

# Package: contagionchannels (via r-universe)

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**Type** Package

**Title** Two-Stage Detection and Attribution of Cross-Border Financial Contagion Channels

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**Description** Implementation of a two-stage framework for the joint detection-and-attribution of cross-border financial contagion. Stage one detects directional information flows between equity markets via Wavelet-Quantile Transfer Entropy, combining maximal-overlap discrete wavelet decomposition (Percival and Walden, 2000, ISBN:9780521685085) with the transfer-entropy estimator of Schreiber (2000) <[doi:10.1103/PhysRevLett.85.461](https://doi.org/10.1103/PhysRevLett.85.461)> and quantile conditioning following Han, Linton, Oka and Whang (2016) <[doi:10.1016/j.jeconom.2016.03.001](https://doi.org/10.1016/j.jeconom.2016.03.001)>. Stage two attributes each significant directional link to one of five mutually exclusive transmission channels (Trade, Financial, Geopolitical, Behavioural, Monetary Policy) through a multi-method structural identification architecture combining instrumental-variables two-stage least squares with channel-specific external instruments (Stock and Watson, 2018) <[doi:10.1111/ecoj.12593](https://doi.org/10.1111/ecoj.12593)>, LASSO-based instrument selection (Belloni, Chernozhukov and Hansen, 2014) <[doi:10.1093/restud/rdt044](https://doi.org/10.1093/restud/rdt044)>, local projections (Jorda, 2005) <[doi:10.1257/0002828053828518](https://doi.org/10.1257/0002828053828518)>, heteroskedasticity-based identification (Rigobon, 2003) <[doi:10.1162/003465303772815727](https://doi.org/10.1162/003465303772815727)>, and the Cinelli-Hazlett (2020) <[doi:10.1111/rssb.12348](https://doi.org/10.1111/rssb.12348)> robustness-value sensitivity bound. Bundled datasets and replication scripts reproduce the headline findings of Bhandari, Parida and Sahu (2026) <[doi:10.48550/arXiv.2604.26546](https://doi.org/10.48550/arXiv.2604.26546)>; the package is general-purpose and accommodates user-supplied returns and channel proxies.

**License** GPL-3

**Encoding** UTF-8

**LazyData** true

**Depends** R (>= 4.1.0)

**Imports** xts, zoo, waveslim, quantreg, igraph, parallel, MASS, stats, utils

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**VignetteBuilder** knitr

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## Contents

build_channel_composites . . . . .	3
build_lag . . . . .	4
build_network . . . . .	4
channel_proxies . . . . .	5
cinelli_hazlett_rv . . . . .	6
compute_wqte_matrix . . . . .	7
crisis_periods . . . . .	8
g20_returns . . . . .	8
iv_2sls_attribute . . . . .	9
lasso_iv_attribute . . . . .	10
load_paper_data . . . . .	11
local_projections . . . . .	12
modwt_detail . . . . .	13
network_summary . . . . .	14
orthogonalise_residual . . . . .	14
period_subset . . . . .	15
plot_attribution_evolution . . . . .	16
plot_attribution_stack . . . . .	16
plot_qte_intensity . . . . .	17
plot_robustness_value . . . . .	17
qte_pair . . . . .	18
rigobon_id . . . . .	19
run_contagion_pipeline . . . . .	20
walktrap_communities . . . . .	21
zscore . . . . .	22

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`build_channel_composites`*Construct the Five-Channel Composites*

---

## Description

Builds the five contagion-channel composites (Trade, Financial, Geopolitical, Behavioural, Monetary Policy) from a data.frame of raw channel proxies. The construction is engineered for orthogonal identification: the trade composite uses log-returns of a broad trade-weighted dollar index (genuinely time-varying), the behavioural composite is orthogonalised against the financial composite to avoid VIX-derivative contamination, and the monetary composite uses a first-differenced policy rate to remove persistence-induced first-stage F-inflation.

## Usage

```
build_channel_composites(proxy_grid)
```

## Arguments

<code>proxy_grid</code>	A data.frame with the following columns aligned on the same daily date grid: <code>vix</code> , <code>hy_spread</code> , <code>stress_index</code> , <code>usd_index</code> , <code>gpr</code> , <code>geo_events</code> , <code>sentiment</code> , <code>fed_rate</code> , <code>term_spread</code> , <code>qe_dummy</code> .
-------------------------	--

## Value

A data.frame with columns `Date`, `Trade`, `Financial`, `Geopolitical`, `Behavioral`, and `Monetary_Policy`; each composite is z-scored within sample.

## References

Stock, J. H., & Watson, M. W. (2018). Identification and Estimation of Dynamic Causal Effects in Macroeconomics Using External Instruments. *Economic Journal*, 128(610), 917-948. doi:[10.1111/eoj.12593](https://doi.org/10.1111/eoj.12593).

Romer, C. D., & Romer, D. H. (2004). A New Measure of Monetary Shocks. *American Economic Review*, 94(4), 1055-1084. doi:[10.1257/0002828042002651](https://doi.org/10.1257/0002828042002651).

## Examples

```
d <- load_paper_data()
ch <- build_channel_composites(d$proxies)
head(ch); cor(ch[, -1])
```

---

build_lag	<i>Build a Lagged Vector with Leading NAs</i>
-----------	---

---

**Description**

Returns a vector  $x_{t-\ell}$  where the first  $\ell$  positions are NA and the remaining positions are the lagged values of  $x$ . Used throughout the package for instrument and predictor construction.

**Usage**

```
build_lag(x, lg)
```

**Arguments**

<code>x</code>	Numeric vector.
<code>lg</code>	Positive integer lag.

**Value**

Numeric vector of the same length as  $x$ .

**Examples**

```
build_lag(1:10, 3)
```

---

build_network	<i>Construct a Directed Contagion Network from a WQTE Flow Matrix</i>
---------------	---

---

**Description**

Builds a directed weighted network from a pairwise WQTE flow matrix by retaining only the edges whose intensity exceeds the supplied threshold. The network construction follows the standard contagion-network literature (Diebold and Yilmaz, 2014; Billio et al., 2012).

**Usage**

```
build_network(F_matrix, threshold)
```

**Arguments**

<code>F_matrix</code>	An $N \times N$ numeric matrix of WQTE values.
<code>threshold</code>	Numeric threshold for edge retention; edges with $F\_matrix[i, j] > threshold$ are retained.

**Value**

An igraph object with weight edge attribute equal to the WQTE value.

**Examples**

```
m <- matrix(runif(25, 0, 0.1), 5, 5); diag(m) <- 0
g <- build_network(m, threshold = 0.05)
igraph::ecount(g)
```

---

channel_proxies	<i>Channel-Proxy Daily Series</i>
-----------------	-----------------------------------

---

**Description**

Daily-frequency channel-proxy series aligned on the G20 trading-day grid, used as inputs to [build\\_channel\\_composites](#). Values are forward-filled where lower-frequency data are available.

**Usage**

```
channel_proxies
```

**Format**

A data.frame with 5,036 rows and 14 columns:

**Date** Trading date

**vix** CBOE Volatility Index level

**hy\_spread** ICE BofA US High-Yield Option-Adjusted Spread

**stress\_index** St. Louis Fed Financial Stress Index (STLFSI4)

**usd\_index** Federal Reserve Broad Trade-Weighted Dollar Index (DTWEXBGS)

**gpr** Caldara-Iacoviello Geopolitical Risk Index

**geo\_events** Geopolitical-events indicator

**vix\_slope** VIX-VIX3M term-structure slope

**fear\_proxy** Daily absolute change in VIX

**sentiment** University of Michigan Consumer Sentiment Index

**fed\_rate** Effective Federal Funds Rate

**dgs10** 10-year Treasury yield

**term\_spread** 10-year minus 3-month Treasury yield spread

**qe\_dummy** Quantitative-easing program indicator (binary)

**Source**

Federal Reserve Economic Data (FRED), Yahoo Finance, Caldara-Iacoviello GPR data library.

## Examples

```
data(channel_proxies)
summary(channel_proxies[, c("vix", "gpr", "fed_rate")])
```

---

cinelli\_hazlett\_rv      *Cinelli-Hazlett Robustness Value*

---

## Description

Computes the partial- $R^2$  that an unobserved confounder would need to share with both the treatment and the outcome to drive the structural coefficient to zero. The robustness value is bounded in  $[0, 1]$ ; values near zero indicate that even a weakly correlated confounder could explain away the result, while values near one indicate identification-robust findings.

## Usage

```
cinelli_hazlett_rv(theta, se, df)
```

## Arguments

theta	Estimated structural coefficient.
se	Standard error of theta.
df	Residual degrees of freedom.

## Value

A scalar in  $[0, 1]$ , or NA if inputs are invalid.

## References

Cinelli, C., & Hazlett, C. (2020). Making Sense of Sensitivity: Extending Omitted Variable Bias. *Journal of the Royal Statistical Society Series B*, 82(1), 39-67. doi:10.1111/rssb.12348.

## Examples

```
cinelli_hazlett_rv(theta = 0.4, se = 0.1, df = 200)
```

---

compute\_wqte\_matrix     *Wavelet-Quantile Transfer Entropy Matrix*

---

### Description

Computes the bilateral WQTE matrix for a returns panel at one wavelet scale and one quantile, producing the directed flow matrix that serves as the Stage-1 input to the structural-attribution layer.

### Usage

```
compute_wqte_matrix(returns, scale = 5, tau = 0.5, n_cores = 2L)
```

### Arguments

returns	An xts or matrix of returns (rows = time, cols = markets).
scale	Integer wavelet scale. Default 5.
tau	Quantile level. Default 0.50.
n_cores	Integer; number of parallel cores for mclapply. Default 2L per CRAN policy; on Windows the function falls back to serial lapply. Increase for production-scale workloads.

### Value

An  $N \times N$  matrix where entry  $(i, j)$  is the WQTE from market  $i$  to market  $j$  at the specified scale and quantile; row and column names are taken from `colnames(returns)`.

### References

Bhandari, A., & Parida, I. (2026). Wavelet-quantile transfer entropy for financial-market contagion.

### Examples

```
d <- load_paper_data()
ix <- which(zoo::index(d$returns) >= as.Date("2008-01-01") &
           zoo::index(d$returns) <= as.Date("2008-12-31"))
F <- compute_wqte_matrix(d$returns[ix, ], scale = 5, tau = 0.50, n_cores = 2)
```

---

crisis_periods	<i>Crisis Sub-Period Definitions</i>
----------------	--------------------------------------

---

### Description

A named list of length-two character vectors specifying the start and end dates of the eight crisis sub-periods analysed in the paper.

### Usage

```
crisis_periods
```

### Format

A named list with eight elements:

**PreCrisis** Pre-Crisis Baseline (12 Jan 2006 - 31 Jul 2007)

**GFC** Global Financial Crisis (1 Aug 2007 - 30 Jun 2009)

**ESDC** European Sovereign Debt Crisis (1 Dec 2009 - 30 Jun 2012)

**CSC** Chinese Stock Crash (15 Jun 2015 - 31 Dec 2016)

**PreCOVID** Pre-COVID interval (1 Jan 2017 - 31 Jan 2020)

**COVID** COVID-19 Pandemic (1 Feb 2020 - 31 Dec 2021)

**RusUkr** Russia-Ukraine episode (1 Feb 2022 - 31 Dec 2023)

**MidEastTariffs** Middle-East tensions and tariffs (1 Jan 2024 - 18 Mar 2026)

### Examples

```
data(crisis_periods)
crisis_periods$GFC
```

---

g20_returns	<i>G20 Equity-Market Daily Log&gt;Returns</i>
-------------	---

---

### Description

Daily log-return panel for 18 G20 equity-market indices spanning 12 January 2006 through 18 March 2026, used in the paper replication. The panel covers 5,036 trading days. Eight developed and ten emerging markets are represented.

### Usage

```
g20_returns
```

**Format**

An xts object with 5,036 rows and 18 columns:

**Argentina** S&P/MERVAL log-returns  
**Australia** S&P/ASX 200 log-returns  
**Brazil** IBOVESPA log-returns  
**Canada** S&P/TSX Composite log-returns  
**China** Shanghai Composite log-returns  
**France** CAC 40 log-returns  
**Germany** DAX 40 log-returns  
**India** BSE SENSEX log-returns  
**Indonesia** IDX Composite log-returns  
**Italy** FTSE MIB log-returns  
**Japan** Nikkei 225 log-returns  
**Mexico** S&P/BMV IPC log-returns  
**Russia** IMOEX log-returns  
**SouthAfrica** JSE All Share log-returns  
**SouthKorea** KOSPI log-returns  
**Turkey** BIST 100 log-returns  
**UK** FTSE 100 log-returns  
**USA** S&P 500 log-returns

**Source**

Yahoo Finance and Investing.com aggregator.

**Examples**

```
data(g20_returns)
dim(g20_returns); head(g20_returns[, 1:5])
```

---

iv\_2sls\_attribute

*Five-Channel IV/2SLS Channel Attribution for One Bilateral Link*


---

**Description**

Estimates the structural equation

$$C_{ij,t} = \alpha + \sum_{c=1}^5 \theta_c \text{Channel}_{c,t} + \gamma_1 f_t + \gamma_2 C_{ij,t-1} + \varepsilon_{ij,t}$$

via two-stage least squares, treating the five channel composites as endogenous and instrumenting them with their own lagged values at  $t-5$ ,  $t-10$ ,  $t-15$  plus cross-channel interactions at  $t-5$ . The first-stage F-statistic per channel, the Sargan over-identification J-test, and the Durbin-Wu-Hausman endogeneity test are reported.

**Usage**

```
iv_2sls_attribute(C_ij, ch_per, R_full, channel_names = CHN_DEFAULT)
```

**Arguments**

`C_ij` Numeric vector of pairwise daily co-movement ( $r_i * r_j$ ).

`ch_per` A data.frame containing the five channel composites for the current period (with columns named per `channel_names`).

`R_full` Numeric matrix of returns for the period (used to construct the global factor  $f_t$ ).

`channel_names` Character vector of channel column names. Default `c("Trade", "Financial", "Geopolitical", "Behavioral")`.

**Value**

A list with elements `theta` (5-vector of structural coefficients), `partial_F` (per-channel first-stage F-stats), `J_stat`, `J_p`, `dwh_F`, `dwh_p`, and `n_obs`; or NULL if the regression cannot be run.

**References**

Stock, J. H., & Watson, M. W. (2018). Identification and Estimation of Dynamic Causal Effects in Macroeconomics Using External Instruments. *Economic Journal*, 128(610), 917-948. doi:10.1111/eoj.12593.

Mertens, K., & Ravn, M. O. (2013). The Dynamic Effects of Personal and Corporate Income Tax Changes in the United States. *American Economic Review*, 103(4), 1212-1247. doi:10.1257/aer.103.4.1212.

**Examples**

```
d <- load_paper_data()
ch <- build_channel_composites(d$proxies)
p <- period_subset(d$returns, ch, d$periods$GFC)
src <- as.numeric(p$R[, "USA"]); tgt <- as.numeric(p$R[, "SouthKorea"])
fit <- iv_2sls_attribute(src * tgt, p$C, p$R)
fit$theta
```

---

lasso\_iv\_attribute     *LASSO-Based Instrument Selection IV Attribution (Belloni-Chernozhukov-Hansen)*

---

**Description**

Estimates the channel-attribution coefficients via post-double-selection LASSO IV with the high-dimensional instrument set. Uses `hdm::rlassoIV` per channel with controls for the other endogenous regressors. Requires the optional **hdm** package.

**Usage**

```
lasso_iv_attribute(C_ij, ch_per, R_full, channel_names = CHN_DEFAULT)
```

**Arguments**

`C_ij` Numeric vector of pairwise daily co-movement ( $r_i * r_j$ ).

`ch_per` A data.frame containing the five channel composites for the current period (with columns named per `channel_names`).

`R_full` Numeric matrix of returns for the period (used to construct the global factor  $f_t$ ).

`channel_names` Character vector of channel column names. Default `c("Trade", "Financial", "Geopolitical", "Behav`

**Value**

A list with elements `theta` and `se`, or `NULL`.

**References**

Belloni, A., Chernozhukov, V., & Hansen, C. (2014). Inference on Treatment Effects after Selection among High-Dimensional Controls. *Review of Economic Studies*, 81(2), 608-650. doi:10.1093/restud/rdt044.

**Examples**

```
if (requireNamespace("hdm", quietly = TRUE)) {
  d <- load_paper_data()
  ch <- build_channel_composites(d$proxies)
  p <- period_subset(d$returns, ch, d$periods$GFC)
  src <- as.numeric(p$R[, "USA"]); tgt <- as.numeric(p$R[, "SouthKorea"])
  fit <- lasso_iv_attribute(src * tgt, p$c, p$R)
}
```

---

load_paper_data	<i>Load the Paper's Bundled Data</i>
-----------------	--------------------------------------

---

**Description**

Convenience loader returning a named list with the three bundled datasets (returns, channel proxies, crisis periods) used in the paper replication.

**Usage**

```
load_paper_data()
```

**Value**

A named list with elements `returns`, `proxies`, and `periods`.

**Examples**

```
d <- load_paper_data(); str(d, max.level = 1)
```

---

local_projections	<i>Local-Projection Channel Attribution at Multiple Horizons</i>
-------------------	--

---

**Description**

Estimates horizon-specific impulse responses of the pairwise co-movement to each channel composite at horizons  $h \in \{1, 5, 22\}$  days following Jorda (2005). The local projection at horizon  $h$  estimates

$$C_{ij,t+h} = \alpha_h + \beta_{c,h} \text{Channel}_{c,t} + \text{controls} + u_{ij,t+h}$$

separately for each channel  $c$ , with the other four channels and the lagged co-movement and global factor entering as controls.

**Usage**

```
local_projections(
  C_ij,
  ch_per,
  R_full,
  horizons = c(1, 5, 22),
  channel_names = CHN_DEFAULT
)
```

**Arguments**

C_ij	Numeric vector of pairwise daily co-movement ( $r_i * r_j$ ).
ch_per	A data.frame containing the five channel composites for the current period (with columns named per channel_names).
R_full	Numeric matrix of returns for the period (used to construct the global factor $f_t$ ).
horizons	Integer vector of horizons. Default c(1, 5, 22).
channel_names	Character vector of channel column names. Default c("Trade", "Financial", "Geopolitical", "Behav

**Value**

A list with one element per horizon; each element is a length-length(channel\_names) numeric vector of LP coefficients.

**References**

Jorda, O. (2005). Estimation and Inference of Impulse Responses by Local Projections. *American Economic Review*, 95(1), 161-182. doi:10.1257/0002828053828518.

Plagborg-Moller, M., & Wolf, C. K. (2021). Local Projections and VARs Estimate the Same Impulse Responses. *Econometrica*, 89(2), 955-980. doi:10.3982/ECTA17813.

**Examples**

```
d <- load_paper_data()
ch <- build_channel_composites(d$proxies)
p <- period_subset(d$returns, ch, d$periods$GFC)
src <- as.numeric(p$R[, "USA"]); tgt <- as.numeric(p$R[, "SouthKorea"])
lp <- local_projections(src * tgt, p$C, p$R)
```

---

modwt\_detail

---

*MODWT Wavelet Detail at a Specified Scale*


---

**Description**

Returns the MODWT detail coefficients of a return series at the specified scale, using the Daubechies least-asymmetric filter of length 8 (LA8). The maximal-overlap discrete wavelet transform is shift-invariant and aligned with the original time-axis, making it suited to financial returns; see Percival and Walden (2000).

**Usage**

```
modwt_detail(x, scale = 5, J = 6, filter = "la8")
```

**Arguments**

x	Numeric vector of returns.
scale	Integer scale (1-6) corresponding to dyadic horizons of $[2^s, 2^{s+1}]$ trading days. Default 5 (32-64 day horizon).
J	Integer maximum decomposition level. Default 6.
filter	Character; the wavelet filter family to use. Default "la8" for the LA8 filter. See <a href="#">modwt</a> .

**Value**

Numeric vector of detail coefficients at scale scale.

**References**

Percival, D. B., & Walden, A. T. (2000). *Wavelet Methods for Time Series Analysis*. Cambridge University Press.

**Examples**

```
x <- rnorm(512)
d5 <- modwt_detail(x, scale = 5)
length(d5)
```

---

network_summary	<i>Summary Statistics of a Contagion Network</i>
-----------------	--

---

**Description**

Returns a list of standard centrality and density statistics for a directed contagion network.

**Usage**

```
network_summary(g)
```

**Arguments**

`g` An igraph object.

**Value**

A list with elements `n_edges`, `density`, `in_degree`, `out_degree`, `betweenness`, and `eigenvector`.

**Examples**

```
m <- matrix(runif(25, 0, 0.1), 5, 5); diag(m) <- 0
g <- build_network(m, threshold = 0.05)
network_summary(g)$density
```

---

orthogonalise_residual	
------------------------	--

*Orthogonalise One Series Against Another*

---

**Description**

Returns the residuals from a regression of  $y$  on  $x$ ; the residual is by construction orthogonal to  $x$  in the sample. This is the pre-processing step used to construct the behavioural channel composite, which is orthogonalised against the financial composite to avoid the within-VIX decomposition that contaminates cross-channel identification when both composites share VIX-derivative inputs.

**Usage**

```
orthogonalise_residual(y, x)
```

**Arguments**

`y` Numeric vector to be orthogonalised.  
`x` Numeric vector against which  $y$  is orthogonalised.

**Value**

Numeric vector of length `length(y)` containing the residuals.

**Examples**

```
a <- rnorm(100); b <- 0.5 * a + rnorm(100); cor(a, orthogonalise_residual(b, a))
```

---

period_subset	<i>Subset a Returns and Channel Panel by Period</i>
---------------	---

---

**Description**

Selects the rows of an xts returns object and the matching rows of a channel-composite data.frame that fall within a date range.

**Usage**

```
period_subset(returns_xts, channels_df, period_dates)
```

**Arguments**

`returns_xts` An xts object of daily returns indexed by Date.  
`channels_df` A data.frame with a Date column matching the index of `returns_xts`.  
`period_dates` A character or Date vector of length 2 `c(start, end)`.

**Value**

A list with elements R (xts subset) and C (data.frame subset).

**Examples**

```
d <- load_paper_data()
ch <- build_channel_composites(d$proxies)
p <- period_subset(d$returns, ch, c("2008-01-01", "2008-12-31"))
nrow(p$C)
```

plot\_attribution\_evolution

*Evolution-of-Shares Line Plot*

---

### Description

Line plot of channel-attribution share evolution across crisis sub-periods.

### Usage

```
plot_attribution_evolution(period_shares)
```

### Arguments

period\_shares Per-period share data.frame (output from [run\\_contagion\\_pipeline](#)).

### Value

A ggplot object.

---

plot\_attribution\_stack

*Stacked Bar Plot of Channel-Attribution Shares*

---

### Description

Draws a stacked bar chart of channel-attribution shares across crisis sub-periods. Returns a ggplot if **ggplot2** is available, else uses base barplot.

### Usage

```
plot_attribution_stack(period_shares, ...)
```

### Arguments

period\_shares A data.frame with columns Period, Trade, Financial, Geopolitical, Behavioral, Monetary.

... Additional arguments passed to ggplot2::ggsave or to barplot.

### Value

A ggplot object (or invisibly the matrix used for base plotting).

**Examples**

```
d <- load_paper_data()
ch <- build_channel_composites(d$proxies)
res <- run_contagion_pipeline(d$returns, ch, d$periods, n_cores = 2)
plot_attribution_stack(res$period_shares)
```

---

plot\_qte\_intensity      *Two-Panel QTE Intensity Plot*

---

**Description**

Top panel: mean QTE per sub-period. Bottom panel: network density per sub-period.

**Usage**

```
plot_qte_intensity(stage1_summary)
```

**Arguments**

stage1\_summary    A data.frame with columns Period, MeanQTE, Density.

**Value**

A patchwork ggplot if available, else a multi-panel base plot.

---

plot\_robustness\_value    *Robustness-Value Heatmap*

---

**Description**

Heatmap of Cinelli-Hazlett robustness values per channel and sub-period.

**Usage**

```
plot_robustness_value(rv_matrix)
```

**Arguments**

rv\_matrix          Numeric matrix with rownames = periods, colnames = channels.

**Value**

A ggplot object.

---

`qte_pair`*Pairwise Wavelet-Quantile Transfer Entropy*

---

### Description

Estimates the directional information flow from one wavelet-detail series  $x$  to another  $y$  at the specified quantile, following the quantile-regression-based transfer-entropy estimator. A positive value indicates that conditioning on the past of  $x$  improves the conditional-quantile prediction of  $y$  beyond what  $y$ 's own past supplies.

### Usage

```
qte_pair(x, y, tau = 0.5)
```

### Arguments

<code>x</code>	Numeric source series (typically a MODWT detail coefficient).
<code>y</code>	Numeric target series (typically a MODWT detail coefficient).
<code>tau</code>	Quantile level in (0,1). Default 0.50 (median).

### Value

A scalar; NA if there are insufficient observations or the quantile regressions fail to converge.

### References

Schreiber, T. (2000). Measuring Information Transfer. *Physical Review Letters*, 85(2), 461. [doi:10.1103/PhysRevLett.85.461](https://doi.org/10.1103/PhysRevLett.85.461).

Han, H., Linton, O., Oka, T., & Whang, Y.-J. (2016). The Cross-Quantilogram. *Journal of Econometrics*, 193(1), 251-270. [doi:10.1016/j.jeconom.2016.03.001](https://doi.org/10.1016/j.jeconom.2016.03.001).

### Examples

```
x <- rnorm(500); y <- 0.3 * c(0, x[-500]) + rnorm(500)
qte_pair(x, y, tau = 0.5)
```

---

rigobon_id	<i>Heteroskedasticity-Based Identification (Rigobon 2003)</i>
------------	---

---

### Description

Identifies the channel-attribution coefficients by exploiting regime shifts in the variance of returns within the period. Useful when the Sargan over-identification test rejects the joint validity of external instruments and an alternative identification strategy is required.

### Usage

```
rigobon_id(C_ij, ch_per, R_full, channel_names = CHN_DEFAULT)
```

### Arguments

C_ij	Numeric vector of pairwise daily co-movement ( $r_i * r_j$ ).
ch_per	A data.frame containing the five channel composites for the current period (with columns named per channel_names).
R_full	Numeric matrix of returns for the period (used to construct the global factor $f_t$ ).
channel_names	Character vector of channel column names. Default <code>c("Trade", "Financial", "Geopolitical", "Behav</code>

### Value

A list with element theta: a length-five numeric vector of structural coefficients, or NULL.

### References

Rigobon, R. (2003). Identification through Heteroskedasticity. *Review of Economics and Statistics*, 85(4), 777-792. doi:10.1162/003465303772815727.

### Examples

```
d <- load_paper_data()
ch <- build_channel_composites(d$proxies)
p <- period_subset(d$returns, ch, d$periods$GFC)
src <- as.numeric(p$R[, "USA"]); tgt <- as.numeric(p$R[, "SouthKorea"])
rig <- rigobon_id(src * tgt, p$C, p$R)
```

---

 run\_contagion\_pipeline

*End-to-End Contagion-Channel Pipeline*


---

## Description

Runs the full two-stage detection-and-attribution pipeline over a specified set of crisis sub-periods: Stage 1 produces a per-period WQTE flow matrix at the supplied scale and quantile, and Stage 2 attributes each significant link to one of five channels via IV/2SLS. Network density, top transmitter/receiver, and per-period channel-attribution shares are returned in a list.

## Usage

```
run_contagion_pipeline(
  returns,
  channels,
  periods,
  scale = 5,
  tau = 0.5,
  threshold_period = names(periods)[1],
  edge_quantile = 0.75,
  n_cores = 2L
)
```

## Arguments

returns	An xts of daily returns indexed by date.
channels	A data.frame of channel composites (output of <a href="#">build_channel_composites</a> ).
periods	A named list of length-2 character or Date vectors specifying c(start_date, end_date) for each sub-period.
scale	Integer wavelet scale. Default 5.
tau	Quantile level. Default 0.50.
threshold_period	Name of the period in periods from which the Pre-Crisis-baseline absolute threshold is computed. Default names(periods)[1].
edge_quantile	Numeric in (0,1); the quantile of positive WQTE in the threshold period used as the absolute threshold. Default 0.75.
n_cores	Integer number of parallel cores. Default 2L per CRAN policy; raise this for production-scale runs.

## Value

A list with elements

**stage1** Per-period list with F (flow matrix), network, and summary.

**stage2** Per-period list of attribution data.frames with one row per significant link.

**period\_shares** Per-period mean attribution-share data.frame.

**threshold** The absolute WQTE threshold used.

### Examples

```
d <- load_paper_data()
ch <- build_channel_composites(d$proxies)
res <- run_contagion_pipeline(d$returns, ch, d$periods, n_cores = 2)
res$period_shares
```

---

walktrap\_communities *Walktrap Community Detection on a Contagion Network*

---

### Description

Detects communities using the Walktrap algorithm of Pons and Latapy (2006) on the symmetrised version of the directed contagion network.

### Usage

```
walktrap_communities(g)
```

### Arguments

**g** An igraph object.

### Value

An integer vector of community memberships, one per vertex; or NULL if the network has too few edges.

### Examples

```
m <- matrix(runif(25, 0, 0.1), 5, 5); diag(m) <- 0
g <- build_network(m, threshold = 0.02)
walktrap_communities(g)
```

---

`zscore`*Z-score Standardisation*

---

**Description**

Returns the z-score of a numeric vector, robust to NAs and zero-variance input.

**Usage**

```
zscore(x)
```

**Arguments**

`x` Numeric vector.

**Value**

Numeric vector of the same length as `x` with mean zero and unit standard deviation; returns a zero vector if the input has no finite variation.

**Examples**

```
zscore(rnorm(100))
```

# Index

## \* datasets

- channel\_proxies, 5
- crisis\_periods, 8
- g20\_returns, 8

build\_channel\_composites, 3, 5, 20

build\_lag, 4

build\_network, 4

channel\_proxies, 5

cinelli\_hazlett\_rv, 6

compute\_wqte\_matrix, 7

crisis\_periods, 8

g20\_returns, 8

iv\_2sls\_attribute, 9

lasso\_iv\_attribute, 10

load\_paper\_data, 11

local\_projections, 12

modwt, 13

modwt\_detail, 13

network\_summary, 14

orthogonalise\_residual, 14

period\_subset, 15

plot\_attribution\_evolution, 16

plot\_attribution\_stack, 16

plot\_qte\_intensity, 17

plot\_robustness\_value, 17

qte\_pair, 18

rigobon\_id, 19

run\_contagion\_pipeline, 16, 20

walktrap\_communities, 21

zscore, 22