for it potentially had deeper effects, for example on market fundamentals as our results suggest. Commodity futures markets have a history of being the classic physical commodity price risk shifting conduit for commodity specialists. Over the financialisation period we observe a decrease in effectiveness of commercial hedging pressure (CHP) at explaining commodity futures returns at the individual level accompanied by a more pronounced increase at the aggregate level. We further note an overall shift in the nature of hedger's behaviour away from the traditional Keynesian view of risk transfer with higher individual risk premiums over phases of cross-sectional contango as opposed to backwardation. Both patterns suggest that the phenomenon may have altered this long-standing paradigm of commercial risk transfer with the entry of long only commodity index traders (CITs) with very different hedging and volume demands from traditional participants. This change could have altered the hedging ontology in these markets and our asset pricing framework is tailored to detect

¹Starting around 2004, with a view of commodity markets as an emerging asset class and in an effort to hedge against inflation and seek diversification benefits (Büyükşahin and Robe, 2014; Singleton, 2013), institutional investors sent forth an unprecedented flow of capital into commodity futures markets. This phenomenon is commonly referred to as the financialisation of commodities (Domanski and Heath, 2007).

²See section Section 2.1 for a detailed discussion.

these more subtle effects driven by changes in the nature of market participation.

Factor model techniques are well suited to isolating common driving factors and detecting co-movement (Fama and French, 1993; Carhart, 1997; Asness and Frazzini, 2013; Fama and French, 2015; Hou et al., 2015; Asness et al., 2015; Frazzini and Pedersen, 2014; Asness et al., 2019) at a broader level, beyond individual pairs of assets, and are commonly used to study commodity market fundamentals (Schwartz and Smith, 2000; Miffre and Rallis, 2007; Gorton et al., 2012; Cortazar et al., 2013; Yang, 2013; Daskalaki et al., 2014; Szymanowska et al., 2014; Fernandez-Perez et al., 2018; Bakshi et al., 2019; Boons and Prado, 2019; Sakkas and Tessaromatis, 2020); they consequently seem well adapted to study these two issues simultaneously. Building on theory we therefore develop a bespoke commodity-based asset pricing framework comprised of factor models based both on fundamentals as well as liquidity considerations.

One of these is based on the Keynes-Hicks (Keynes, 1923, 1930; Hicks, 1939, 1946) risk transfer approach to the term structure; the corresponding "theory of normal backwardation", complemented for the possibility of contango (Houthakker, 1957a; Cootner, 1960), postulates that futures prices for a given commodity are inversely related to the extent that commercial hedgers are short or long and using commercial hedging pressure (CHP) as a proxy for hedgers net market position we construct factor mimicking portfolios that capture the impact of hedging pressure as a systemic factor (Basu and Miffre, 2013). The cross-sectional nature of financialisation (Cheng and Xiong, 2014; Basak and Pavlova, 2016) further leads us to extend this paradigm by incorporating aggregate CHP (CHP), a market wide measure of hedging pressure first considered in Hong and Yogo (2012). In this context, returns for individual commodities should be high in periods where CHP is low (backwardation) and low in periods where CHP is high (contango). Working's contending approach that relates futures price formation to storage costs in the theory of "the price of storage" (Working, 1949) forms the basis of another model that we consider where, following Szymanowska et al. (2014) and Fuertes et al. (2015) who demonstrate that the term structure factor has explanatory power for the cross-section of commodity returns, we construct factor mimicking portfolios using roll-yield as a proxy for the front end shape of individual commodity term structures. We further consider a liquiditybased model following Hong and Yogo (2012) who find that open interest growth has predictive power for individual commodity returns and again, using the latter measure as a proxy for individual commodity liquidity, we construct long-short factor mimicking portfolios.

We study the performance of these models on a set of twenty-four US traded commodities and six UK traded (London Metal Exchange) base metals over the 1997/2018 period that we divide into four distinct sub-periods: pre-financialisation (1997/2003); financialisation ($2004/2008^3$); financial crisis and aftermath, including loose monetary policy regimes that followed (2008/2013); and post-crisis (2014-2018).

Consistent with earlier studies, we observe that mean returns for the US traded commodities are sharply higher over the financialisation period compared to pre-financialisation and this pattern extends to UK traded metals.

Over the first period the cross-sectional average returns during phases of **backwardation** for both the US and UK commodities are positive and overall higher than during phases of **contango**, providing broad support for an aggregate version of the Keynesian hedging pressure hypothesis. Over the second period this pattern reverses, with both the US and UK commodities having higher returns during phases of **contango** as opposed to **backwardation**. This change in pattern suggests the onset of financialisation altered the traditional risk shifting nature of hedger's behaviour, implied by the Keynesian hypothesis.

This change of paradigm due to the arrival of financial investors, on the other hand, raises the question of whether they functioned as liquidity providers or demanders. Arguments have been put forward in favour of both possibilities with Moskowitz et al. (2012) arguing that financial investors provide liquidity and Kang et al. (2020) taking the opposite stance. Our analysis of individual commodity returns' relation with both

³Starting point based on earlier studies (Baker, 2021; Christoffersen et al., 2014).

CHP and CHP before and during financialisation sheds a new light on this issue with our results suggesting that financial investors may have been liquidity demanders leading hedgers to become liquidity providers.

Goldstein et al. (2014) and Goldstein and Yang (2022) argue that the arrival of financial investors had an impact on the extent of normal backwardation in commodity futures markets. Based on their analysis we would expect our average CHP factor, based on backwardation as measured by hedger's positions (Keynesian perspective, theory of normal backwardation: Keynes (1930), Hicks (1939), Houthakker (1957b), Telser (1958)), as well our term structure factor, based on backwardation as measured by roll yield (Working's perspective, theory of storage: Working (1933), Kaldor (1939), Working (1948), Brennan (1958), Cootner (1960), Weymar (1966), Danthine (1978), Turnovsky (1983), Schwartz (1997)), to be best suited to measure the impact of financialisation.

We analyse the pricing dynamics over the different periods by first studying the time series explanatory performance of the various factors on all the US traded commodities with the factors constructed using the latter set of commodities. The market-based factor outperforms the other three in the first period and, although its performance increases over the financialisation period, that of the CHP factor shows a greater improvement in a relative sense. The performance of the market factor is driven by agricultural and energy commodities in the first period while metals strongly contribute to the performance improvement observed over the second period. Similarly, the dramatic improvement of the CHP factor in the second period, is driven by the metals complex, particularly that of the long leg where they predominantly load. Average pairwise correlations amongst individual factors' best performing commodities increase across the board over the financialisation period, far beyond the average for the whole cross-section of US traded commodities. The increase is particularly strong for the long-short factors, in a relative sense, on which the metal complex loads significantly more during financialisation. This increase is due to the dramatic rise in average pairwise correlation amongst the US traded metals all of which exhibited high R^2 with respect to the CHP factor during financialisation which in turn is due to their common low level of CHP over this period. The CHP factor model thus seems able to establish the strengthening link between a non-price attribute and price dynamics in the US metal sub-sector during financialisation.

We explore this issue further by constructing the four factors independently with the top eight commodities for a given factor (factor picks) in each period⁴ and evaluating its performance on these selected commodities as well as on all US traded commodities. The results of this exercise provide evidence of convergence of Keynesian and Working's theories of the term structure in detecting price co-movement during financialisation and strongly suggest that the term structure related attributes (hedging pressure/roll) co-movement and price co-movement link was concentrated in the metals sub-sector.

The UK traded base metals provide a useful set of assets to test how widespread were the changes in pricing dynamics brought about by the onset of financialisation. To that end, we examine the performance of the four factors on the six UK metals, when constructed from each set of factor picks. The results show that there was co-movement in global metals returns in the second period and that this co-movement could be detected by our term structure related models, particularly so by our hedging pressure-based model.

The onset of the financial crisis and the monetary policy regimes that followed, on the other hand, also appear to have induced significant changes. Commodity futures returns fell dramatically into the crisis and the pricing results indicate the emergence of a systematic factor across the entire cross-section of the US commodities as well as evidence of cross-market linkages with the UK traded commodity subset. The post-crisis results, in contrast, show a substantial pull-back in performance for the market factor across the board as well as evidence of decoupling between the US and UK markets. Besides, starting in the crisis the Keneysian paradigm of returns begins to revert with higher mean returns during low (backwardation) relative to high (contango) CHP phases and this pattern continues, more pronounced, over the post-crisis period. As pointed out in Basu and Bauthéac (2024), the financial crisis and ensuing accommodative mon-

⁴See section Section 2.3 for further details.

etary policy therefore seem to have initiated a process of reversion toward pre-financialisation fundamentals in the global commodity futures complex.

Financialisation was an issue of such policy importance that it triggered legislative action. The debate was initially framed around adequacy of speculation, the burning issue of the day, and the ensuing academic analysis focused on the more mechanical effects of financialisation.

With the benefit of hindsight, it appears this approach was perhaps too narrow, and it now seems necessary to address the phenomenon from a broader perspective. Commodity price dynamics appear to have altered substantially in quite different ways over the financialisation and the financial crisis periods and our commodity futures asset pricing factors based approach seems able to provide new insights into the nature of these changes: financialisation was a phenomenon endogenous to the commodity markets transmitted via the commodity futures markets and which effects were particularly strong for the metals sector while the crisis and its aftermath seems to have delivered an exogenous shock across the entire cross-section of liquid commodities and thereby potentially triggering a reversion to pre-financialisation dynamics.

The initial view of financialisation was that it consisted of speculative flows which had the effect of driving up and creating bubble like conditions in commodity futures prices⁵. However, the fundamental question of the nature of the impact of financialisation across the entire cross-section of commodity futures markets has not yet been completely answered. We provide an empirical complement to a new stream of theoretical studies that try to model the impact of financialisation on various aspects of commodity futures markets⁶ and demonstrate that the effects of financialisation extend beyond the mechanical effects induced by indexation as outlined in Tang and Xiong (2012) and Basak and Pavlova (2016).

⁵Masters (2008); Masters and White (2008); UNCTAD and Cooperation (2009); De Schutter (2010); Gilbert (2010a); Gilbert (2010b); Herman et al. (2011); Schumann (2011); Singleton (2013).

⁶Etula (2013), Acharya et al. (2013), Cheng et al. (2014), Leclercq and Praz (2014), Sockin and Xiong (2015), Goldstein et al. (2014), Ekeland et al. (2016), Goldstein and Yang (2022), Ekeland et al. (2019), Isleimeyyeh (2020).

2. Data & methods

2.1. Background

The period from 1998 to 2008 saw an influx of investment into commodity index linked products most of which made its way into the commodity futures market. Popular commodity indexes whose goal was to track the broad movement of commodity prices became accessible by means of swaps, exchange traded funds and exchange traded notes and attracted at least \$100 billion of net super-hoc investment over the 1998-2008 period.

Concerns over the consequences of this financialisation phenomenon eventually led to legislative changes including the approval of Rule 76 FR 4752 issued by the US Commodity Futures Trading Commission (CFTC) on January 26th, 2011. This provision emanates from the Dodd-Frank Wall Street and Consumer Protection Act of 2010 (Title VII, Section 737) that mandates the CFTC to use position limits to restrict the flow of speculative capital into a number of commodity markets. The Rule was approved in a close 3-2 vote and the ensuing rule-making process was extremely contentious with several commissioners (Michael Dunn and Scott D. O'Malia in particular) expressing reservations about the lack of supporting evidence and the Rule also triggering thousands of comment letters as well as a lawsuit against the CFTC spearheaded by two Wall Street trade groups, the International Swaps & Derivatives Association (ISDA) and the Securities Industries & Financial Markets Association (SIFMA)⁷.

A world-wide debate ensued about the role of index funds in commodity markets. The first responses to the 2007/2008 crisis of escalating food and energy prices took the form of policy reports, many of which reasoned that the growth of commodity index funds came along with an influx of largely speculative capital that was responsible for driving commodity prices beyond their historic highs (De Schutter, 2010; Gilbert, 2010b; Herman et al., 2011; Schumann, 2011; UNCTAD and Cooperation, 2009).

Early US Senate investigations on the matter drew on pricing and trading data supplied by the CFTC as well as interviews with numerous experts including hedge fund manager Michael Masters who linked the growing presence of index investors in US agricultural and energy markets to the observed surges in both futures and cash prices (Masters, 2008; Masters and White, 2008). This contention has since come to be known as the Masters' hypothesis and was largely endorsed by the final US Senate report on the issue (Senate, 2009).

At this stage, the question thrust on the academic sphere was whether "excessive speculation" (understood as the market activity peculiar to index type investors) was linked to escalating energy and agricultural commodity prices. A number of ensuing academic studies, most of which used commodity specific correlation and causality based analysis have since disproved this contention: Irwin et al. (2009), Sanders et al. (2010), Sanders and Irwin (2011a), Sanders and Irwin (2011b), Irwin (2013), Brunetti and Reiffen (2014), Hamilton and Wu (2015) and Bruno et al. (2017) investigated the issue in the context of the agricultural markets; meanwhile, Büyükşahin and Harris (2011), Tokic (2012), Fattouh et al. (2013), Kilian and Murphy (2014), Knittel and Pindyck (2016) and Manera et al. (2016) examined the energy markets while Bohl and Stephan (2013), Kim (2015) and Boyd et al. (2016) studied both energy and agricultural markets and Irwin and Sanders (2011), Irwin and Sanders (2012) as well as Stoll and Whaley (2011) examined the commodity markets in general. Other studies have underlined various alterations in commodity pricing dynamics over the period. Gilbert and Pfuderer (2014) reject the view that financialisation has not had any effects in the grains markets and demonstrate that trades originated by financial market participants, and specifically index investors, can move prices but tend to be typically volatility-reducing. Juvenal and Petrella (2015) contend that the oil price increase between 2004 and 2008 was mainly driven by the strength of global demand but that the financialisation process of commodity markets also played a role. Likewise, Henderson et al. (2015) show that non-information-based financial investments have important impacts on commodity prices. Cheng and Xiong (2014), on the other hand, show that in the case of index commodities, financialisation has transformed the risk sharing, and information discovery functions of commodity futures markets.

 $^{^7 \}mathrm{ISDA}$ & SIFMA v US CFTC; Complaint, 1:11-cv-02146; December $2^{\mathrm{nd}},\,2011.$

2.2. Data

Commodity index swaps, commodity index funds (CIFs), commodity-based exchange traded funds (ETFs) and notes (ETNs) or commodity linked notes (CLNs) were amongst the main investment vehicles that allowed institutional and retail investors to build up commodity exposure in the early 2000s (Boons et al., 2012; Henderson et al., 2015; Irwin and Sanders, 2011; UNCTAD and Cooperation, 2009; Schumann, 2011). Most of these products were primarily designed to track prominent commodity indexes among which the S&P-GSCI, the most popular.

Most of its constituents at the time of writing form the basis of the broad cross-section of commodities that we consider in this study: corn-#2 yellow (XCBT), oats (XCBT), soybean meal (XCBT), soybean oil (XCBT), soybeans (XCBT), wheat-SRW (XCBT), cattle-feeder (XCME), cattle-live (XCME), lean hogs (XCME), cocoa (IFUS), coffee-C (IFUS), cotton-#2 (IFUS), lumber (XCME), orange juice (IFUS), sugar-#11 (IFUS), natural gas (XNYM), crude oil-WTI (XNYM), gasoline (XNYM), heating oil (XNYM), copper (XCEC), gold (XCEC), palladium (XNYM), platinum (XNYM) and silver (XCEC). Unleaded gasoline, the NYMEX legacy contract for gasoline, stopped trading in 2006 when it was replaced by Reformulated Gasoline Blendstock for Oxygen Blending (RBOB) with the two futures series trading alongside for most of the year. For our NYMEX gasoline data series, we consider unleaded gasoline up to September 1st, 2006 and RBOB gasoline thereafter, date at which liquidity for the latter overtook that for the former.

We also consider the UK metal complex with aluminium (XLME), copper (XLME), lead (XLME), nickel-primary (XLME), tin-refined (XLME) and zinc (XLME).

The commodity futures trading commission (CFTC) operates a comprehensive system of collecting information on market participants known as the large trader reporting program (LTRP) where it records and reports to the public position data on market participants who have position levels in excess of a particular market specific threshold⁸. The commission collects market data and position information daily from clearing members, futures commission merchants (FCMs) and foreign brokers and publishes corresponding summaries weekly in a series of market specific reports. In these reports, individual traders are categorised according to the nature of their trading activity on the basis of self-reported information⁹ that is subject to review by CFTC staff for reasonableness. Position information is then aggregated by category and asset type with each report providing an aggregation for a set of report specific categories and, for each category, a breakdown between futures only positions and positions in futures and options combined.

In its legacy report, the commitment of trader report (COT) with data dating back to 1962, the CFTC distinguishes "commercial" and "non-commercial" market participants where a "commercial" participant is defined as one "[...] engaged in business activities hedged by the use of futures and option markets" 10. A third category, "non-reportable", aggregates positions for participants not meeting the reporting threshold. In response to concerns related to category accuracy 11 the CFTC now refines its classification in the disaggregated commitment of trader report (DCOT) where participant categories include "Producer/Merchant/Processor/User", "Swap dealer", "Money manager" and "Other reportable" as well as in the supplemental commitment of trader report (SCOT) that details commodity index trader aggregate positions for thirteen agricultural commodity futures markets. Data for DCOT and SCOT date back to 2006 and 2007 respectively.

We study the 1997/2018 period where we observe futures only position data from the COT report, the only CFTC report with data available for the whole period of interest; as well as futures term structure price and open interest data from Bloomberg.

2.3. Methods

On the one hand, futures markets have a long history of serving commodity producers to alleviate their commodity price and output risks. Keynes (1930) and Hicks (1939) emphasise the "normal" behaviour of

⁸Current reporting level thresholds available at: www.ecfr.gov.

⁹CFTC form 40. Available at: www.cftc.gov.

 $^{^{10}\}mathrm{See}$ CFTC regulation 1.3, 17 CFR 1.3(z) for details.

¹¹See for example Ederington and Lee (2002).

naturally short hedgers outsourcing their commodity risk to naturally long commodity specialist speculators. This market configuration is referred to as "backwardation" where futures are expected to trade at a discount to spot prices, providing speculators with incentives to take the long side. The opposite market paradigm is referred to as "contango". Working (Working, 1933, 1948, 1953) and Kaldor (Kaldor, 1939) somehow relax this "supply-of-speculative-services" (Till, 2007) approach by introducing processors and merchants, also commodity experts, on the hedgers side and relating the notions of backwardation and contango more directly to the term structure via the theory of storage where the shape of the front end of the term structure (basis) is directly related to the market price of storage.

Financialisation, on the other hand, refers to the entry of financial investors into the commodity futures markets who, for the most part, are not commodity experts. Besides, these new market participants tend to exhibit herding like behaviour, often taking massive long-only positions on the whole cross-section of indexed commodities in an attempt to enhance returns while hedging against inflation and reaping further diversification benefits for portfolios with, in most cases, existing large positions across various asset classes (Brunetti et al., 2016; Boyd et al., 2016; Cheng et al., 2014; Juvenal and Petrella, 2015; Singleton, 2013; Tang and Xiong, 2012). This contrasts with the traditional approaches of Keynes and Working and in this context the issues of co-movement and aggregate market participants behaviour seem particularly relevant.

By grouping together assets by shared attribute dynamics in common portfolios, asset pricing techniques—and factor models in particular—are well suited to study co-movement (Fama and French, 1993; Carhart, 1997; Asness and Frazzini, 2013; Fama and French, 2015; Hou et al., 2015; Asness et al., 2015; Frazzini and Pedersen, 2014; Asness et al., 2019). Careful, theory-based, attribute selection further enables a refined analysis of market fundamentals (Schwartz and Smith, 2000; Miffre and Rallis, 2007; Gorton et al., 2012; Cortazar et al., 2013; Yang, 2013; Daskalaki et al., 2014; Szymanowska et al., 2014; Fernandez-Perez et al., 2018; Bakshi et al., 2019; Boons and Prado, 2019; Sakkas and Tessaromatis, 2020). We therefore develop a futures-based asset pricing framework that includes factors constructed using both theoretical and liquidity considerations and use this framework to carry out our analysis over four periods of interest. The first, the pre-financialisation period, starts in 1997 and is naturally followed by financialisation, spanning the 2004/2008 period, and the financial crisis and its aftermath, spanning the 2008/2013 period. The last period spans the 2014/2018 period starting with the beginning of the tapering/end of the US Quantitative Easing program (QE).

Using the weekly price, open interest and market position data described above we construct market portfolios of individual commodity nearby futures as well as mimicking portfolios for market, commercial hedging pressure, term structure and liquidity risk factors in returns. While our market portfolios are long only, equally weighted portfolios of particular commodity sets, our other mimicking portfolios for risk factors are constructed as combinations of two sub-portfolios (legs), one held long, one held short, which respective constituents are pooled on a weekly basis according to risk factor specific criteria.

For each leg, the position weight for a particular asset for a particular period is contingent on the value of the corresponding sorting variable and its rank relative to the other assets in the pool. The sorting variable values for the asset pool are first rescaled so that the highest (lowest) value becomes 1 (0). Weights are then determined pairwise, starting from the extremes with the highest and lowest figures. A coefficient α is determined by dividing their difference (1 in this case) by the mean of the rescaled values. The high figure is attributed a score of $\frac{1}{n} \times \alpha$ where n is the number of assets in the pool while the score for the low figure is $\frac{1/n}{\alpha}$. The operation is repeated for the pair of assets showing the second highest and second lowest values for the sorting variable and so on until all the assets in the pool are attributed a score. The individual asset weights are then determined by dividing each score by the sum of the scores.

The period return for a particular mimicking portfolio is calculated as the difference between the weighted average return for constituents of the long portfolio and that for the constituents of the short portfolio. i.e, return on the mimicking portfolio for risk factor i (f_i) for period t:

$$r_t^{f_i} = \sum_{j=1}^x w_{j,t}^{long} \times r_{j,t} - \sum_{j=n-x}^n w_{j,t}^{short} \times r_{j,t}$$

 $n \equiv$ number of commodities in the set considered for mimicking portfolio construction.

 $s \equiv \text{selection threshold } (\frac{1}{3} \text{ here}).$

 $w_{j,t}^* \equiv \text{period } t \text{ weight for commodity } j \text{ on the long or short leg.}$

 $r_{i,t} \equiv \text{period } t \text{ front contract return for commodity } j.$

 $j \equiv \text{commodity rank in the ordered set } Y_t^{f_i}$.

 $Y_t^{f_i} \equiv \text{period } t \text{ ordered set of the commodities considered for mimicking portfolio construction where the$ ordering rule is specific to f_i .

For our portfolio CHP (commercial hedging pressure), we sort the set of commodities considered on past twenty-six-week average CHP from lowest to highest. The bottom third constituents (lowest \overline{CHP} s) of the corresponding ordered set form the long leg of the portfolio while the top third form the short leg:

$$Y_{t}^{CHP} \equiv \left\{ \left\{ \overline{CHP_{1}}, \overline{CHP_{2}}, ..., \overline{CHP_{n}} \right\}, \leq \right\}$$

$$\overline{CHP_j} = \frac{1}{26} \sum_{k=0}^{25} \frac{L_{j,t-k}}{L_{i,t-k} + S_{j,t-k}}$$

 $\overline{CHP_j} = \tfrac{1}{26} \sum_{k=0}^{25} \tfrac{L_{j,t-k}}{L_{j,t-k} + S_{j,t-k}}$ $L_{j,t} \equiv \text{number of long positions held by commercial hedgers for period } t \text{ on the futures series of commodity}$

 $S_{j,t} \equiv$ number of short positions held by commercial hedgers for period t on the futures series of commodity

For our portfolio TS (term structure), we sort the set of commodities considered on past twenty-six-week average roll-yield from highest to lowest. The bottom third constituents (highest \overline{RY} s) of the corresponding ordered set form the long leg of the portfolio while the top third form the short leg:

$$Y_{t}^{TS}\equiv\left\{ \left\{ \overline{RY_{1}},\overline{RY_{2}},...,\overline{RY_{n}}\right\} ,\geq\right\}$$

$$\overline{RY_j} = \frac{1}{26} \sum_{k=0}^{25} \left(\frac{F_{j,t-k}^1}{F_{j,t-k}^2} - 1 \right)$$

 $\begin{array}{l} \overline{RY_j} = \frac{1}{26} \sum_{k=0}^{25} (\frac{F_{j,t-k}^1}{F_{j,t-k}^2} - 1) \\ F_{j,t}^1 \equiv \text{period } t \text{ close price for commodity } j \text{ first nearby futures contract.} \\ F_{j,t}^2 \equiv \text{period } t \text{ close price for commodity } j \text{ second nearby futures contract.} \end{array}$

For our portfolio OI (open interest), we sort the set of commodities considered on past twenty-six-week average front contract open interest growth from highest to lowest. The bottom third constituents (highest $OI^{\Delta\%}$ s) of the corresponding ordered set form the long leg of the portfolio while the top third form the short leg.

$$Y_t^{OI} \equiv \left\{ \left\{ \overline{OI_1^{\Delta\%}}, \overline{OI_2^{\Delta\%}}, ..., \overline{OI_n^{\Delta\%}} \right\}, \geq \right\}$$

$$\overline{OI_{j}^{\Delta\%}} = \frac{1}{26} \sum_{k=0}^{25} (\frac{OI_{j,t-k}}{OI_{j,t-k-1}} - 1)$$

 $\begin{array}{l} \overline{OI_{j}^{\Delta\%}} = \frac{1}{26} \sum_{k=0}^{25} (\frac{OI_{j,t-k}}{OI_{j,t-k-1}} - 1) \\ OI_{j,t} \equiv \text{open interest at close of period } t \text{ on the front (nearby) contract for commodity } j. \end{array}$

We use a time series regression approach where we start with regressions of the US traded individual commodity weekly front contract returns on returns to the mimicking portfolio for a particular risk factor where the portfolio is constructed using the whole cross-section of US traded commodities. The corresponding R^2 s are averaged and the process is repeated for each of our four risk factors (market, hedging pressure, term structure and liquidity) and periods (1997/2003, 2004/2008, 2008/2013 and 2014/2018). For the long-short factors the results are reported independently for models where the regressor is the factor itself or one of its legs (long vs. short). The same operation is repeated for the two (p=2) and three (p=3) factor model cases, where regressors include in turn all combinations of any two and three risk factors respectively, as well as for the four (p = 4) factor model case where all the risk factor mimicking portfolios are included. Results are reported for the one-factor model case with the other results available upon request.

For each period we also implement the analysis described above over periods of low and high aggregate CHP (CHP), a notion that we believe is particularly relevant in the context of financialisation where issues of co-movement and aggregate market behaviour play a central role. We define CHP as the cross-sectional average of individual CHPs for a particular pool of assets. i.e., period t CHP for asset pool x:

$$\mathbf{CHP}_t^x = \frac{1}{n} \sum_{i=1}^n CHP_{j,t}$$

 $n \equiv \text{number of commodities in set } x.$

 $CHP_{i,t} \equiv \text{period } t \text{ CHP for commodity } i \text{ as defined above.}$

We further define aggregate backwardation (backwardation) and contango (contango) CHP regimes as, for a given period, CHP levels below and above the period's median respectively and, in contrast with (Hong and Yogo, 2012) who aggregate across sub-sectors, the asset pool we consider for CHP construction include the whole cross-section of US traded commodities.

In the next part of the analysis, our mimicking portfolios for risk factors, constructed from the set of all US traded commodities are used as commodity picking devices on the latter commodity set. For each risk factor and period, individual US traded commodity front returns are regressed on returns for the corresponding mimicking portfolio. Individual commodities are sorted by R^2 accordingly from highest to lowest and the top third (highest R^2 s) are selected as commodity picks for the risk factor for the period. i.e, set of picks $P_t^{f_i}$ for risk factor i (f_i) over period t:

$$P_t^{f_i} \equiv \{P_1, P_2, ..., P_x\}$$

 $n \equiv$ number of commodities in the set considered for the picking exercise (twenty-four here).

 $s\equiv$ selection threshold ($\frac{1}{3}$ here). $P_j\equiv$ commodity pick j for risk factor f_i over period t

 $j \equiv {\rm rank}$ in the ordered set $Y_t^{f_i}$

$$Y_t^{f_i} \equiv \left\{ \left\{ \tilde{R_1^2}, \tilde{R_2^2}, ..., \tilde{R_n^2} \right\}, \geq \right\}$$

 $n \equiv$ number of commodities in the set considered for picking (twenty-four here).

 $\tilde{R}_{j}^{2} \equiv \text{risk factor } f_{i} \text{ returns' explanatory power on commodity } j \text{ nearby futures returns over period } t \text{ as}$

For each period we report the proportion of time each pick is held in the long and (/or) short leg of the corresponding risk factor mimicking portfolio as well as the average pairwise correlation amongst every set of picks.

In the last part of the analysis, for each period our mimicking portfolios for risk factors in returns are constructed using each set of commodity picks independently as the asset pool for portfolio construction. A similar time series approach to that described above is then implemented using these newly formed mimicking portfolios. In this case, a first set of dependent variables includes the sets of commodity picks themselves, a second set includes the whole cross-section of US traded commodities while a third set includes the six UK traded metals considered in the study.

By relating returns on attribute based risk factor mimicking portfolios to returns on individual as well as group of assets, our time series approach therefore draws the link between attribute and pricing dynamics over different, theory-based market regimes, and thereby enables us to study market fundamentals and co-movement simultaneously, two key issues in the financialisation phenomenon.

3. Results & discussion

Table 1 displays descriptive statistics for the commodity series considered in the study; the results show a clear pattern of increase in average return during financialisation with an equally weighted portfolio of the US traded assets showing a mean returns of 15.9% significant at the 5% level vs. 6.9% with a 10% significance in the pre-financialisation period. The increase is even more pronounced for an equally weighted portfolio of GB traded assets (metals), 2.47% (not significant) to 21.8% (significant at the 10% level), which sets out the global nature of the phenomenon. Most commodities show a higher mean return while five—Lumber (XCME), Natural gas (XNYM), Palladium (XNYM), Platinum (XNYM) and Nickel-primary (XLME)—show a lower figure with five significant figures—Crude oil-WTI (XNYM), Heating oil (XNYM), Copper (XCEC), Copper (XLME) and Tin-refined (XLME)—against none pre-financialisation. Sugar-#11 wins the race with a mean return increasing from -4.6% to 24.9% followed by copper with sizeable increases for both markets: 1.5% to 28.2% (XCEC) and 0.5% to 28.6% (XLME).

The pattern of results for volatilities is similar although the increase is less pronounced. Both equally weighted portfolios shows an increase, more pronounced for the GB assets, 10% to 13.6% (US) vs. 15.7% to 26.9% (GB), while out of the thirty assets, eleven show a drop in volatility: Oats (XCBT), Lean hogs (XCME), Cocoa (IFUS), Coffee-C (IFUS), Lumber (XCME), Sugar-#11 (IFUS), Natural gas (XNYM), Crude oil-WTI (XNYM), Gasoline (XNYM), Heating oil (XNYM), Palladium (XNYM). Financialisation seems however to have impacted the commodity complex beyond these mechanical effects of increased long participation ensuing from the entry on index type investments as our results suggest next.

The cross-sectional nature of financialisation naturally leads us to extend the individual hedging pressure paradigm by incorporating aggregate CHP (CHP), a market wide measure of hedging pressure. The traditional Keynesian hedging pressure paradigm postulates a negative relationship between the return on an individual commodity futures and its hedging pressure. The extended Keynesian hedging pressure paradigm, in the context of financialisation, should also imply that returns for individual commodities should be high in periods where CHP is low and vice et versa. In Table 2 the mean returns are shown over high (contango) and low (backwardation) CHP regimes for each time period.

Over the first period eighteen of the twenty-four US commodities have higher mean returns during periods of **backwardation** as compared to **contango**, with the difference being statistically significant at the 10% level for two commodities, at the 5% level for one and at the 1% level for another. For the remaining six commodities the higher mean return during **contango** is not statistically significant for any of them. For the six GB commodities, all metals, five of them had higher mean returns during phases of **backwardation** as compared to **contango** in the first period, with one of these differences being statistically significant at the 10% level while for the remaining asset (Zinc (XLME)) the higher mean return during **contango** is not statistically significant. An equally weighted portfolio of the twenty-four US commodities had a mean return of 15.4% during **backwardation** phases, statistically significant at the 1% level, and a mean return of -1.1% during **contango**, both in the first period, with the difference between them statistically significant at the 5% level. The corresponding figures for an equally weighted portfolio of the GB metals were 4.4% and 0.3% respectively. These results provide broad support for an aggregate version of the Keynesian hedging pressure hypothesis and suggest that hedgers in the aggregate were engaged in risk transfer during this period.

Over the second period the pattern essentially reverses. For the twenty-four US commodities over this period, fourteen have higher mean returns during phases of **contango** over **backwardation**. Four of the six GB metals achieved higher mean returns over **contango** relative to **backwardation** phases with the difference being statistically significant for one commodity. The equally weighted US portfolio had a mean return of 15.7% over **backwardation** against 16.2% during phases of **contango**, both statistically significant at the 10% level. The difference was considerably more pronounced for the equally weighted GB metals portfolio with a mean return of 8% during **backwardation** rising to 35.6% during **contango**.

The onset of financialisation thus seems to have engendered a change in the nature of hedger's behaviour, at the aggregate level in particular, taking it away from the traditional Keynesian view of risk shifting.

This phenomenon has been noted in earlier theoretical and empirical work (Danthine, 1978; Stout, 1998) in different contexts from which two competing models of behaviour have emerged.

The information arbitrage model (Grossman and Stiglitz, 1980; Danthine, 1978; Kyle, 1989) implies that hedgers may often be seeking out counterparties to trade with rather than being purely passive. This issue is also raised in both Cheng and Xiong (2014) and Stulz (1996) who point out that hedgers may be taking a view on prices just as speculators do. In fact, by hedging away some of their risk, hedgers are able to speculate more heavily based on their disagreement against speculators regarding futures price movement (Simsek, 2013), which fits into a heterogeneous expectation theory of speculation¹². As Stout (1998) points out, this disagreement-based theory of trading is one of the main reason the public and the law disapproves speculation as this form of trading is regarded as non-productive¹³.

In the second model hedgers are seen as liquidity providers as in Kang et al. (2020) with this explanation also consistent with the nature of financialisation which led to the arrival of long term, long only investors who require substantial liquidity to roll-over their positions. Table 3 supports this view. We regress commodity futures front returns on individual commercial hedging pressure (CHP) as well as aggregate CHP (CHP), our market wide measure of CHP, in separate models and find that the explanatory power of CHP at the individual commodity level decreases over the financialisation period with the US cross-sectional average R^2 falling down to 20.8% from 25.8% in the previous period while that of **CHP** increases (7.1% vs 5.4%). Financialisation therefore seems to have had the effect of loosening the extent to which futures risk premiums relate to commodity specific commercial hedging motives while strengthening that to which they relate to broader dynamics such as cross-sectional liquidity provision. The pattern is strongest for the metals sector where the average R^2 drops down to 19.1% from 28.1% between the two periods at the individual level while it rises from 8.4% to 15.7% at the aggregate level. The symmetry breaks in the crisis period, with the US cross-sectional average R^2 at the individual level bouncing back up (22.5% vs 20.8%) while that at the aggregate level rises further (14.7% vs 7.1%), before eventually resuming post-crisis with a further rise at the individual level (27.2% vs 22.5%) and a drop at the aggregate level (7.5% vs 14.7%) suggesting a return to legacy commodity specific dynamics.

Financialisation was a global phenomenon that, through the channel of index investment, simultaneously influenced the entire spectrum of liquid commodities. This sweeping impact spurred the transformation of commodity markets from a collection of idiosyncratic markets into a unified asset class with interconnectedness emerging as a crucial aspect that warrants thorough examination.

We set out to address this issue by first examining the average of weekly pairwise correlations between commodities across the entire cross-section of the commodity assets considered in the study as well as by country, sector and sub-sector. Table 4 shows the results for the four periods of interest with the results reported for the entire period as well as for periods of **backwardation** and **contango**. For the entire sample the average pairwise correlation jumps from 0.07 in the first period to 0.16 over financialisation. These findings indicate higher interdependence within the full cross-section of commodities starting with the financialisation period, consistent with the empirical findings in Fry-McKibbin and McKinnon (2023) and Mayer et al. (2017) as well as with the theoretical model of Basak and Pavlova (2016). The sector analysis sheds light on the driving force behind these changes. The US metals see the sharpest increase in pairwise correlations going from 0.22 over the past period to 0.56 over financialisation while the three other sectors see a much smaller percentage increase during financialisation relative to the past period. These findings thus suggest that the US metals sector was the driving force behind increased average pairwise correlations during financialisation, a new observation to the best of our knowledge. The pattern of correlations is very similar across phases of backwardation and contango for all the sectors and sub-sectors over the past and financialisation periods which indicates that increased commodity interdependence over financialisation was not a phase driven phenomenon in the same way that individual commodity returns are, as demonstrated in Table 2.

 $^{^{12}\}mathrm{Hirshleifer}$ (1975); Hirshleifer (1976); Hirshleifer (1977)

 $^{^{13}}$ Duffie (2014) also discusses some of the challenges faced by a policy treatment of speculative trading motivated by differences in beliefs.

We further examine these correlations by year in Table 5 to uncover any distinct breakpoints. For the full sample of assets, the average pairwise correlations increase from 0.08 to 0.12 from 2003 to 2004 seemingly driven by US metals average pairwise correlations, more than doubling from 0.22 to 0.49, suggesting 2004 as the onset of financialization as noted in several studies (Baker, 2021; Tang and Xiong, 2012) and our contribution is to observe that the US metals sub-sector seems to have played a major role in this.

We investigate further the issue of co-movement from an asset pricing perspective using a bespoke asset pricing framework starting with a market factor mimicking portfolio constituted of the twenty-four US traded commodities considered in the study with equal weighting. This portfolio enables a refined examination of cross correlations across sectors as the numerator of the beta of the regression of the returns for an individual commodity against those for the market factor is driven by covariances while the denominator is driven by the individual commodity variance. Thus, an increase in a commodity's correlation with other commodities should lead to an increase in its beta relative to the market factor portfolio. We therefore regress the weekly individual commodity front futures returns against those for our market factor portfolio and average the resulting betas across the entire set of commodity assets considered as well as by country, sector and subsector for each of the four periods of interest with the results shown in Table 6. The average beta for the US metals sector jumps sharply to 1.12 over financialisation from 0.63 over the previous period; the increase is even more dramatic for the base metals sub-sector (GB traded assets) in a relative sense. The group had a much lower beta of 0.27 over the past period, suggesting a low level of correlation with US commodities; this figure increases to 0.85 over the financialisation period, even beyond that for the US agricultural sector of 0.80. In contrast, the corresponding figures for the energy sector show a substantial decrease, down to 1.61 during financialisation from 2.33 over the past period.

The corresponding pricing performance displays a similar pattern as shown in Table 7 where for each period the corresponding R^2 values are independently averaged over the entire period as well as over periods of contango and backwardation. The average R^2 for the market factor across the entire cross-section of commodity assets considered in the study, including the GB traded base metals, increases to around 20% during financialisation compared to around 10% over the previous period with the rise driven by the metals sector at large. The sector average for the US traded assets jumps to 31% during financialisation from 7.5% over the past period while for the GB traded assets the corresponding figure increases from less than 2%, or virtually no explanatory power for the (US based) market factor to almost 12% over the financialisation period. The results for our other risk factor mimicking portfolios show a similar pattern and provide a deeper insights into co-movement dynamics in this context. All three factors—CHP, open interest and term structure—performed poorly in the first period, indicating that the commodities with extreme factor attributes (high or low CHP for example) were uncorrelated with the rest of the commodity cross-section, and the overall performance consistent across periods of contango and backwardation. The onset of financialisation leads to a substantial improvement in the performance of the CHP factor with its average R^2 , albeit still small at 5.4%, almost tripling from the past to the financialisation period in contrast to the other factors that show weaker dynamics with a slight decrease and a much smaller increase for the open interest and term structure factors respectively. Here again the dramatic increase in pricing performance for the CHP factor is driven by the metals sector, particularly the precious sub-sector with the average R^2 for the group jumping to almost 26% during financialisation from 1.2% over the past period. The corresponding figures of 22% and 1.1% for the entire US metals sector stand in sharp contrast with the agricultural sector, which exhibits no notable change, and the energy sector, where pricing performance drops substantially from 9.2% down to 1.6%. The improvement in the CHP factor's pricing performance remarkably comes predominantly from the long leg (low CHP assets: Keynesian "normal backwardation") of the factor \{Results available from the authors upon request.\} with the corresponding mimicking portfolio's performance itself driven by the metals, the precious sub-sector in particular. The average R^2 for the group jumps to 55% during financialisation from less than 12% over the past period while the short leg sees a far smaller increase, up to 8.9% from 3.3% over the past period. Interestingly, although the factor itself only sees a very modest increase, its short leg (low roll-yield assets: Working's backwardation) shows a sizeable rise in pricing performance, yet again driven the metals sector, with the average R^2 for the group's US traded assets jumping to 28.2% during financialisation from 6.2% over the past period. The pattern of results for both factors extends to the GB trade assets and suggest a convergence of Keneysian and Working's term structure perspective over the financialisation period.

Taken together these results emphasise the global nature of the impact of financialisation on the metals complex and provide further evidence that global metals co-movement could be detected by US based factor models that combine the Keynesian and Working's term structure paradigms. There seems thus to be information in the factor constructed from hedging pressure based on the Keynesian paradigm relevant to the term structure factor built on the Working paradigm and these findings therefore appear to complement the results of Kang et al. (2020) who find that hedging pressure both conveys information about liquidity provision in the short term (Working paradigm) and risk transfer in the longer term (Keynesian paradigm).

The crisis period (late 2008 to mid-2013) shows a contrasting pattern of results. Returns fall sharply relative to the financialisation period (Table 1) with eighteen US commodities along with all six GB traded metals showing lower mean returns. Returns for equally weighted portfolios of both sets of assets fall down to 9.4% and 5.3% from 15.9% and 21.8% respectively. Returns are mostly higher during phases of **backwardation** relative to **contango** (Table 2) with fifteen US assets and five of the GB traded metals showing higher mean returns in **backwardation** and the pattern extending to equally weighted portfolios of the two sets of assets. Volatility is moderately higher for both sets with fifteen US commodities and three GB metals showing higher figures and interestingly eminently higher over phases of **contango**; out of the US complex only Sugar-#11 (IFUS) and none of the GB traded metals show higher volatility figures over **backwardation** phases. The symmetry between the dynamics of CHP and **CHP**'s explanatory power on individual commodity returns breaks in the crisis period (Table 3) with both now seemingly converging; the US cross-sectional average R^2 at the individual level bouncing back up (22.5% vs 20.8%) while that at the aggregate level rises further (14.7% vs 7.1%).

// TODO Add correlation results discussion here.

The performance of the market factor dominates the US commodity complex over the period with a 50% increase in mean average R^2 to over 30% (Table 7) while that of the CHP factor decreases and, although increasing, remains modest for the open interest growth and term structure factors. Taken together, these results indicate the presence of a systematic factor across the entire cross-section of the US commodities.

The returns pull-back continues into the post-crisis period (Table 1) with twenty-two US and four UK traded assets showing lower mean returns relative to the crisis while the same number of US commodities and all UK metals show lower figures relative to the financialisation period; mean returns for equally weighted portfolios of US and UK assets drop down to -0.2% and 2.7% respectively. Interestingly volatility also recedes substantially with twenty US assets and all UK metals showing lower volatility figures relative to both the crisis and financialisation periods suggesting that investment outflows from the commodity futures complex have had a lower impact on volatility than the corresponding inflows witnessed over the financialisation period. The figures for the equally weighted portfolios of US and UK assets drop to 9.9% and 16.2% respectively, very close to pre-financialisation levels.

The pattern of reversion toward the pre-financialisation Keynesian paradigm initiated during the crisis continues stronger into the post-crisis period (Table 2) with mean returns higher during phases of **backwardation** relative to **contango**, more so than over the crisis period with seventeen US commodities showing higher mean returns over **backwardation** phases. The difference is significant at the 5% level for Lumber (XCME) and at the 10% level for Palladium (XNYM) while no difference is significant for the assets showing higher figures over phases of **contango**. The pattern is more pronounced for the UK traded subset with all six assets showing higher mean returns in **backwardation** with the difference significant at the 1% level for Nickel-primary (XLME), 5% level for Aluminium (XLME) and 10% level for Zinc (XLME). It also extends to equally weighted portfolios of US and UK traded assets that both show higher mean returns during **backwardation** with the difference significant at the 5% level for the UK portfolio and the figure close to the 10% level significance threshold for the portfolio of US assets. Volatilities are overall higher in

contango although less so than over the previous period with seventeen US commodities showing higher figures over contango against twenty-three in the crisis period. The pattern extends to three of the six UK traded metals with the difference significant at the 1% level for Copper (XLME) and Tin (XLME) and at the 5% level for Zinc (XLME) while differences are not significant for Aluminium (XLME), Lead (XLME) and Nickel-primary (XLME) that exhibit higher volatility over **backwardation** phases. Equally weighted portfolios of US and UK traded assets on the other hand both show higher volatility in contango with the difference significant at the 1% level for both. Over the financialisation period in contrast volatilities were predominantly higher in backwardation for US traded assets with fourteen commodities showing higher volatility figures in these phases; the post-crisis pattern of results for volatilities thus likewise shows signs of reversion to pre-financialisation standards. After showing signs of convergence in the crisis period the dynamics of the explanatory power of CHP and CHP part ways (Table 3), seemingly reverting to the Keynesian paradigm observed pre-financialisation, further suggesting a return to legacy commodity specific dynamics. The recovery initiated in the crisis period for that of CHP is confirmed with a stronger increase in the post-crisis period, 27.2% vs. 22.5%, strongest for the metals sector, 37.3% vs. 23.7%, while the energy sector shows the opposite pattern: 11% vs. 16%. In contrast, after increasing stronger in the crisis, that for CHP falls overall in the post-crisis period, 7.5% vs 14.7%, with the decrease strongest for the energy sector in a relative sense: 2.5% vs. 14.8%.

Taken together the patterns of results over the crisis and post-crisis periods echo those in Basu and Bauthéac (2024) in suggesting that the exogenous shock delivered by the financial crisis combined with the accommodative monetary policy that ensued seem to have initiated a reversion toward pre-financialisation fundamentals in the commodity futures complex.

4. Conclusion

In the early 2000s, against a backdrop of a low yield environment and poor stock market performance, a combination of financial innovations¹⁴ and regulatory changes¹⁵ led to large inflows of institutional capital into the commodity futures markets. This process known as "financialisation" spurred a heated public policy debate about whether the ensuing increase in open interest and trading volume in commodity futures exerted upward pressure on prices.

The debate spread to the legislative sphere, with the US senate launching formal investigations, and was thrust onto the academic community as a matter of urgency. Perhaps in response to this, most of the early studies focused on the more mechanical effects of financialisation in individual markets, relying on commodity specific causality and correlation-based analysis. The contention eventually triggered legislative action and new position limits were introduced in a number of grains and energy futures markets.

With the benefit of hindsight, this focus appears to have been too narrow. As our results suggests, financialisation was a phenomenon global in nature and studying the direct effects of increased market participation in single markets thus seems amiss in this context.

The results of our analysis further suggest that financialisation had cross-sectional effects on market fundamentals across the commodity complex beyond energy and agricultural commodities and in fact strongest for the metals sector at the global level; an aspect largely overlooked in the original approach that we believe the existing literature has not fully analysed.

The onset of the financial crisis and the monetary policy regimes that followed, on the other hand, also appear to have induced significant changes in commodity pricing dynamics. In contrast with financialisation, our results suggest that the crisis and its aftermath have delivered an exogenous shock across the entire cross-section of liquid commodities and seem to have triggered a motion of reversion to legacy pre-financialisation fundamentals.

¹⁴Commodity price indexes and the ad-hoc financial instruments enabling investment in them were the main financial innovations which allowed large global banks to offer commodity investment products to institutional and retail investors. In 1991 Goldman Sachs created the S&P-GSCI which provides investors with buy-side exposure to commodities via the OTC swap market and thus without having to participate in the formal futures markets with their position limit restrictions. At this stage, these restrictions still applied to the issuing institutions though, as they hedged the corresponding commodity swap exposure in the futures markets. The first commodity-based ETFs were created through buying physical precious metals with gold and silver ETFs offered as early as 2002/2003. The regulatory hurdle here related to the licensing of commodity trading professionals. Typically, investors had to sign a statement with their broker stating that they understood the risks of commodity investments; a rather inconvenient paperwork for a product designed to trade like a stock, as set forth by a number of industry players at the time.

¹⁵In 2000, the Commodity Futures Modernization Act (CFMA) granted non-agricultural commodity futures statutory exemption from regulation ("Enron loophole"). It still required agricultural commodity derivatives be traded on a CFTC-regulated exchange however. Eventually, the CFTC classified swap dealers as "bona fide" hedgers, granting them position limits exemption ("swap dealer loophole"). In 2005, the CFTC waived the rule that required commodity investors to sign a statement saying they understood the risks, letting the funds replace it with their prospectus. The regulatory bottleneck that prevented the large-scale expansion of commodity index investment was no more.

5. Tables

Table 1: This table shows mean returns and volatility (sd) for the twenty four individual US commodities and the six GB metals considered in the study as well as for two equally weighted portfolios formed from the US commodities and the GB metals respectively across the four periods of interest (past: 1997-2003; financialisation: 2004-2008; crisis: 2008-2013; post-crisis: 2013-2018). Mean values significant at the 1%, 5% and 10% level are marked with ***, ** and * respectively. The results are discussed in section Section 3.

mean sd	3.48%	21.37%	6.89%	
			0.00/0	-4.51%
	23%	29.54 %	36.15%	21.8%
mean	5.49%	21.62 %	8.89%	-0.87%
sd	$\boldsymbol{32.22\%}$	$\boldsymbol{31.04\%}$	$\boldsymbol{34.67\%}$	$\boldsymbol{29.5\%}$
mean	7.29%	12.06%	8.12%	-0.33%
sd	$\boldsymbol{26.34\%}$	$\boldsymbol{32.55\%}$	31.78%	$\boldsymbol{25.11\%}$
mean	6.21%	$\boldsymbol{15.57\%}$	$\boldsymbol{4.04\%}$	-7.39%
sd	$\boldsymbol{21.43\%}$	$\boldsymbol{28.35\%}$	$\boldsymbol{25.75\%}$	18.95%
mean	$\boldsymbol{6.89\%}$	13.36%	6.19%	-3.82%
$\operatorname{\mathbf{sd}}$	21.47%	30.04%	28%	21.22 %
mean	5.07%	17.63%	6.68%	-1.58%
sd	24.93 %	33.03%	$\boldsymbol{37.62\%}$	$\boldsymbol{26.58\%}$
mean	0.39%	7.43%	7.13%	1.87%
sd	$\boldsymbol{13.23\%}$	15.16%	$\boldsymbol{15.24\%}$	18.69%
mean	3.45%	8.77%	4.13%	1.92%
sd	17.1%	17.43%	16.32%	18.3%
mean	0.73%	8.4%	$\boldsymbol{12.69\%}$	-1.09%
sd	38.33%	31.89%	30.37%	36.94%
mean	3.66%	16.35%	1.39%	3.65%
sd	$\boldsymbol{32.66\%}$	29.99 %	31.03%	24.69 %
mean	-5.91%	20.38%	2.2%	1.99%
sd	43.84%	32.3%	30.17%	31.86%
mean	3.24%	0.95%	12.53%	0.5%
sd	$\boldsymbol{26.38\%}$	$\boldsymbol{29.01\%}$	33.26 %	$\boldsymbol{20.61\%}$
mean	1.9%	-3.58%	11.74%	5.06%
sd	30.58%	29.39 %	$\pmb{36.31\%}$	25.1%
mean	$\boldsymbol{1.67\%}$	$\boldsymbol{13.87\%}$	14.4%	3.8%
sd	$\boldsymbol{29.5\%}$	$\boldsymbol{33.01\%}$	$\boldsymbol{34.43\%}$	$\boldsymbol{29.37\%}$
mean	-4.56%	24.87 %	11.82%	-1.16%
sd	34.88%	33.25%	37.9%	28.72%
mean	35.38%	16.25%	0.93%	8.33%
sd	61.55%	54.6%	53.58%	44.58%
mean	15.43%	*28.74%	9.67%	-5.64%
sd	$\boldsymbol{39.87\%}$	$\boldsymbol{32.56\%}$	$\boldsymbol{42.84\%}$	$\boldsymbol{34.69\%}$
mean	16.65%	30.95%	10.59%	-5.37%
$\operatorname{\mathbf{sd}}$	42.77%	41.76%	40.07%	37.09%
mean	$\boldsymbol{16.31\%}$	*29.88%	6.83%	-3.44%
sd	$\boldsymbol{40.75\%}$	35.09%	$\boldsymbol{33.17\%}$	30.56%
mean	$\boldsymbol{1.45\%}$	*28.15%	6.19%	-0.1%
sd	$\boldsymbol{21.23\%}$	$\boldsymbol{31.89\%}$	34.91 %	19.56%
mean	4.5%	14.35%	14.01%	0.53%
sd	15.14%	18.89%	$\boldsymbol{21.46\%}$	14.41%
mean	9.12%	9.03%	*29.12%	13.75%
$\operatorname{\mathbf{sd}}$	38.27%	32.71%	$\boldsymbol{36.14\%}$	$\boldsymbol{25.38\%}$
mean	12.93%	10.56%	7.29%	-7.92%
sd	$\boldsymbol{22.62\%}$	22.66%	25.53%	19.48%
mean	6.19%	17.54%	$\boldsymbol{21.96\%}$	-2.43%
sd	22%	33.92%	39.87%	$\boldsymbol{24.35\%}$
mean	1.13%	13.78%	-3.58%	3.15%
sd	17.6%	$\boldsymbol{26.41\%}$	$\boldsymbol{26.48\%}$	18.56%
mean	0.45%	**28.59%	5.74%	0.03%
sd	19.25%	31.09%	34.27%	18.92%
	mean sd mean	sd 21.43% mean 6.89% sd 21.47% mean 5.07% sd 24.93% mean 0.39% sd 13.23% mean 3.45% sd 17.1% mean 0.73% sd 38.33% mean 3.66% sd 32.66% mean -5.91% sd 43.84% mean 3.24% sd 26.38% mean 1.9% sd 30.58% mean 1.67% sd 29.5% mean -4.56% sd 34.88% mean 15.43% sd 39.87% mean 16.65% sd 42.77% mean 16.31% sd 40.75% mean 1.45% sd 21.23% mean 4.5%	sd 21.43% 28.35% mean 6.89% 13.36% sd 21.47% 30.04% mean 5.07% 17.63% sd 24.93% 33.03% mean 0.39% 7.43% sd 13.23% 15.16% mean 3.45% 8.77% sd 17.1% 17.43% mean 0.73% 8.4% sd 38.33% 31.89% mean 3.66% 16.35% sd 32.66% 29.99% mean -5.91% 20.38% sd 43.84% 32.3% mean -5.91% 20.38% sd 43.84% 32.3% mean -5.91% 20.38% sd 26.38% 29.01% mean 1.9% -3.58% sd 26.38% 29.01% mean 1.67% 13.87% sd 29.5% 33.01% mean -4.56% 24.87% sd 34.88% 33.25%	sd 21.43% 28.35% 25.75% mean 6.89% 13.36% 6.19% sd 21.47% 30.04% 28% mean 5.07% 17.63% 6.68% sd 24.93% 33.03% 37.62% mean 0.39% 7.43% 7.13% sd 13.23% 15.16% 15.24% mean 3.45% 8.77% 4.13% sd 17.1% 17.43% 16.32% mean 0.73% 8.4% 12.69% sd 38.33% 31.89% 30.37% mean 3.66% 16.35% 1.39% sd 32.66% 29.99% 31.03% mean -5.91% 20.38% 2.2% sd 32.48% 32.3% 30.17% mean 3.24% 0.95% 12.53% sd 26.38% 29.01% 33.26% mean 1.67% 13.87% 14.4% sd 29.5%

Lead (XLME)	mean	3.83%	28.99%	10.74%	2.07%
Lead (XLME)	sd	19.64%	41.74%	$\boldsymbol{41.05\%}$	23.26 %
Nickel-primary (XLME)	mean	19.31%	12.5%	3.52%	0.31%
Nickel-primary (XLME)	sd	32.26%	$\boldsymbol{44.27\%}$	41.45%	$\boldsymbol{28.98\%}$
Tin-refined (XLME)	mean	4.07%	*29.1%	7.46%	$\boldsymbol{1.68\%}$
Tin-refined (XLME)	sd	15.38%	31.71%	$\boldsymbol{35.61\%}$	$\boldsymbol{18.79\%}$
Zinc (XLME)	mean	-4.52%	$\boldsymbol{20.13\%}$	7.62%	9.07%
Zinc (XLME)	sd	$\boldsymbol{21.73\%}$	38.77%	36.17%	23.25 %
US commodities	mean	*6.89%	**15.89%	9.39%	-0.18%
US commodities	sd	10%	$\boldsymbol{13.62\%}$	17.58%	9.86%
GB commodities	mean	2.47%	*21.77%	5.25%	2.72%
GB commodities	sd	15.66%	$\boldsymbol{26.91\%}$	$\boldsymbol{30.24\%}$	$\boldsymbol{16.19\%}$

Table 2: This table shows mean returns and volatility (sd) for the twenty four individual US commodities and the six GB metals considered in the study as well as for two equally weighted portfolios formed from the US commodities and the GB metals respectively across the four periods of interest (past: 1997-2003; financialisation: 2004-2008; crisis: 2008-2013; post-crisis: 2013-2018). Figures are shown independently for phases of aggregate backwardation (aggregate CHP \leq period median) and aggregate contango (aggregate CHP > period median). Mean values significant at the 1%, 5% and 10% level are marked with ***, ** and * respectively. Aggregate CHP construction and corresponding regime definitions are discussed in section Section 2.3 while the results are discussed in section Section 3.

asset	regime	estimate	past	financialisation	crisis	post-crisis
Corn-#2 yellow (XCBT)	backwardation	mean	-4.86%	19.26%	20.48%	-8.74%
Corn-#2 yellow (XCBT)	backwardation	sd	23.07%	29.14%	31.59%	21%
Corn-#2 yellow (XCBT)	contango	mean	11.3%	23.54 %	-4.43%	0.78%
Corn-#2 yellow (XCBT)	contango	sd	23%	30.01%	40.14%	$\boldsymbol{22.68\%}$
Oats (XCBT)	backwardation	mean	-5.39%	7.82%	7.85%	$\boldsymbol{2.56\%}$
Oats (XCBT)	backwardation	sd	29.02%	33.82%	31.43%	31.55%
Oats (XCBT)	contango	mean	16.95%	*35.38%	11.68%	-3.17%
Oats (XCBT)	contango	sd	35.22%	28.1%	37.7%	27.5%
Soybean meal (XCBT)	backwardation	mean	13.15%	8.22%	16.04%	-3.52%
Soybean meal (XCBT)	backwardation	sd	27.77%	33.06%	25.7%	25.1%
Soybean meal (XCBT)	contango	mean	1.52%	15.91%	2.32%	2.22%
Soybean meal (XCBT)	contango	sd	24.94%	32.13%	36.85%	25.32%
` ,	backwardation			14.62%	20.34%	
Soybean oil (XCBT)		mean	14.79%			-9.44%
Soybean oil (XCBT)	backwardation	sd	21.35%	28.94%	22.33%	18.19%
Soybean oil (XCBT)	contango	mean	-2.68%	16.62%	-10.01%	-4.12%
Soybean oil (XCBT)	contango	$\operatorname{\mathbf{sd}}$	$\boldsymbol{21.55\%}$	$\boldsymbol{27.91\%}$	$\boldsymbol{28.65\%}$	19.79%
Soybeans (XCBT)	backwardation	mean	13.69%	0.93%	$\boldsymbol{13.42\%}$	-11.45%
Soybeans (XCBT)	backwardation	$\operatorname{\mathbf{sd}}$	21.93 %	34.58 %	24.39 %	22.04 %
Soybeans (XCBT)	contango	mean	-0.16%	25.71%	0.96%	3.75%
Soybeans (XCBT)	contango	$\operatorname{\mathbf{sd}}$	$\boldsymbol{21.09\%}$	24.74 %	31.11%	$\boldsymbol{20.56\%}$
Wheat-SRW (XCBT)	backwardation	mean	-2.7%	19.02%	0.98%	0.44%
Wheat-SRW (XCBT)	backwardation	sd	25.79%	36%	37.53%	26.57%
Wheat-SRW (XCBT)	contango	mean	11.78%	16.29%	14.57%	-2.77%
Wheat-SRW (XCBT)	contango	sd	24.04%	29.84%	37.73%	26.72%
Cattle-feeder (XCME)	backwardation	mean	2.31%	**20.52%	7.51%	6.58%
Cattle-feeder (XCME)	backwardation	sd	12.79%	14.93%	13.39%	20.14%
Cattle-feeder (XCME)	contango	mean	-0.81%	-5.49%	7.28%	-3.12%
,	contango	sd	13.67%	15.35%	16.91%	17.23%
Cattle-feeder (XCME) Cattle-live (XCME)	backwardation	mean	8.39%	14.2%	5.87%	-2.07%
Cattle-live (XCME)	backwardation	mean sd	16.65%	17.91%	14.97%	19.9%
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Cattle-live (XCME)	contango	mean	-0.34%	3.43%	2.7%	5.82%
Cattle-live (XCME)	contango	sd	17.49%	16.99%	17.57%	16.66%
Lean hogs (XCME)	backwardation	mean	12.34%	16.3%	18.23%	19.24%
Lean hogs (XCME)	backwardation	sd	37.6%	35.73%	27.83%	35.61%
Lean hogs (XCME)	contango	mean	-10.5%	0.6%	7.5%	-20.66%
Lean hogs (XCME)	contango	\mathbf{sd}	39.22 %	$\boldsymbol{27.67\%}$	$\boldsymbol{32.78\%}$	38.28 %
Cocoa (IFUS)	backwardation	mean	4.12%	9.25%	-9.56%	-2.98%
Cocoa (IFUS)	backwardation	sd	$\boldsymbol{33.57\%}$	31.05%	$\boldsymbol{26.93\%}$	25.86 %
Cocoa (IFUS)	contango	mean	3.67%	23.43 %	$\boldsymbol{12.92\%}$	10.29%
Cocoa (IFUS)	contango	sd	$\boldsymbol{31.81\%}$	$\boldsymbol{28.91\%}$	34.66 %	23.57 %
Coffee-C (IFUS)	backwardation	mean	4.51%	26.49%	2.66%	14.36%
Coffee-C (IFUS)	backwardation	sd	48.82%	32.6%	29.92%	31.52%
Coffee-C (IFUS)	contango	mean	-16.87%	14.38%	3.4%	-10.24%
Coffee-C (IFUS)	contango	sd	38.35%	32.06%	30.22%	32.24%
Cotton-#2 (IFUS)	backwardation	mean	$\boldsymbol{22.64\%}$	7.85%	35.53%	10.84%
Cotton-#2 (IFUS)	backwardation	sd	26.02%	29.71%	33.94%	21.33%
Cotton-#2 (IFUS)	contango	mean	-16.46%	-5.92%	-9.36%	-8.91%
Cotton-#2 (IFUS)	contango	sd	26.77%	28.35%	$\frac{-9.30\%}{32.59\%}$	19.95%
Lumber (XCME)	backwardation	mean	-8.48%	-2.37%	29.11%	**26.99%
Lumber (XCME)	backwardation	sd	$\frac{-6.48\%}{30.43\%}$	30.72%	$\frac{29.11\%}{35.45\%}$	21.26%
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Lumber (XCME)	contango	mean	11.68%	-4.77%	-5.04%	-17.14%
Lumber (XCME)	contango	sd	30.56%	28.2%	37.23%	28.45%
Orange juice (IFUS)	backwardation	mean	-2.38%	-2.22%	$\boldsymbol{18.24\%}$	18.64%

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Orange juice (IFUS)	backwardation	\mathbf{sd}	28.01%	33.01%	30.09%	28.64%
Orange juice (IFUS)	contango	mean	5.83%	29.93%	12.29%	-8.64%
Orange juice (IFUS)	contango	\mathbf{sd}	31.05%	33.1%	38.06%	30.11%
Sugar-#11 (IFUS)	backwardation	mean	11.98%	29.51%	2.7%	-8.07%
Sugar-#11 (IFUS)	backwardation	sd	33.23%	33.6%	41.29%	27.69%
Sugar-#11 (IFUS)	contango	mean	-20.49%	20.35%	21.47%	4.7%
Sugar-#11 (IFUS)	contango	sd	36.48%	32.98%	34.09%	29.83%
Sugar #11 (11 CS)	o .	БС		02.0070	01.0070	20.0070
Natural gas (XNYM)	backwardation	mean	**85.84%	$\boldsymbol{4.76\%}$	-2.32%	18.33%
Natural gas (XNYM)	backwardation	\mathbf{sd}	63.46 %	47.3%	48.85%	42.59 %
Natural gas (XNYM)	contango	mean	-11.98%	$\boldsymbol{27.84\%}$	4.77%	3.25%
Natural gas (XNYM)	contango	\mathbf{sd}	$\boldsymbol{59.76\%}$	$\boldsymbol{61.3\%}$	$\mathbf{58.18\%}$	$\boldsymbol{46.41\%}$
Crude oil-WTI (XNYM)	backwardation	mean	$\boldsymbol{29.17\%}$	*35.61%	11.47%	14.24 %
Condo all WTI (VNVM)	backwardation	sd	36.74%	32.09%	28.19%	26.94%
Crude oil-WTI (XNYM)			2.21%	22.04%	9.95%	-24.85%
Crude oil-WTI (XNYM)	contango	mean				
Crude oil-WTI (XNYM)	contango	sd	42.74%	33.1%	53.7%	41.08%
Gasoline (XNYM)	backwardation	mean	39%	19.82%	23.98%	9.69%
Gasoline (XNYM)	backwardation	sd	42.47%	39.38%	31.41%	32.1%
Gasoline (XNYM)	contango	mean	-5.22%	42.13%	-0.09%	-19.47%
Gasoline (XNYM)	contango	\mathbf{sd}	43.14%	44.13%	47.11%	41.71%
Heating oil (XNYM)	backwardation	mean	36.32%	29.41%	25.56%	15.16%
Heating oil (XNYM)	backwardation	sd	41.37%	33.33%	25.95%	24.62%
Heating oil (XNYM)	contango	mean	-2.74%	30.5%	-10.77%	-21.62%
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Heating oil (XNYM)	contango	$\operatorname{\mathbf{sd}}$	40.19%	36.98%	39.16%	35.69%
Copper (XCEC)	backwardation	mean	3.14%	22.17%	16.53%	12.1%
Copper (XCEC)	backwardation	sd	20.71%	30.05%	25.73%	18.16%
Copper (XCEC)	contango	mean	-0.28%	34.17%	-3.46%	-12.15%
Copper (XCEC)	contango	\mathbf{sd}	$\mathbf{21.78\%}$	33.69%	$\boldsymbol{42.28\%}$	$\boldsymbol{20.87\%}$
Gold (XCEC)	backwardation	mean	9.19%	13.37%	9.37%	4.21%
Gold (XCEC)	backwardation	sd	16.89%	19.2%	17.03%	12.36%
Gold (XCEC)	contango	mean	-0.01%	15.37%	19.1%	-3.1%
Gold (XCEC)	contango	sd	13.24%	18.58%	25.21%	16.23%
Palladium (XNYM)	backwardation	mean	**49.79%	24.47%	30.01%	**31.93%
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Palladium (XNYM)	backwardation	\mathbf{sd}	34.98%	34.23%	33.49%	23.82%
Palladium (XNYM)	contango	mean	-30.19%	-6.3%	30.58%	-4.88%
Palladium (XNYM)	contango	sd	41.22%	31.2%	38.71%	26.8%
	backwardation	mean	**29.46%		10.46%	1.17%
Platinum (XNYM)				*25.07%		
Platinum (XNYM) Platinum (XNYM)	backwardation	sd	22.1%	22.51%	21.24%	18.04%
Platinum (XNYM)		sd mean	22.1% -3.22%			
Platinum (XNYM) Platinum (XNYM)	contango			$\boldsymbol{22.51\%}$	$\boldsymbol{21.24\%}$	18.04%
Platinum (XNYM) Platinum (XNYM) Platinum (XNYM)		mean sd	-3.22% 23.12%	22.51% -3.81% 22.81%	21.24% 8.04% 28.75%	18.04% -17.28% 20.8%
Platinum (XNYM) Platinum (XNYM) Platinum (XNYM) Silver (XCEC)	contango contango backwardation	mean	-3.22% 23.12% 5.98%	22.51% -3.81% 22.81% 14.29%	21.24% 8.04% 28.75% 19.66%	18.04% -17.28% 20.8% 3.44%
Platinum (XNYM) Platinum (XNYM) Platinum (XNYM) Silver (XCEC) Silver (XCEC)	contango contango backwardation backwardation	mean sd mean sd	-3.22% 23.12% 5.98% 21.07%	22.51% -3.81% 22.81% 14.29% 32.5%	21.24% 8.04% 28.75% 19.66% 37.17%	18.04% $-17.28%$ $20.8%$ $3.44%$ $20.3%$
Platinum (XNYM) Platinum (XNYM) Platinum (XNYM) Silver (XCEC) Silver (XCEC) Silver (XCEC)	contango contango backwardation backwardation contango	mean sd mean sd mean	-3.22% 23.12% 5.98% 21.07% 6.57%	22.51% -3.81% 22.81% 14.29% 32.5% 20.83%	21.24% 8.04% 28.75% 19.66% 37.17% 26.77%	18.04% -17.28% 20.8% 3.44% 20.3% -8.09%
Platinum (XNYM) Platinum (XNYM) Platinum (XNYM) Silver (XCEC) Silver (XCEC) Silver (XCEC)	contango contango backwardation backwardation contango contango	mean sd mean sd mean sd	-3.22% 23.12% 5.98% 21.07% 6.57%	22.51% -3.81% 22.81% 14.29% 32.5% 20.83% 35.34%	21.24% 8.04% 28.75% 19.66% 37.17% 26.77% 42.44%	18.04% -17.28% 20.8% 3.44% 20.3% -8.09% 27.77%
Platinum (XNYM) Platinum (XNYM) Platinum (XNYM) Silver (XCEC) Silver (XCEC) Silver (XCEC) Silver (XCEC) Aluminium (XLME)	contango contango backwardation backwardation contango	mean sd mean sd mean	-3.22% 23.12% 5.98% 21.07% 6.57%	22.51% -3.81% 22.81% 14.29% 32.5% 20.83%	21.24% 8.04% 28.75% 19.66% 37.17% 26.77%	18.04% -17.28% 20.8% 3.44% 20.3% -8.09%
Platinum (XNYM) Platinum (XNYM) Platinum (XNYM) Silver (XCEC) Silver (XCEC) Silver (XCEC) Silver (XCEC) Aluminium (XLME) Aluminium (XLME)	contango contango backwardation backwardation contango contango backwardation backwardation	mean sd mean sd mean sd mean sd	-3.22% 23.12% 5.98% 21.07% 6.57% 22.93% 2.56% 14.82%	22.51% -3.81% 22.81% 14.29% 32.5% 20.83% 35.34% 14.78% 25.57%	21.24% 8.04% 28.75% 19.66% 37.17% 26.77% 42.44% 16.25% 24.94%	18.04% -17.28% 20.8% 3.44% 20.3% -8.09% 27.77% **23.68% 18.94%
Platinum (XNYM) Platinum (XNYM) Platinum (XNYM) Silver (XCEC) Silver (XCEC) Silver (XCEC) Silver (XCEC) Aluminium (XLME) Aluminium (XLME) Aluminium (XLME)	contango contango backwardation contango contango backwardation backwardation contango	mean sd mean sd mean sd mean sd mean	-3.22% 23.12% 5.98% 21.07% 6.57% 22.93% 2.56% 14.82% -0.38%	22.51% -3.81% 22.81% 14.29% 32.5% 20.83% 35.34% 14.78% 25.57% 12.86%	21.24% 8.04% 28.75% 19.66% 37.17% 26.77% 42.44% 16.25% 24.94% -23.04%	18.04% -17.28% 20.8% 3.44% 20.3% -8.09% 27.77% **23.68% 18.94% -16.74%
Platinum (XNYM) Platinum (XNYM) Platinum (XNYM) Silver (XCEC) Silver (XCEC) Silver (XCEC) Silver (XCEC) Aluminium (XLME) Aluminium (XLME)	contango contango backwardation backwardation contango contango backwardation backwardation	mean sd mean sd mean sd mean sd	-3.22% 23.12% 5.98% 21.07% 6.57% 22.93% 2.56% 14.82%	22.51% -3.81% 22.81% 14.29% 32.5% 20.83% 35.34% 14.78% 25.57%	21.24% 8.04% 28.75% 19.66% 37.17% 26.77% 42.44% 16.25% 24.94%	18.04% -17.28% 20.8% 3.44% 20.3% -8.09% 27.77% **23.68% 18.94%
Platinum (XNYM) Platinum (XNYM) Platinum (XNYM) Silver (XCEC) Silver (XCEC) Silver (XCEC) Aluminium (XLME) Aluminium (XLME) Aluminium (XLME) Aluminium (XLME)	contango contango backwardation backwardation contango contango backwardation backwardation contango contango	mean sd mean sd mean sd mean sd mean sd	-3.22% 23.12% 5.98% 21.07% 6.57% 22.93% 2.56% 14.82% -0.38% 20.13%	22.51% -3.81% 22.81% 14.29% 32.5% 20.83% 35.34% 14.78% 25.57% 12.86% 27.29%	21.24% 8.04% 28.75% 19.66% 37.17% 26.77% 42.44% 16.25% 24.94% -23.04% 28.01%	18.04% -17.28% 20.8% 3.44% 20.3% -8.09% 27.77% **23.68% 18.94% -16.74% 18.23%
Platinum (XNYM) Platinum (XNYM) Platinum (XNYM) Silver (XCEC) Silver (XCEC) Silver (XCEC) Silver (XCEC) Aluminium (XLME) Aluminium (XLME) Aluminium (XLME) Aluminium (XLME) Copper (XLME)	contango contango backwardation backwardation contango contango backwardation contango contango contango contango	mean sd mean sd mean sd mean sd mean	-3.22% 23.12% 5.98% 21.07% 6.57% 22.93% 2.56% 14.82% -0.38% 20.13%	22.51% -3.81% 22.81% 14.29% 32.5% 20.83% 35.34% 14.78% 25.57% 12.86% 27.29%	21.24% 8.04% 28.75% 19.66% 37.17% 26.77% 42.44% 16.25% 24.94% -23.04% 28.01% 18.66%	18.04% -17.28% 20.8% 3.44% 20.3% -8.09% 27.77% **23.68% 18.94% -16.74% 18.23% 11.99%
Platinum (XNYM) Platinum (XNYM) Platinum (XNYM) Silver (XCEC) Silver (XCEC) Silver (XCEC) Silver (XCEC) Aluminium (XLME) Aluminium (XLME) Aluminium (XLME) Aluminium (XLME) Copper (XLME) Copper (XLME)	contango contango backwardation backwardation contango contango backwardation contango contango contango contango	mean sd mean sd mean sd mean sd mean sd mean sd	-3.22% 23.12% 5.98% 21.07% 6.57% 22.93% 2.56% 14.82% -0.38% 20.13% 1.54% 17.98%	22.51% -3.81% 22.81% 14.29% 32.5% 20.83% 35.34% 14.78% 25.57% 12.86% 27.29% 25.22% 29.01%	21.24% 8.04% 28.75% 19.66% 37.17% 26.77% 42.44% 16.25% 24.94% -23.04% 28.01%	18.04% -17.28% 20.8% 3.44% 20.3% -8.09% 27.77% **23.68% 18.94% -16.74% 18.23% 11.99% 17.92%
Platinum (XNYM) Platinum (XNYM) Platinum (XNYM) Silver (XCEC) Silver (XCEC) Silver (XCEC) Silver (XCEC) Aluminium (XLME) Aluminium (XLME) Aluminium (XLME) Copper (XLME) Copper (XLME) Copper (XLME)	contango contango backwardation contango contango backwardation backwardation contango contango backwardation contango contango backwardation backwardation contango	mean sd mean sd mean sd mean sd mean sd mean	-3.22% 23.12% 5.98% 21.07% 6.57% 22.93% 2.56% 14.82% -0.38% 20.13% 1.54% 17.98% -0.95%	22.51% -3.81% 22.81% 14.29% 32.5% 20.83% 35.34% 14.78% 25.57% 12.86% 27.29% 25.22% 29.01% 32%	21.24% 8.04% 28.75% 19.66% 37.17% 26.77% 42.44% 16.25% 24.94% -23.04% 28.01% 18.66% 25.53% -6.78%	18.04% -17.28% 20.8% 3.44% 20.3% -8.09% 27.77% **23.68% 18.94% -16.74% 18.23% 11.99% 17.92% -11.89%
Platinum (XNYM) Platinum (XNYM) Platinum (XNYM) Silver (XCEC) Silver (XCEC) Silver (XCEC) Silver (XCEC) Aluminium (XLME) Aluminium (XLME) Aluminium (XLME) Aluminium (XLME) Copper (XLME) Copper (XLME)	contango contango backwardation backwardation contango contango backwardation contango contango contango contango	mean sd mean sd mean sd mean sd mean sd mean sd	-3.22% 23.12% 5.98% 21.07% 6.57% 22.93% 2.56% 14.82% -0.38% 20.13% 1.54% 17.98% -0.95% 20.41%	22.51% -3.81% 22.81% 14.29% 32.5% 20.83% 35.34% 14.78% 25.57% 12.86% 27.29% 25.22% 29.01%	21.24% 8.04% 28.75% 19.66% 37.17% 26.77% 42.44% 16.25% 24.94% -23.04% 28.01% 18.66% 25.53% -6.78% 41.31%	18.04% -17.28% 20.8% 3.44% 20.3% -8.09% 27.77% **23.68% 18.94% -16.74% 18.23% 11.99% 17.92% -11.89% 19.86%
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Platinum (XNYM) Platinum (XNYM) Platinum (XNYM) Platinum (XNYM) Silver (XCEC) Silver (XCEC) Silver (XCEC) Silver (XCEC) Aluminium (XLME) Aluminium (XLME) Aluminium (XLME) Copper (XLME) Copper (XLME) Copper (XLME) Copper (XLME) Lead (XLME) Lead (XLME) Lead (XLME) Lead (XLME) Nickel-primary (XLME) Nickel-primary (XLME)	contango contango backwardation backwardation contango contango backwardation contango contango backwardation backwardation backwardation contango contango backwardation contango backwardation backwardation backwardation contango contango contango contango contango backwardation contango contango backwardation contango contango backwardation backwardation	mean sd mean	-3.22% 23.12% 5.98% 21.07% 6.57% 22.93% 2.56% 14.82% -0.38% 20.13% 1.54% 17.98% -0.95% 20.41% 7.59% 19.16% -0.09% 20.12% 23.25% 30.23% 15.46%	22.51% -3.81% 22.81% 14.29% 32.5% 20.83% 35.34% 14.78% 25.57% 12.86% 27.29% 29.01% 32% 32.98% 11.21% 39.91% *46.48% 43.16% -39.24% 41.62% **63.47%	21.24% 8.04% 28.75% 19.66% 37.17% 26.77% 42.44% 16.25% 24.94% -23.04% 28.01% 18.66% 25.53% -6.78% 41.31% 9.87% 35.35% 12.65% 46.22% 16.3% 34.48% -7.67%	18.04% -17.28% 20.8% 3.44% 20.3% -8.09% 27.77% **23.68% 18.94% -16.74% 18.23% 11.99% 17.92% -11.89% 19.86% 17.05% 23.37% -13.41% 23.09% **35.09% 29.13%
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Platinum (XNYM) Platinum (XNYM) Platinum (XNYM) Platinum (XNYM) Silver (XCEC) Silver (XCEC) Silver (XCEC) Silver (XCEC) Aluminium (XLME) Aluminium (XLME) Aluminium (XLME) Copper (XLME) Copper (XLME) Copper (XLME) Copper (XLME) Lead (XLME) Lead (XLME) Lead (XLME) Lead (XLME) Nickel-primary (XLME) Nickel-primary (XLME) Nickel-primary (XLME) Nickel-primary (XLME) Tin-refined (XLME)	contango contango backwardation backwardation contango contango backwardation contango contango contango backwardation backwardation contango contango backwardation contango contango backwardation contango contango contango contango backwardation contango backwardation backwardation backwardation backwardation backwardation backwardation contango contango backwardation	mean sd mean	-3.22% 23.12% 5.98% 21.07% 6.57% 22.93% 2.56% 14.82% -0.38% 20.13% 1.54% 17.98% -0.95% 20.41% 7.59% 19.16% -0.09% 20.12% 23.25% 30.23% 15.46% 34.28% *15.83%	22.51% -3.81% 22.81% 14.29% 32.5% 20.83% 35.34% 14.78% 25.57% 12.86% 27.29% 29.01% 32% 32.98% 11.21% 39.91% *46.48% 43.16% -39.24% 41.62% **63.47% 46.72% 34.62%	21.24% 8.04% 28.75% 19.66% 37.17% 26.77% 42.44% 16.25% 24.94% -23.04% 28.01% 18.66% 25.53% -6.78% 41.31% 9.87% 35.35% 12.65% 46.22% 16.3% 34.48% -7.67% 47.44% 23.7%	18.04% -17.28% 20.8% 3.44% 20.3% -8.09% 27.77% **23.68% 18.94% -16.74% 18.23% 11.99% 17.92% -11.89% 19.86% 17.05% 23.37% -13.41% 23.09% **35.09% 29.13% **-34.64% 28.6% 4.48%

Tin-refined (XLME)	contango	\mathbf{sd}	15.77%	$\boldsymbol{32.08\%}$	41.13%	$\boldsymbol{21.68\%}$
Zinc (XLME)	backwardation	mean	-9.06%	2.95%	$\boldsymbol{10.05\%}$	**29.42%
Zinc (XLME)	backwardation	sd	$\boldsymbol{21.15\%}$	$\boldsymbol{36.86\%}$	33.28 %	$\boldsymbol{22.12\%}$
Zinc (XLME)	contango	mean	-0.13%	37.11%	5.43%	-11.31%
Zinc (XLME)	contango	sd	$\boldsymbol{22.38\%}$	40.51%	$\boldsymbol{39.03\%}$	24.32 %
US commodities	backwardation	mean	***15.44%	*15.66%	13.4%	6.82%
US commodities	backwardation	$\operatorname{\mathbf{sd}}$	9.88%	14.21 %	14.83%	8.59%
US commodities	contango	mean	-1.09%	*16.16%	4.9%	-6.33%
US commodities	contango	\mathbf{sd}	$\boldsymbol{10.09\%}$	$\boldsymbol{13.05\%}$	19.98%	$\boldsymbol{10.95\%}$
GB commodities	backwardation	mean	4.41%	8.01%	15.06%	**19.95%
GB commodities	backwardation	\mathbf{sd}	14.39%	$\boldsymbol{25.92\%}$	$\boldsymbol{25.94\%}$	15.13%
GB commodities	contango	mean	0.34%	*35.56%	-5.24%	-14.05%
GB commodities	contango	sd	$\boldsymbol{16.87\%}$	$\boldsymbol{27.88\%}$	$\boldsymbol{34.06\%}$	$\boldsymbol{17.1\%}$

Table 3: This table shows the average time series \mathbb{R}^2 for models where the returns series for the twenty four individual US commodities are independently regressed against relative change in their own commercial hedging pressure (CHP) series as well as against that in aggregate CHP, a market wide measure of CHP which calculation method is discussed in section Section 2.3. Averages are calculated independently across the whole cross-section of US traded commodity assets considered in the study as well as across commodity sectors and subsectors for the four periods of interest (past: 1997-2003; financialisation: 2004-2008; crisis: 2008-2013; post-crisis: 2013-2018). The results are discussed in section Section 3.

regressor	sector	subsector	past	financialisation	crisis	post-crisis
$\Delta\%$ commodity CHP	all	all	$\boldsymbol{25.79\%}$	20.72%	$\boldsymbol{22.46\%}$	$\boldsymbol{27.17\%}$
$\Delta\%$ commodity CHP	agriculturals	all	$\boldsymbol{25.19\%}$	21.6%	23.91 %	$\boldsymbol{28.12\%}$
$\Delta\%$ commodity CHP	agriculturals	grains	33.18%	$\boldsymbol{27.86\%}$	$\boldsymbol{29.15\%}$	$\boldsymbol{36.24\%}$
$\Delta\%$ commodity CHP	agriculturals	livestock	$\boldsymbol{6.23\%}$	10.53%	$\boldsymbol{6.89\%}$	$\boldsymbol{4.34\%}$
$\Delta\%$ commodity CHP	agriculturals	softs	$\boldsymbol{26.68\%}$	$\boldsymbol{20.89\%}$	$\boldsymbol{27.19\%}$	31.88%
$\Delta\%$ commodity CHP	energy	all	25.12%	19.45%	16%	10.96%
$\Delta\%$ commodity CHP	energy	gas	19.69%	15.65%	7.36%	5.29%
$\Delta\%$ commodity CHP	energy	petroleum	$\boldsymbol{26.93\%}$	$\boldsymbol{20.71\%}$	18.87%	$\boldsymbol{12.85\%}$
$\Delta\%$ commodity CHP	metals	all	$\boldsymbol{28.12\%}$	19.1%	23.27%	37.32%
$\Delta\%$ commodity CHP	metals	base	47.38%	25.86%	18.07%	$\boldsymbol{34.9\%}$
$\Delta\%$ commodity CHP	metals	precious	$\boldsymbol{23.3\%}$	17.41%	$\boldsymbol{24.57\%}$	37.93%
$\Delta\%$ aggregate CHP	all	all	5.37%	7.12%	$\boldsymbol{14.72\%}$	7.49%
$\Delta\%$ aggregate CHP	agriculturals	all	$\boldsymbol{4.91\%}$	4.5%	$\boldsymbol{12.14\%}$	6.23%
$\Delta\%$ aggregate CHP	agriculturals	grains	7.25%	5.4%	19.75%	8.45%
$\Delta\%$ aggregate CHP	agriculturals	livestock	1.35%	0.41%	0.52%	0.26%
$\Delta\%$ aggregate CHP	agriculturals	softs	4.34%	5.64%	10.35%	6.98%
$\Delta\%$ aggregate CHP	energy	all	3.35%	6.27%	14.83%	2.45%
$\Delta\%$ aggregate CHP	energy	gas	$\boldsymbol{2.65\%}$	2.58%	2.6%	0.27 %
$\Delta\%$ aggregate CHP	energy	petroleum	3.59%	7.49%	18.9%	3.17%
$\Delta\%$ aggregate CHP	metals	all	8.36%	15.67%	22.37%	15.33%
$\Delta\%$ aggregate CHP	metals	base	$\boldsymbol{10.12\%}$	$\boldsymbol{12.74\%}$	$\boldsymbol{23.94\%}$	7.67%
$\Delta\%$ aggregate CHP	metals	precious	7.92%	16.4%	$\boldsymbol{21.98\%}$	17.25%

Table 4: This table shows the average pairwise correlation coefficients calculated independently across the whole cross-section of commodity assets considered in the study as well as across countries and commodity sectors and subsectors for the four periods of interest (past: 1997-2003; financialisation: 2004-2008; crisis: 2008-2013; post-crisis: 2013-2018). For each period, the results are calculated independently over the entire period as well as over phases of aggregate backwardation and contango. Aggregate CHP construction and corresponding regime definitions are discussed in section Section 2.3 while the results are discussed in section Section 3.

country	sector	$\operatorname{subsector}$	regime	past	financialisation	crisis	post-crisis
all	all	all	whole period	0.0737	0.1638	0.2849	0.1282
all	all	all	backwardation	0.0724	0.1790	0.2709	0.1001
all	all	all	contango	0.0748	0.1501	0.2985	0.1495
$\mathbf{U}\mathbf{S}$	all	all	whole period	0.0684	0.1545	0.2526	0.1096
\mathbf{US}	all	all	backwardation	0.0642	0.1697	0.2323	0.0859
$\mathbf{u}\mathbf{s}$	all	all	contango	0.0720	0.1402	0.2706	0.1268
US	agriculturals	all	whole period	0.0910	0.1305	0.2322	0.1075
$\mathbf{U}\mathbf{S}$	agriculturals	all	backwardation	0.0799	0.1421	0.1975	0.0955
US	agriculturals	all	contango	0.1018	0.1184	0.2611	0.1186
\mathbf{US}	agriculturals	grains	whole period	0.4421	0.4898	0.5673	0.3728
US	agriculturals	grains	backwardation	0.4312	0.5283	0.5749	0.3289
\mathbf{US}	agriculturals	grains	contango	0.4570	0.4510	0.5675	0.4161
\mathbf{US}	agriculturals	livestock	whole period	0.3340	0.2868	0.3707	0.3333
$\mathbf{U}\mathbf{S}$	agriculturals	livestock	backwardation	0.3053	0.2652	0.3906	0.3220
\mathbf{US}	agriculturals	livestock	contango	0.3605	0.3150	0.3557	0.3483
\mathbf{US}	agriculturals	softs	whole period	0.0391	0.0853	0.1584	0.0764
US	agriculturals	softs	backwardation	0.0437	0.0845	0.1210	0.0685
$\mathbf{U}\mathbf{S}$	agriculturals	softs	contango	0.0334	0.0874	0.1937	0.0833
\mathbf{US}	energy	all	whole period	0.4908	0.5892	0.4687	0.4203
\mathbf{US}	energy	all	backwardation	0.4844	0.5949	0.4601	0.4179
$\mathbf{u}\mathbf{s}$	energy	all	contango	0.4957	0.5869	0.4779	0.4210
$\mathbf{U}\mathbf{S}$	energy	petroleum	whole period	0.7278	0.7846	0.7692	0.7175
US	energy	petroleum	backwardation	0.6962	0.7854	0.7727	0.7282
$\mathbf{U}\mathbf{S}$	energy	petroleum	contango	0.7606	0.7856	0.7716	0.7134
\mathbf{US}	metals	all	whole period	0.2167	0.5553	0.5877	0.4721
\mathbf{US}	metals	all	backwardation	0.2201	0.5418	0.6353	0.4155
US	metals	all	contango	0.2145	0.5727	0.5709	0.5035
$\mathbf{U}\mathbf{S}$	metals	precious	whole period	0.2883	0.6305	0.6704	0.5762
\mathbf{US}	metals	precious	backwardation	0.3070	0.6061	0.6991	0.5386
\mathbf{US}	metals	precious	contango	0.2713	0.6591	0.6615	0.5949
GB	all	all	whole period	0.4047	0.4691	0.6536	0.4412
GB	all	all	backwardation	0.3987	0.4753	0.6729	0.3954
GB	all	all	contango	0.4108	0.4652	0.6456	0.4797

Table 5: This table shows the average pairwise correlation coefficients calculated independently across the whole cross-section of commodity assets considered in the study as well as across countries and commodity sectors and subsectors by year for the whole period of interest (1997-2018). The results are discussed in section Section 3.

country	sector	subsector	decade	0	1	2	3	4	5	6	7	8	9
all	all	all	1990								0.0401	0.0812	0.0822
all	all	all	2000	0.0652	0.0719	0.0796	0.0848	0.1242	0.1247	0.1650	0.1541	0.3334	0.2913
all	all	all	2010	0.2743	0.2837	0.2235	0.1432	0.1139	0.1547	0.1339	0.0963	0.1374	
$\mathbf{U}\mathbf{S}$	all	all	1990								0.0536	0.0811	0.0803
US	all	all	2000	0.0597	$\boldsymbol{0.0652}$	0.0600	0.0723	0.1073	0.1164	0.1502	0.1479	0.3284	0.2621
$\mathbf{U}\mathbf{S}$	all	all	2010	0.2275	0.2541	0.1850	0.1018	0.0986	0.1371	0.1064	0.0905	0.1197	
US	agriculturals	all	1990								0.0827	0.0916	0.1121
US	agriculturals	all	2000	0.0918	0.0967	0.0734	0.0827	0.1111	0.1090	0.1125	0.1131	0.3034	0.2507
US	agriculturals	all	2010	0.2038	0.2304	0.1583	0.0716	0.0846	0.1360	0.1121	0.1148	0.1181	
\mathbf{US}	agriculturals	grains	1990								0.5243	0.4882	0.5650
$\mathbf{U}\mathbf{S}$	agriculturals	grains	2000	0.4870	0.4337	0.3851	0.3402	0.4473	0.5335	0.4594	0.5040	0.6325	0.5681
US	agriculturals	grains	2010	0.5389	0.6424	0.5592	0.3199	0.3134	0.4390	0.3872	0.4123	0.3618	
US	agriculturals	livestock	1990								0.2192	0.3724	0.3453
$\mathbf{U}\mathbf{S}$	agriculturals	livestock	2000	0.2762	0.3315	0.3243	0.3497	0.2937	0.4170	0.2784	0.2744	0.2822	0.3365
\mathbf{US}	agriculturals	livestock	2010	0.3568	0.4772	0.3585	0.1783	0.3532	0.4246	0.3445	0.2948	0.2994	
US	agriculturals	softs	1990								-0.0042	-0.0001	0.0128
US	agriculturals	softs	2000	0.0603	0.0289	0.0470	0.0807	0.0577	0.0624	0.0765	0.0485	0.2484	0.0128
US	agriculturals	softs	2010	0.1273	0.0200	0.1038	0.0568	0.0311	0.0024	0.1280	0.0435	0.0565	0.1710
US	energy	all	1990	0.1210	0.1400	0.1000	0.0000	0.0410	0.0000	0.1200	0.4060	0.4552	0.5256
US	energy	all	2000	0.4362	0.5186	0.5682	0.5248	0.6212	0.6402	0.5082	0.5530	0.6278	0.4840
US	energy	all	2010	0.4549	$\boldsymbol{0.4752}$	0.3796	0.2889	0.4103	0.4282	0.4442	0.3890	0.4270	
US	energy	petroleum	1990	0.0100	0 =000	0.00	0.00.40	0.0004			0.7207	0.7946	0.8527
US	energy	petroleum	2000	0.6122	0.7632	0.8072	0.6948	0.8321	0.7377	0.7703	0.7895	0.8184	0.7460
US	energy	petroleum	2010	0.8835	0.7867	0.6692	0.6075	0.6710	0.7069	0.7547	0.6672	0.7786	0.0440
\mathbf{US}	metals	all	1990								0.2296	0.2877	0.2449
\mathbf{US}	metals	all	2000	0.1860	0.1832	0.1992	0.2290	0.4923	0.4480	0.6328	0.5426	0.5664	0.5547
\mathbf{US}	metals	all	2010	0.6297	0.6458	0.6649	0.6160	0.4531	0.4850	0.4269	0.3584	0.5259	
$\mathbf{U}\mathbf{S}$	metals	precious	1990								0.3063	0.3451	0.3308
US	metals	precious	2000	0.2934	0.2337	0.3053	0.2959	0.5688	0.5674	0.6969	0.6143	0.6428	0.6713
$\mathbf{U}\mathbf{S}$	metals	precious	2010	0.7004	0.7107	0.7089	0.6988	0.5620	0.6088	0.5652	0.4746	0.5716	
GB	all	all	1990								0.2289	0.3601	0.4487
GB	all	all	2000	0.3550	0.4082	0.4775	0.5526	0.5068	0.4792	0.4340	0.4653	0.5795	0.6647
GB	all	all	2010	0.6886	0.6742	0.6122	0.6606	0.4700	0.5003	0.4636	0.3810	0.3598	

Table 6: This table shows the average time series regression coefficients for models where the returns series for all the commodity assets considered in the study are independently regressed against the return series for our market factor, an equally weighted portfolio of the 24 US commodities considered in the study. The results are calculated independently across countries as well as commodity sectors and subsectors for the four periods of interest (past: 1997-2003; financialisation: 2004-2008; crisis: 2008-2013; post-crisis: 2013-2018). The market factor construction is discussed in section Section 2.3 while the results are discussed in section Section 3.

country	sector	${f subsector}$	\mathbf{past}	financial is at ion	crisis	post-crisis
all	all	all	0.8556	0.9709	1.0352	0.9438
$\mathbf{U}\mathbf{S}$	agriculturals	all	0.7706	0.7965	0.8535	0.8252
$\mathbf{U}\mathbf{S}$	agriculturals	grains	0.9891	1.1982	1.1727	1.0301
$\mathbf{U}\mathbf{S}$	agriculturals	livestock	0.4320	0.2059	0.3317	0.5484
$\mathbf{U}\mathbf{S}$	agriculturals	softs	0.7213	0.6900	0.7953	0.7589
$\mathbf{U}\mathbf{S}$	energy	all	2.3308	1.6106	1.3977	1.7639
$\mathbf{U}\mathbf{S}$	energy	gas	2.6992	1.7090	1.0044	1.2585
$\mathbf{U}\mathbf{S}$	energy	petroleum	2.2080	1.5778	1.5288	1.9324
$\mathbf{U}\mathbf{S}$	metals	all	0.6320	1.1252	1.1212	0.9131
\mathbf{US}	metals	base	0.4902	1.1178	1.4140	0.8780
\mathbf{US}	metals	precious	0.6674	1.1271	1.0480	0.9219
GB	all	all	0.2713	0.8520	1.1761	0.7192

Table 7: This table shows the average time series regression R^2 for models where the returns series for all the commodity assets considered in the study are independently regressed against the return series for our mimicking portfolios for risk factors in returns, including the market, CHP, open interest and term structure factors. The results are calculated independently across countries and commodity sectors and subsectors for the four periods of interest (past: 1997-2003; financialisation: 2004-2008; crisis: 2008-2013; post-crisis: 2013-2018); for each period the results are calculated independently over the entire period as well as over phases of aggregate backwardation and contango. Factors as well as aggregate CHP construction and corresponding regime definitions are discussed in section Section 2.3 while the results are discussed in section Section 3.

country	sector	subsector	factor	regime	past	financialisation	crisis	post-crisis
all	all	all	market	whole period	9.69%	19.7%	31.3%	14.73%
all	all	all	market	backwardation	9.44%	$\boldsymbol{21.55\%}$	$\boldsymbol{30.04\%}$	$\boldsymbol{11.82\%}$
all	all	all	market	contango	$\boldsymbol{10.07\%}$	18.15%	$\boldsymbol{32.67\%}$	$\boldsymbol{16.99\%}$
all	all	all	CHP	whole period	$\boldsymbol{1.93\%}$	5.39%	4.43%	3.78%
all	all	all	CHP	backwardation	2.6%	5.61%	4%	4.57%
all	all	all	CHP	contango	1.54%	5.55%	4.99%	4.09%
all	all	all	open interest	whole period	5.98%	5.63%	4.09%	4.32%
all	all	all	open interest	backwardation	5.69%	5.47%	3.89%	4.12%
all	all	all	open interest	contango	6.41%	6.21%	4.73%	4.74%
all	all	all	term structure	whole period	$\boldsymbol{1.52\%}$	2.95%	5.2%	$\boldsymbol{1.39\%}$
all	all	all	term structure	backwardation	1.87%	2.7%	4.56%	2.3%
all	all	all	term structure	contango	$\boldsymbol{1.62\%}$	3.59%	5.93%	1.12%
\mathbf{US}	all	all	market	whole period	11.67%	21.7%	30.69%	15.73%
\mathbf{US}	all	all	market	backwardation	$\boldsymbol{11.26\%}$	23.68 %	$\boldsymbol{28.95\%}$	$\boldsymbol{13.12\%}$
US	all	all	market	contango	$\boldsymbol{12.22\%}$	19.97%	$\boldsymbol{32.48\%}$	17.66%
$\mathbf{u}\mathbf{s}$	all	all	CHP	whole period	2.4%	5.8%	$\boldsymbol{5.39\%}$	4.47%
$\mathbf{U}\mathbf{S}$	all	all	CHP	backwardation	3.19%	5.94%	4.72%	5.54%
$\mathbf{U}\mathbf{S}$	all	all	CHP	contango	$\boldsymbol{1.91\%}$	6.1%	$\boldsymbol{6.14\%}$	$\boldsymbol{4.79\%}$
$\mathbf{U}\mathbf{S}$	all	all	open interest	whole period	7.45%	6.95%	4.57%	5.29%
$\mathbf{U}\mathbf{S}$	all	all	open interest	backwardation	7.1%	$\boldsymbol{6.79\%}$	4.5%	5%
\mathbf{US}	all	all	open interest	contango	7.95%	7.61%	5.24%	5.82%
\mathbf{US}	all	all	term structure	whole period	$\boldsymbol{1.89\%}$	2.87%	5.12%	1.71%
\mathbf{US}	all	all	term structure	backwardation	2.3%	2.78%	$\boldsymbol{4.79\%}$	2.85%
\mathbf{US}	all	all	term structure	contango	2.01%	3.41%	5.73%	1.34%
$\mathbf{u}\mathbf{s}$	agriculturals	all	market	whole period	9.08%	16.04%	$\boldsymbol{25.99\%}$	$\boldsymbol{12.07\%}$
$\mathbf{U}\mathbf{S}$	agriculturals	all	market	backwardation	8.57%	18.68%	23.01 %	11.56%
$\mathbf{U}\mathbf{S}$	agriculturals	all	market	contango	9.85%	13.5%	$\boldsymbol{28.67\%}$	$\boldsymbol{12.91\%}$
$\mathbf{U}\mathbf{S}$	agriculturals	all	CHP	whole period	1.03%	1.6%	1.47%	1.61%
$\mathbf{U}\mathbf{S}$	agriculturals	all	CHP	backwardation	1.14%	2.47%	1.72%	3.11%
$\mathbf{U}\mathbf{S}$	agriculturals	all	CHP	contango	1.21%	1.31%	$\boldsymbol{1.56\%}$	1.1%
US	agriculturals	all	open interest	whole period	1.09%	1.9%	0.45%	1%
\mathbf{US}	agriculturals	all	open interest	backwardation	0.85%	3.74%	$\boldsymbol{1.57\%}$	2.01%
$\mathbf{U}\mathbf{S}$	agriculturals	all	open interest	contango	1.58%	0.97%	0.58%	0.67%
$\mathbf{U}\mathbf{S}$	agriculturals	all	term structure	whole period	$\boldsymbol{1.28\%}$	1.36%	$\boldsymbol{1.73\%}$	2.16%
US	agriculturals	all	term structure	backwardation	1.2%	1.27%	1.93%	3.55%
$\mathbf{U}\mathbf{S}$	agriculturals	all	term structure	contango	1.72%	1.91%	2.15%	1.69%
\mathbf{US}	agriculturals	grains	market	whole period	$\boldsymbol{16.44\%}$	29.28%	$\boldsymbol{42.09\%}$	$\boldsymbol{19.27\%}$
\mathbf{US}	agriculturals	grains	market	backwardation	14.54%	34.68%	$\boldsymbol{38.84\%}$	18.3%

U	S a	griculturals	grains	market	contango	18.73%	23.95 %	44.4%	$\boldsymbol{20.71\%}$
U	S a	griculturals	grains	CHP	whole period	0.64%	1.37%	0.78%	$\boldsymbol{2.98\%}$
U	S a	griculturals	grains	CHP	backwardation	0.77%	2.63%	1.04%	5.46%
U	S a	griculturals	grains	CHP	contango	0.74%	$\boldsymbol{1.06\%}$	0.7%	$\boldsymbol{1.36\%}$
U	S a	griculturals	grains	open interest	whole period	2.22%	4.18%	0.67%	2.04%
U	S a	griculturals	grains	open interest	backwardation	1.38%	7.87%	2.5%	3.8%
U	S a	griculturals	grains	open interest	contango	3.3%	$\boldsymbol{1.92\%}$	0.69%	1.24%
	ra.					4 MODY	4 0 1017	0.0407	0.0007
		griculturals	grains	term structure	whole period	1.78%	1.37%	2.04%	2.08%
		griculturals	grains	term structure	backwardation	1.47%	1.19%	1.87%	3.56%
		griculturals	grains	term structure	contango	2.25%	2.18%	3.3%	1.72%
U		0	livestock	market	whole period	2.96%	1.83%	8.99%	4.84%
U	S a	griculturals	livestock	market	backwardation	2.14%	2.15%	6.2%	$\boldsymbol{4.32\%}$
U	S a	griculturals	livestock	market	contango	4.03%	1.52%	11.26%	5.94%
		griculturals	livestock	CHP	whole period	1.62%	2.06%	3.36%	1.23%
		~	livestock	CHP	backwardation	1.95%	2.94%	3.54%	1.94%
		griculturals	livestock	CHP	contango	1.35%	1.29%	3.71%	0.89%
		griculturals	livestock	open interest	whole period	0.43%	0.55%	0.18%	0.34%
				open meerest	•				
		griculturals	livestock	open interest	backwardation	0.44%	0.88%	0.39%	0.66%
U	S a	griculturals	livestock	open interest	contango	0.5%	0.3%	0.19%	0.28%
		griculturals	livestock	term structure	whole period	0.38%	1.25%	1.15%	$\boldsymbol{1.79\%}$
U	S a	griculturals	livestock	term structure	backwardation	0.59%	$\boldsymbol{1.09\%}$	1.43%	4.47%
U	S a	griculturals	livestock	term structure	contango	0.55%	2.17%	1.1%	0.37%
TI	S a	griculturals	softs	market	whole period	4.78%	9.9%	18.39%	8.48%
		griculturals	softs	market	backwardation	5.82%	10.95%	15.58%	8.44%
		griculturals	softs	market	contango	3.87%	9.02%	21.64%	8.6%
		griculturals	softs	CHP	whole period	1.12%	$\frac{9.02\%}{1.59\%}$	1.22%	0.44%
		griculturals	softs	CHP	backwardation	1.11%	2.07%	1.49%	1.35%
U	o a	igriculturais	SOILS	CHF	Dackwardation	1.11/0	2.01/0	1.49/0	1.55/0
U	S a	griculturals	softs	CHP	contango	1.6%	$\boldsymbol{1.57\%}$	$\boldsymbol{1.36\%}$	$\boldsymbol{0.94\%}$
U	S a	griculturals	softs	open interest	whole period	0.29 %	0.28%	0.36%	0.28%
U	S a	griculturals	softs	open interest	backwardation	0.51%	1.04%	1.23%	0.9%
U	S a	griculturals	softs	open interest	contango	0.4%	0.35%	0.66%	0.3%
U	S a	griculturals	softs	term structure	whole period	1.24%	1.41%	1.71%	$\boldsymbol{2.42\%}$
U	S a	griculturals	softs	term structure	backwardation	1.25%	1.44%	2.25%	3.08%
U		griculturals	softs	term structure	contango	1.78%	1.53%	1.51%	2.32%
		energy	all	market	whole period	26.58%	31.57%	38.94%	25.9%
U		energy	all	market	backwardation	26.08%	32.3%	33.74%	20.94%
		energy	all	market	contango	26.84%	31.47%	42.16%	28.77%
TI	S e	energy	all	CHP	whole period	9.18%	1.61%	3.5%	4.79%
		energy	all	CHP	backwardation	12.81%	1.01%	2.97%	5.68%
		energy	all	CHP	contango	6.05%	2.32%	4.09%	7.91%
			all	open interest	whole period	40%	33.46%	21.16%	27.67%
		energy	all	•	backwardation	38.82%	25.93%	17.59%	22.07%
U	ъ е	energy	an	open interest	DackwaruatiOII	JU.04/0	49.93 /0	T1.99/0	44.01/0
U	S e	energy	all	open interest	contango	41.02%	$\boldsymbol{40.42\%}$	23.88 %	$\boldsymbol{32.13\%}$
U	S e	energy	all	term structure	whole period	3.41%	2.99%	5.5%	$\boldsymbol{1.86\%}$
U	S e	energy	all	term structure	backwardation	7.4%	3.75%	4.93%	3.4%
U	S e	energy	all	term structure	contango	1.16%	2.76%	5.88%	$\boldsymbol{1.26\%}$

$\mathbf{U}\mathbf{S}$	energy	gas	market	whole period	19.37%	18.38%	10.88%	7.86%
$\mathbf{U}\mathbf{S}$	energy	gas	market	backwardation	21.36%	17.54%	$\boldsymbol{12.6\%}$	7.17%
US	energy	gas	market	contango	17.12%	20.07%	9.92%	7.96%
US	energy	gas	CHP	whole period	5.37%	3.62%	12.6%	16.2%
US	energy	gas	CHP	backwardation	10.16%	2.14%	11.42%	19.3%
\mathbf{US}	energy	gas	CHP	contango	1.59%	5.26%	13.46%	13.84%
US	onorgy	gas	open interest	whole period	22.08%	27.47%	15.46%	16.03%
US	energy	gas	open interest	backwardation	23.23%	21.16%	13.40% $14.53%$	16.64%
US	energy	gas	open interest	contango	20.54%	32.65%	16.18%	15.41%
US	energy	gas	term structure	whole period	0.41%	4.24%	7.16%	5.22%
US	energy	gas	term structure	backwardation	0.36%	2.03%	11.66%	5.9%
OB	chergy	gas	term structure	backwai dation		2.0070	11.0070	0.070
\mathbf{US}	\mathbf{energy}	gas	term structure	contango	$\boldsymbol{0.48\%}$	6.51%	4.6%	4.55%
$\mathbf{U}\mathbf{S}$	\mathbf{energy}	petroleum	market	whole period	$\boldsymbol{28.98\%}$	$\boldsymbol{35.97\%}$	$\boldsymbol{48.3\%}$	$\boldsymbol{31.91\%}$
$\mathbf{u}\mathbf{s}$	energy	${f petroleum}$	market	backwardation	$\boldsymbol{27.65\%}$	$\boldsymbol{37.22\%}$	$\boldsymbol{40.79\%}$	25.53%
$\mathbf{U}\mathbf{S}$	energy	petroleum	market	contango	30.08%	35.26%	$\boldsymbol{52.9\%}$	35.71%
$\mathbf{U}\mathbf{S}$	energy	petroleum	CHP	whole period	$\boldsymbol{10.44\%}$	0.94%	0.47%	0.99%
US	onorgy	petroleum	CHP	backwardation	13.69%	0.63%	0.15%	1.14%
US	energy	petroleum	CHP	contango	7.53%	1.34%	0.15%	5.93%
US		petroleum	open interest	whole period	45.97%	35.46%	23.06%	31.56%
US	energy	•	•	backwardation	44.02%	27.52%	18.61%	23.88%
US	energy	petroleum petroleum	open interest open interest	contango	47.85%	43.02%	26.44%	37.7%
US	energy	petroleum	open interest	containgo	41.65/0	45.02/0	20.44/0	31.170
\mathbf{US}	energy	petroleum	term structure	whole period	4.41%	2.57%	$\boldsymbol{4.95\%}$	0.74%
$\mathbf{u}\mathbf{s}$	energy	petroleum	term structure	backwardation	9.74%	4.32%	2.68%	2.56%
$\mathbf{U}\mathbf{S}$	energy	petroleum	term structure	contango	1.38%	1.51%	6.3%	0.16%
$\mathbf{U}\mathbf{S}$	metals	all	market	whole period	7.54%	$\boldsymbol{30.79\%}$	38.2%	18.58%
US	metals	all	market	backwardation	7.5%	31.77%	$\boldsymbol{42.95\%}$	11.54%
US	metals	all	market	contango	7.63%	30.19%	36.15%	23%
US	metals	all	CHP	whole period	1.07%	21.76%	18.67%	12.79%
US	metals	all	CHP	backwardation	1.66%	20.33%	15.15%	12.79%
US	metals	all	CHP	contango	0.72%	23.49%	$\frac{15.15\%}{21.5\%}$	13.38%
US	metals	all		whole period	0.72%	0.91%	3.68%	0.27%
US	metais	an	open interest	whole period	0.5%	0.91%	3.00%	0.2170
$\mathbf{u}\mathbf{s}$	metals	all	open interest	backwardation	0.47 %	0.61%	2.83%	$\boldsymbol{0.32\%}$
$\mathbf{U}\mathbf{S}$	metals	all	open interest	contango	0.61%	1.28%	4.31%	0.24 %
US	metals	all	term structure	whole period	2.48%	7.31%	14.99%	0.24 %
$\mathbf{U}\mathbf{S}$	metals	all	term structure	backwardation	1.5%	6.51%	$\boldsymbol{13.24\%}$	0.33%
\mathbf{US}	metals	all	term structure	contango	3.57%	8.4%	$\boldsymbol{16.35\%}$	0.34%
US	metals	base	market	whole period	5.39%	23.13%	50.77%	19.74%
US	metals	base	market	backwardation	4.96%	25.38%	48.58%	10.15%
US	metals	base	market	contango	5.66%	21.55%	52.45%	27.5%
US	metals	base	CHP	whole period	0.58%	5.78%	0.57%	0.6%
US	metals	base	CHP	backwardation	1.4%	5.18%	1.91%	0.29%
US	metals	base	CHP	contango	0.1%	6.42%	0.21%	0.88%
US	metals	base	open interest	whole period	0.18%	1.01%	2.78%	0.46%
US	metals	base	open interest	backwardation	0.1%	0.34%	1.84%	0.61%
US	metals	base	open interest	contango	0.24%	1.82%	3.42%	0.37%
\mathbf{US}	$_{ m metals}$	base	term structure	whole period	0.06%	9.72%	9.63%	$\boldsymbol{0.24\%}$

$\mathbf{U}\mathbf{S}$	metals	base	term structure	backwardation	0.37 %	7.74%	7.44%	0.12%
$\mathbf{U}\mathbf{S}$	metals	base	term structure	contango	$\boldsymbol{0.02\%}$	$\boldsymbol{11.67\%}$	$\boldsymbol{11.07\%}$	$\boldsymbol{0.43\%}$
\mathbf{US}	metals	precious	market	whole period	8.08%	32.7%	35.06%	18.3%
$\mathbf{U}\mathbf{S}$	metals	precious	market	backwardation	8.13%	33.37%	41.54%	11.89%
$\mathbf{U}\mathbf{S}$	metals	precious	market	contango	8.12%	$\boldsymbol{32.35\%}$	$\boldsymbol{32.07\%}$	$\boldsymbol{21.88\%}$
***			CIID		4 007	0 F F007	00 1007	4 7 0 407
US	metals	precious	CHP	whole period	1.2%	25.76%	23.19%	15.84%
US	metals	precious	CHP	backwardation	1.72%	24.12%	18.46%	15.81%
US	metals	precious	CHP	contango	0.87%	27.76%	26.82%	16.5%
$\mathbf{u}\mathbf{s}$	$_{ m metals}$	precious	open interest	whole period	0.57%	0.89%	3.9%	0.22%
$\mathbf{u}\mathbf{s}$	$_{ m metals}$	precious	open interest	backwardation	0.56%	0.68%	3.07%	$\boldsymbol{0.24\%}$
US	metals	precious	open interest	contango	0.71%	1.15%	4.53%	0.21%
US	metals	precious	term structure	whole period	3.08%	6.71%	16.33%	0.24%
US	metals	precious	term structure	backwardation	1.79%	6.21%	14.69%	0.39%
US	metals	precious	term structure	contango	4.46%	7.58%	17.68%	0.31%
GB	all	all	market					
GB	an	an	market	whole period	1.74%	11.71%	33.71%	10.75%
$_{\mathrm{GB}}$	all	all	market	backwardation	2.14%	$\boldsymbol{13.03\%}$	$\boldsymbol{34.39\%}$	$\boldsymbol{6.62\%}$
GB	all	all	market	contango	1.48%	10.88%	33.45%	14.35%
GB	all	all	CHP	whole period	0.08%	3.74%	0.57%	1.02%
GB	all	all	CHP	backwardation	0.24 %	4.26%	1.09%	0.72%
GB	all	all	CHP	contango	0.03%	3.33%	0.4%	1.31%
a D		,,			0.070	0.0104	0 1 =04	0.4007
GB	all	all	open interest	whole period	0.07%	0.34%	2.17%	0.42%
GB	all	all	open interest	backwardation	0.08%	0.18%	1.44%	0.57%
GB	all	all	open interest	contango	0.22 %	0.59%	2.7%	0.39%
GB	all	all	term structure	whole period	0.04%	3.28%	5.5%	0.09%
GB	all	all	term structure	backwardation	0.17%	2.37%	3.67%	0.09%
$_{ m GB}$	all	all	term structure	contango	0.02%	4.34%	$\boldsymbol{6.72\%}$	$\boldsymbol{0.24\%}$

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