

# Measuring economic uncertainty using news-media textual data

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# Motivation

- Uncertainty affects economic decision-making
- But is unobservable, and existing proxies have drawbacks
- News-media measures promising (Baker, Bloom, Davis, 2015) but need further development and characterisation

# Contribution of this paper

Theory of measurement

Refined methodology

Comparison to stock volatility

Conclusions

# Theory of measurement

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$$U \equiv \frac{\text{\# of articles containing keyphrase(s)}}{\text{total \# of articles in FT}}$$

# What is $U$ measuring?



Intensity of  
the **cognitive  
state** of  
uncertainty

**Propensity to  
express**  
uncertainty in  
natural  
language

**Frequency of  
expressions** of  
uncertainty in  
natural  
language

***$U$  as an unbiased, consistent estimator of  $U^*$***

Articles = independent Bernoulli trials

Success = article expresses uncertainty

$\Pr(\text{Success}) = U^*$

$\rightarrow U \sim$  scaled Poisson binomial with mean  $U^*$

**Noise-to-signal ratio:**

Daily 0.87

Weekly 0.21

Monthly 0.05

Annual 0.01

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# Uncertainty keyphrases: baseline

- {"uncertain", "uncertainty", "uncertainties"}
- Including "uncertainties" boosts signal by 14% without material noise increase
- Measurement error should be modest *a priori*
  - Fairly compact semantic space
  - Semantics stable over time
  - Negation is rare because already self-negated

# Uncertainty keyphrases: extensions?

Added keyphrase	Gain in U	Correlation with baseline (monthly)
unclear*	23%	0.51
not clear*	14%	no data
unpredictabl*	7%	0.59
not sure*	7%	0.47
unsure*	4%	0.45
not certain*	1%	0.38
not predictabl*	0%	0.05
<b>all the above</b>	<b>54%</b>	<b>no data</b>
risk*	238%	no data

**For data availability reasons I stick to “uncertain\*”  
Advantage of a relatively ‘clean’ measure**

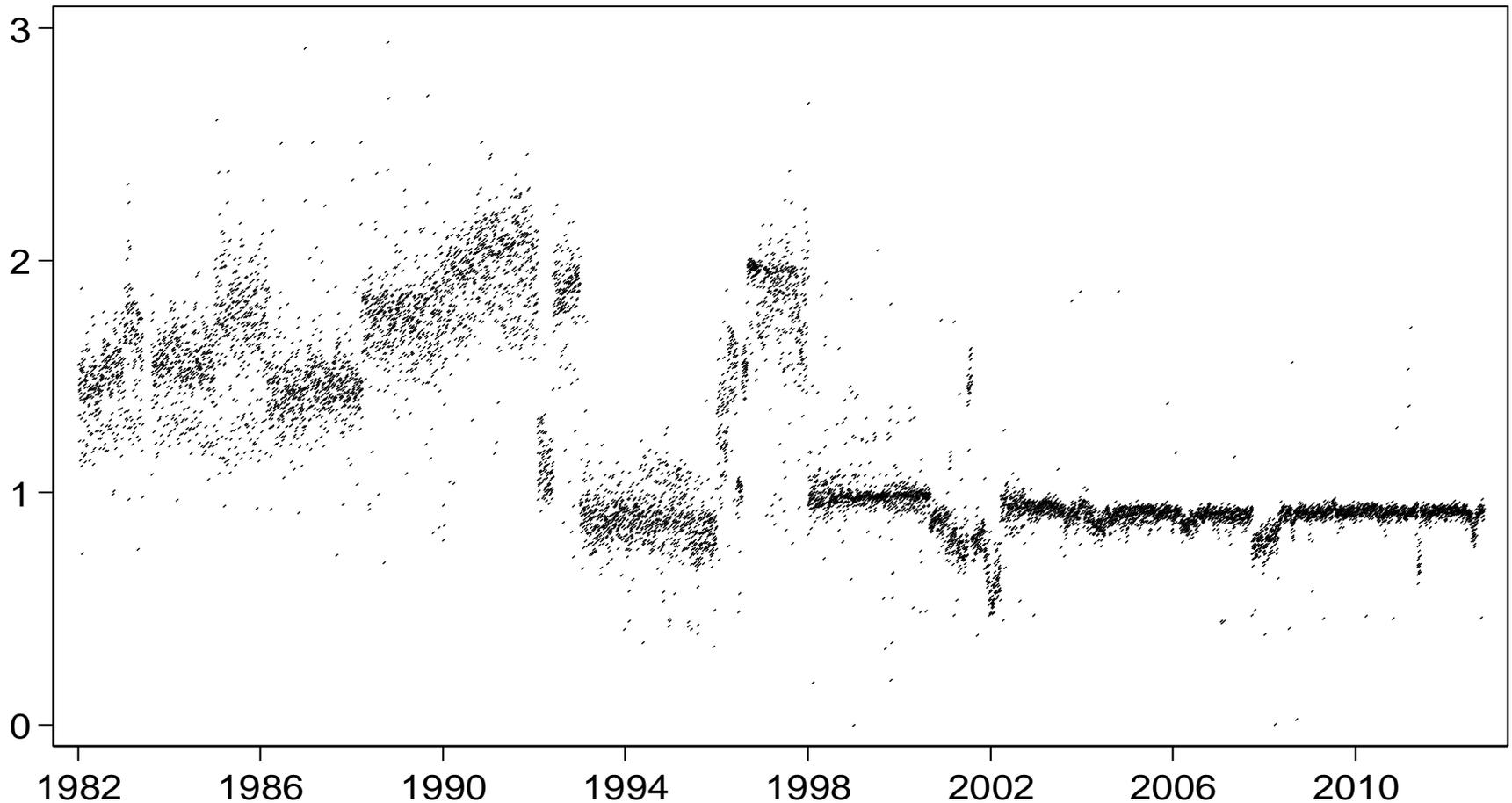
# Isolating economic uncertainty

- Literature using generalist newspapers has to filter out non-economic articles
  - Typically only use articles containing “economic” or “economy”
- Unnecessary with FT
  - Almost all articles have an economic angle
  - May even give stronger signal than subset of articles from multiple generalist papers

% of “uncertain*” articles that do NOT contain “economic” or “economy” but DO contain...	
financ*	66%
bank*	34%
debt*	14%
credit*	15%
bond*	11%
equity   equities	16%
money*	18%
business*	38%
profit*	30%
earnings*	15%
revenue*	11%
wage*	2%
government*	36%
politic*	19%
policy*	13%
Eurozone	2%
oil	10%
gas	6%
coal	2%
<b>any of the above</b>	<b>95%</b>

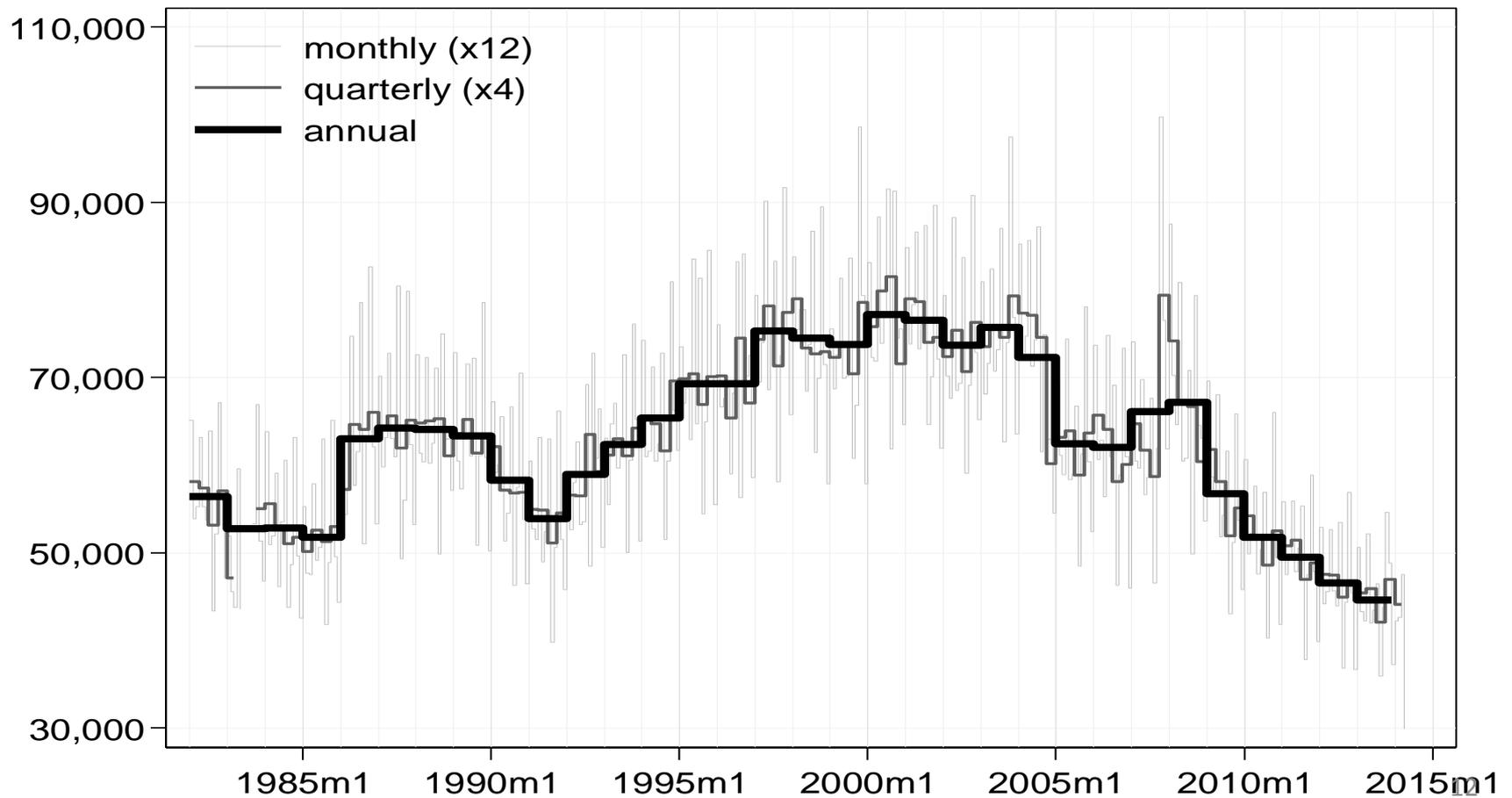
# De-duplication

Ratio of daily FT record counts in Factiva vs. Nexis UK

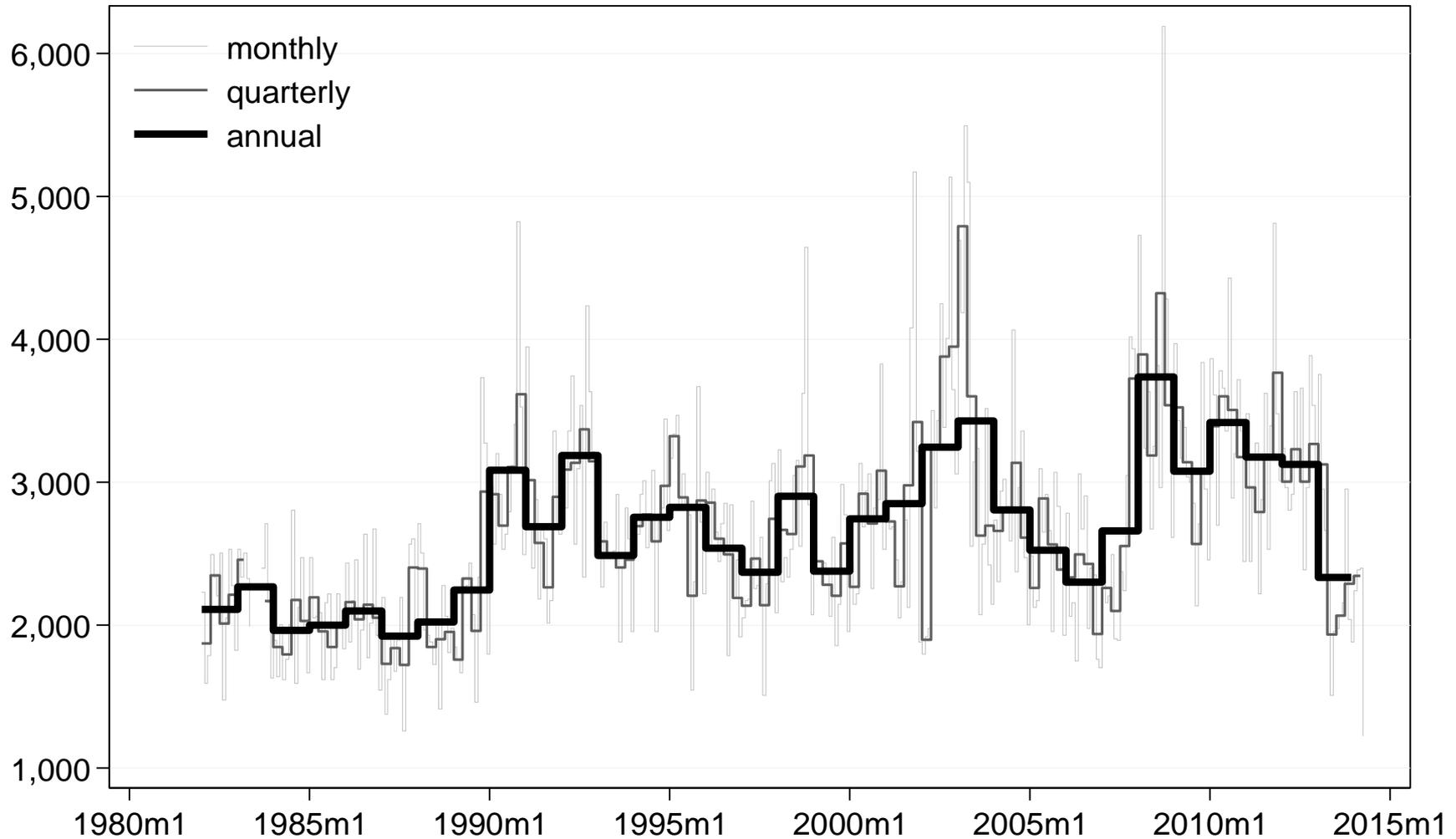


# Normalising by news volume

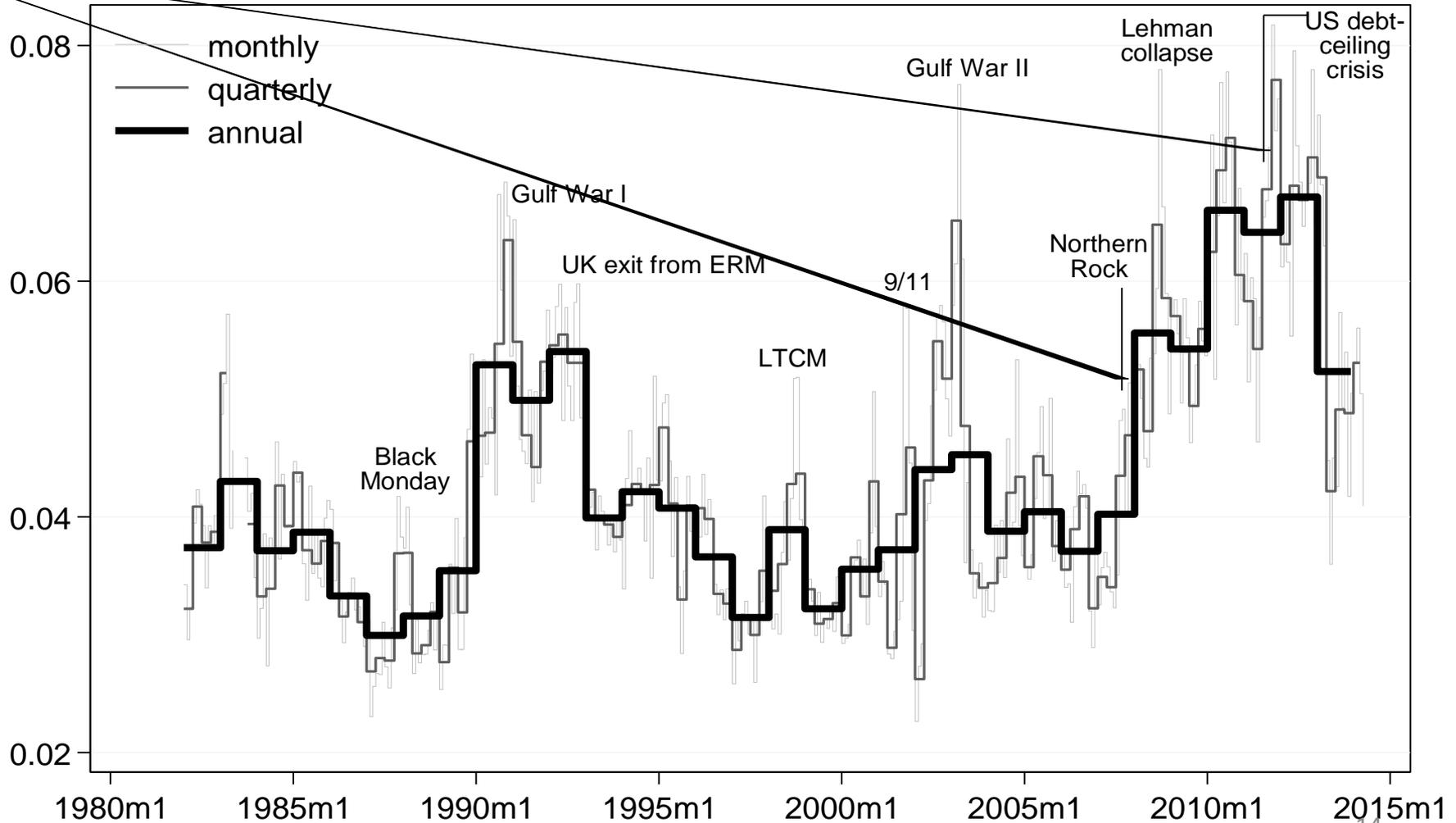
Total # of articles per period



# Without normalisation (numerator of $U$ )

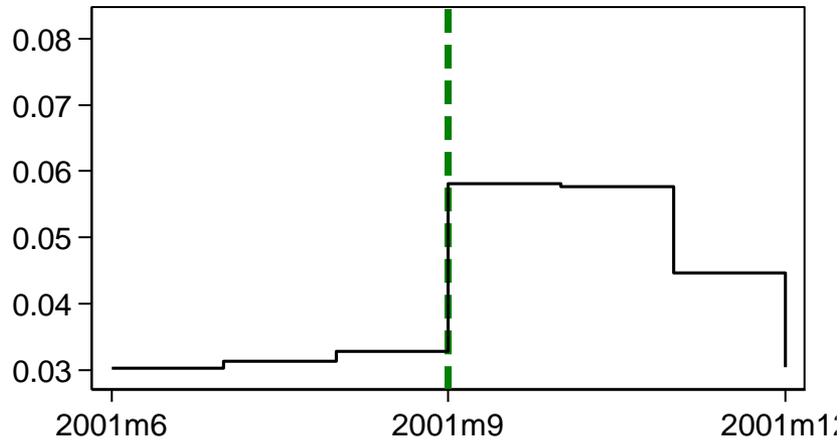


# With normalisation (news-media uncertainty, $U$ )

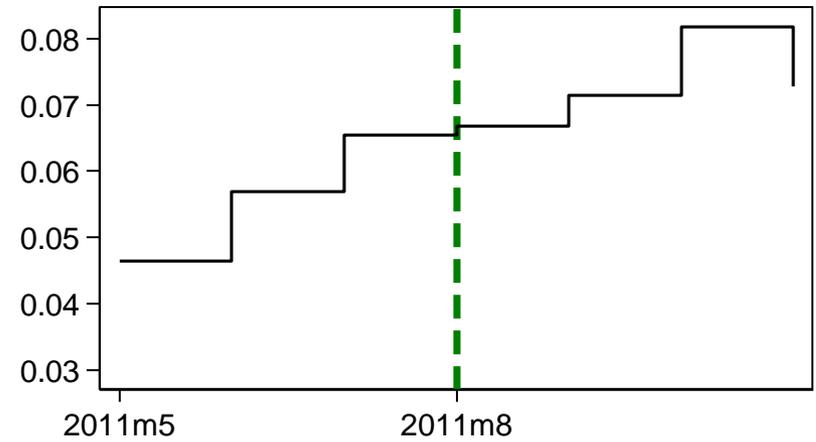


# Moves around major narrative events

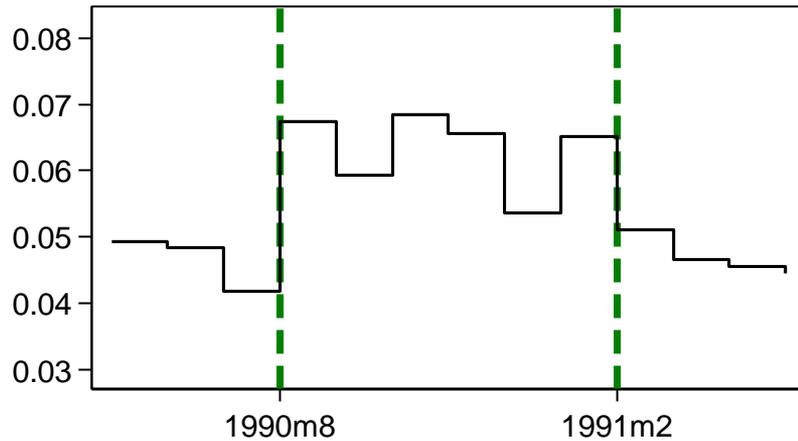
9/11 attacks



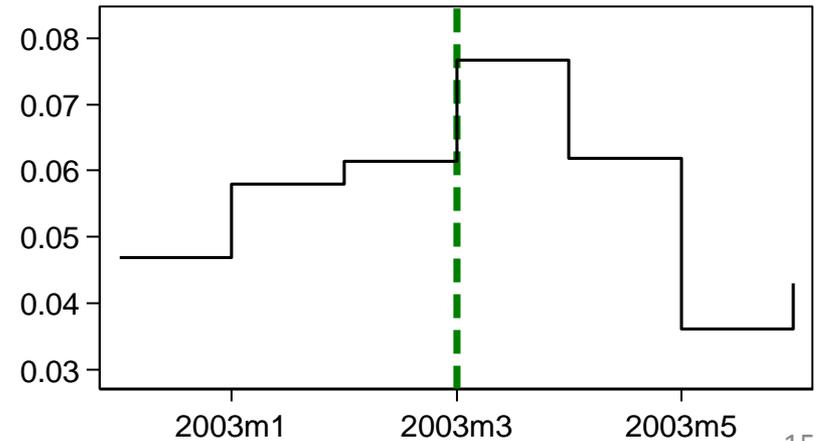
US debt-ceiling crisis



Gulf War I



Gulf War II / Iraq War



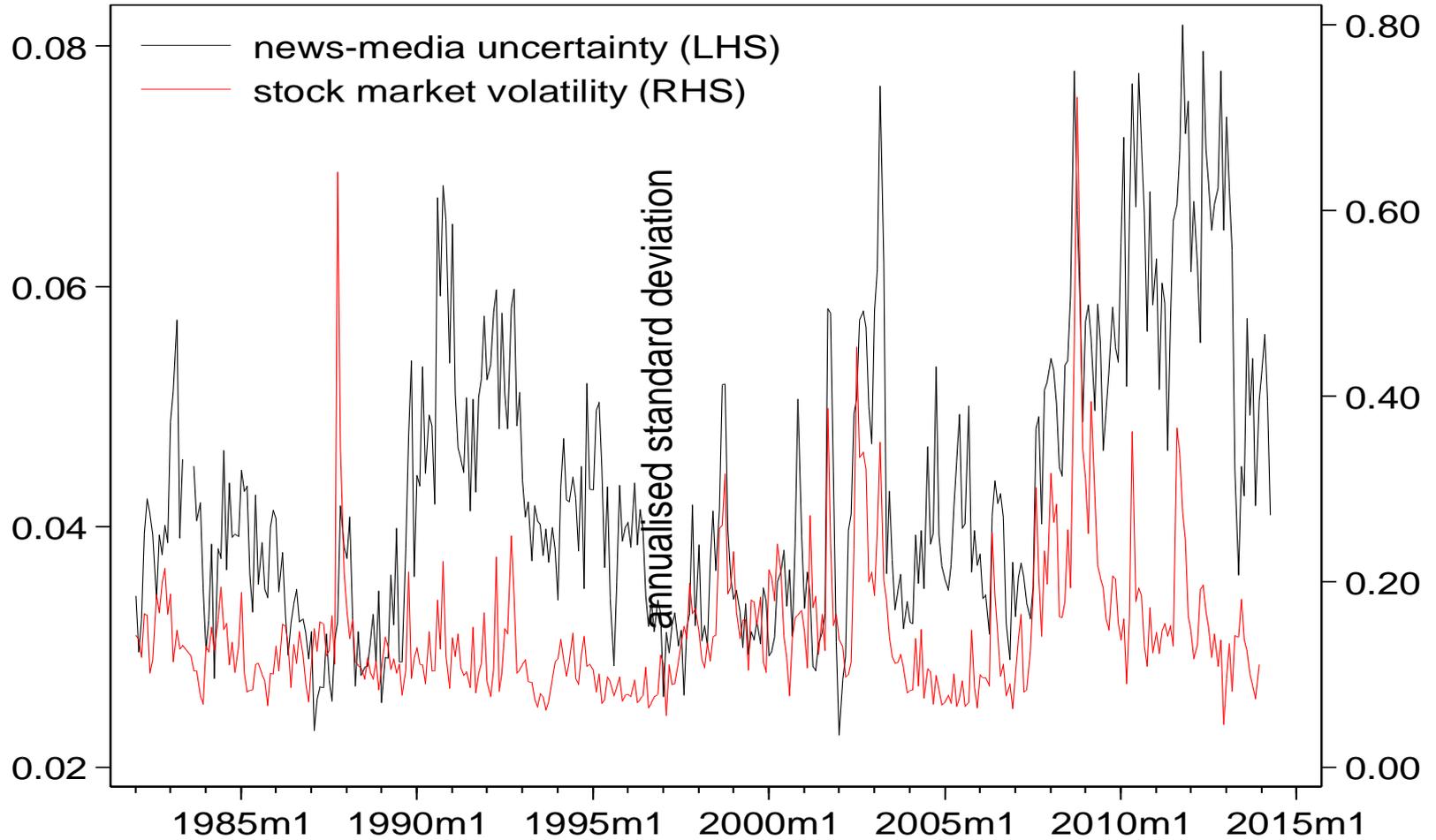
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# $U$ vs. $\sigma$



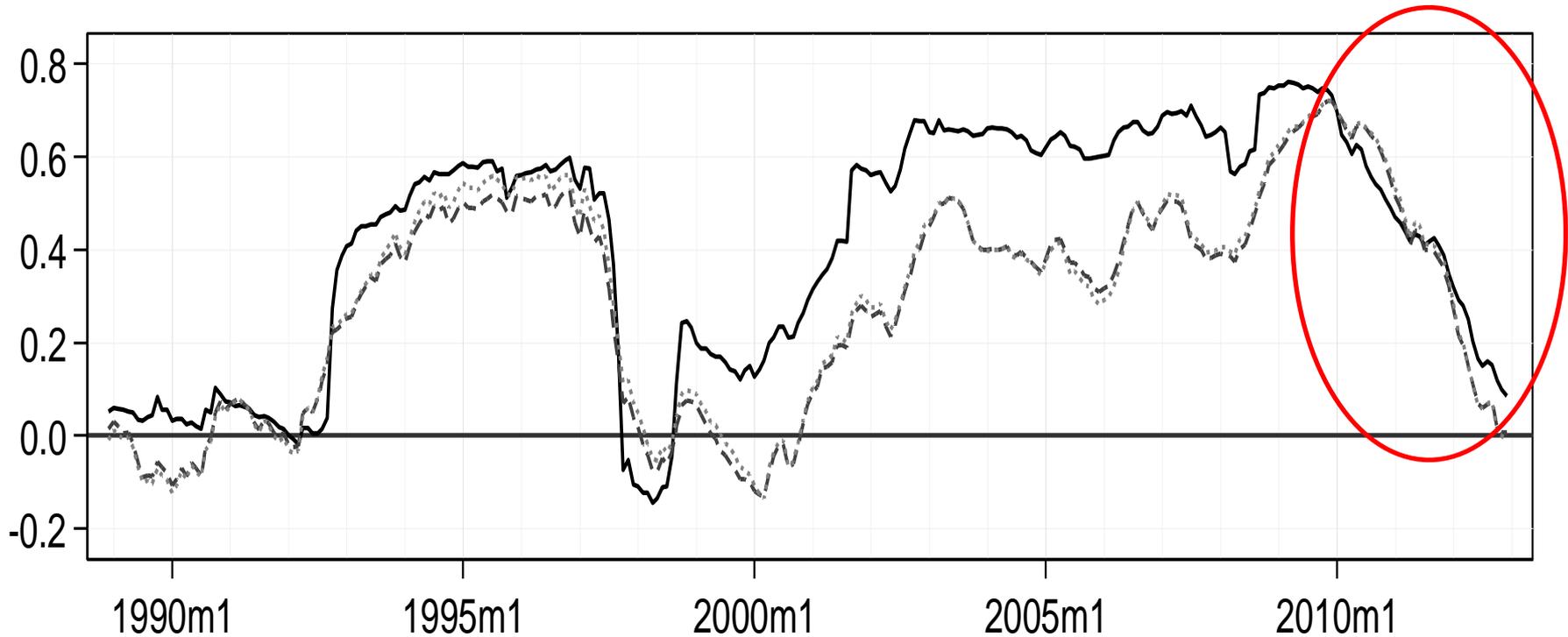
# Granger causality tests

- Bivariate daily VAR with RV estimates\* of  $\sigma$  (2000-2012)
- $U \nrightarrow \sigma$  [p=0.487 on joint exclusion of all lags]
- $\sigma \rightarrow U$  [p=0.010 on joint exclusion of all lags]
- But markets lead by only one day, consistent with FT being published before markets open [p=0.202 for joint exclusion restriction on  $\sigma$  lags 2+]

\* Realized volatility (RV) estimates of Heber, Lunde, Shephard, & Sheppard (2009) based on intra-day tick data

# Correlation of $U$ and $\sigma$

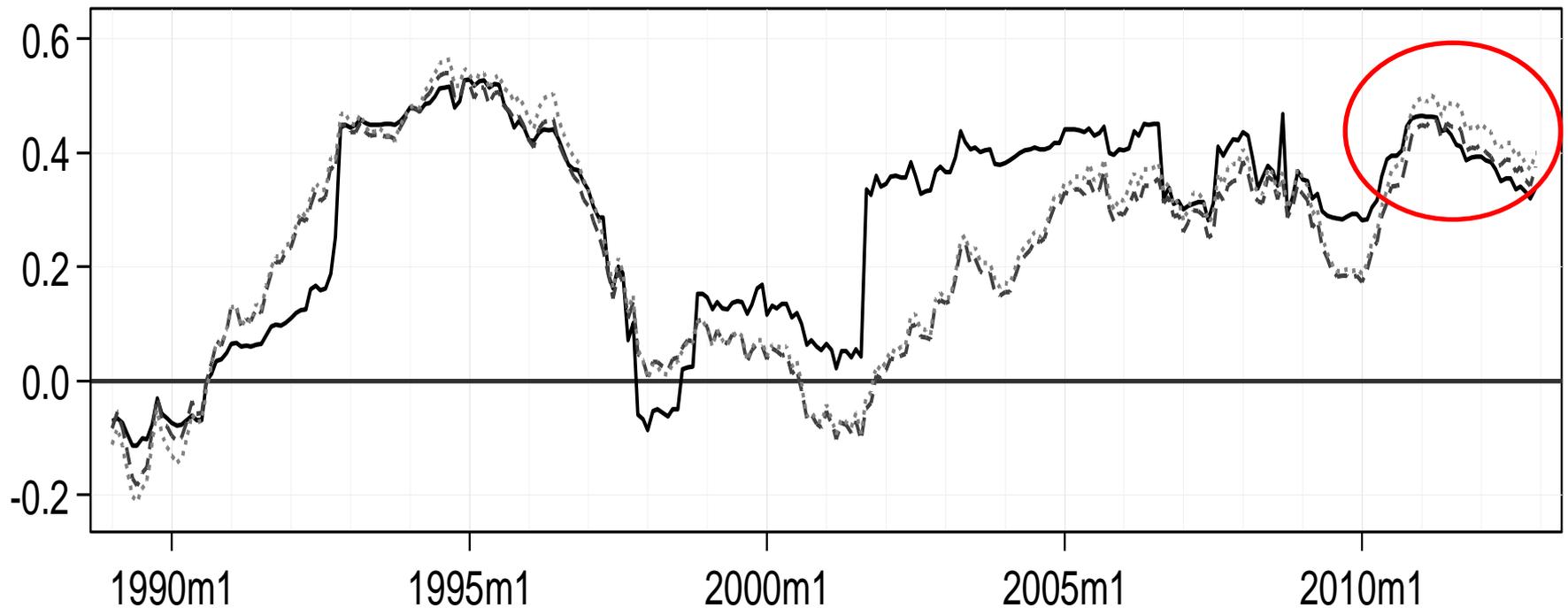
60-month rolling



- Pearson's  $\rho$
- - - Spearman's  $r_s$
- ..... Grenier's  $r_g$

# Correlation of $\Delta U$ and $\Delta\sigma$

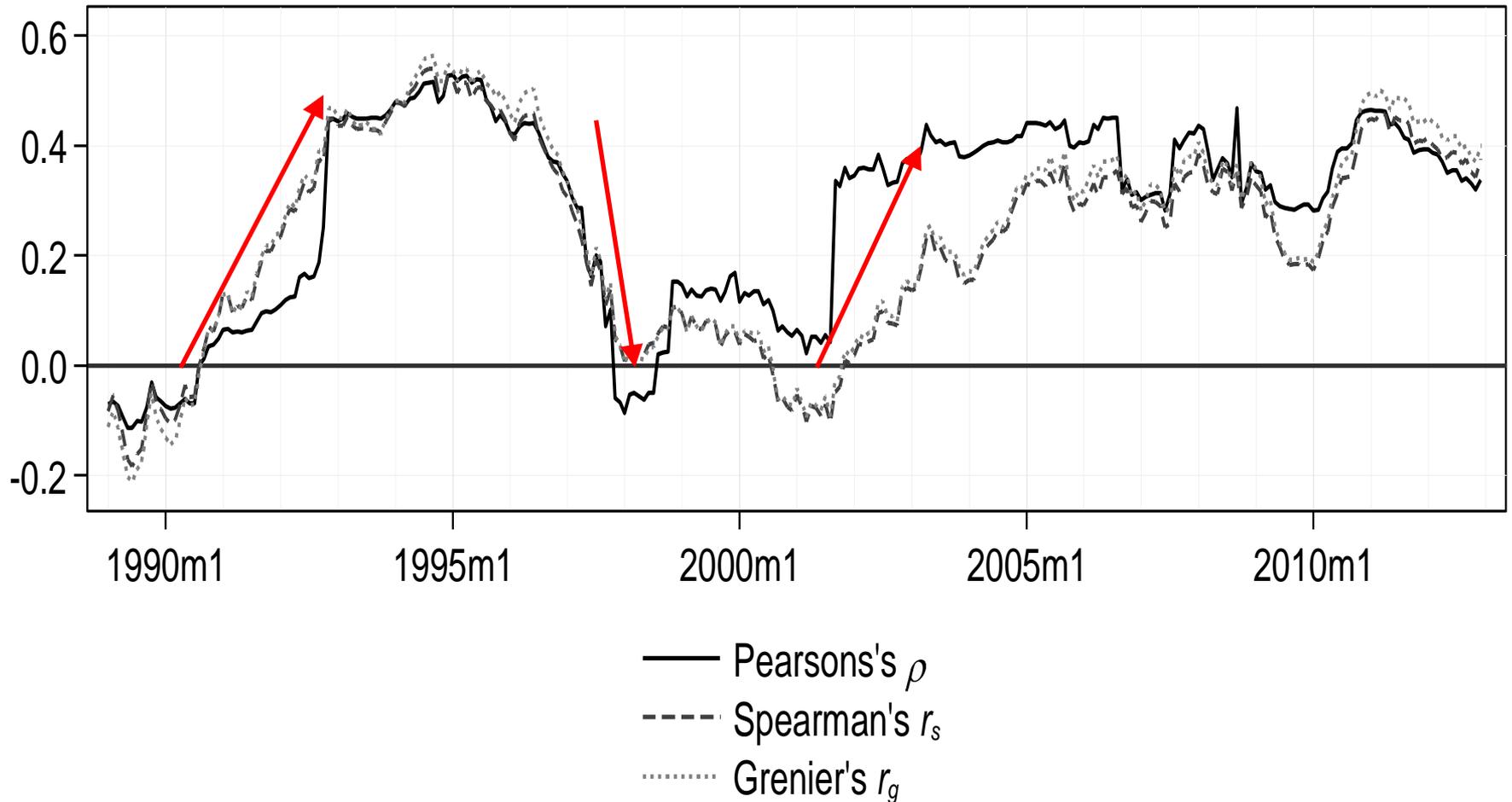
60-month rolling



- Pearson's  $\rho$
- - - Spearman's  $r_s$
- ..... Grenier's  $r_g$

# Switching between high and low correlation

60-month rolling



# Possible explanation: asymmetry in $U$

- Upside risk not naturally expressed as “uncertainty”
- Whereas  $\sigma$  treats upside and downside symmetrically
- Circumstantial evidence consistent with hypothesis
  - Correlation lower when perceived risk more skewed to the upside, e.g. in booms
  - Correlation higher when downside risk is higher, which under our hypothesis will be when  $U$  (or  $|\Delta U|$ ) is larger.

Consistent with this, correlation is 0.331 for  $U$  above its median vs. 0.076 below. Similar result with  $|\Delta U|$ , and not for placebo with  $\sigma$  above vs. below its median.

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# Conclusions

- Theory of measurement clarifies what we are measuring, and how much noise to expect
- Some simple methodology refinements give material reduction in measurement error
- Others are shown to have a modest impact, which gives greater confidence in the apparently simplistic approach of counting small numbers of keywords
- Relationship between  $U$  and  $\sigma$  is time-varying . We proposed an explanation for further testing.

**Questions?**

# Key references

- **Alexopolous & Cohen (2009)**. “Uncertain times, uncertain measures.” *University of Toronto Department of Economics*
- **Baker & Bloom (2013)**. “Measuring Economic Policy Uncertainty.” *www.policyuncertainty.com*
- **Bekaert et al (2009)**. “Risk, uncertainty and asset prices.” *Journal of Financial Economics*
- **Bloom (2014)**. “Fluctuations in Uncertainty.” *Journal of Economic Perspectives*

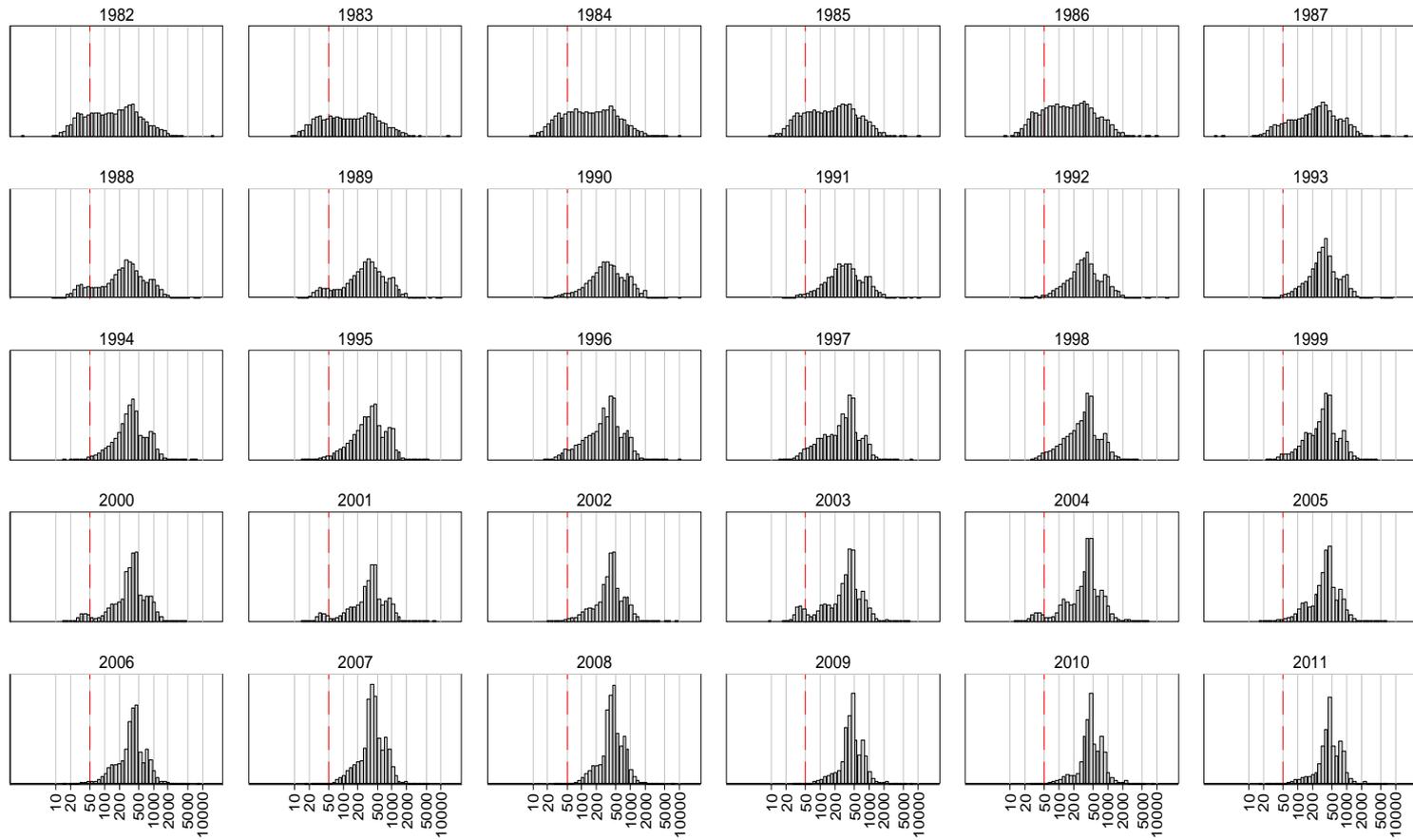
# **APPENDIX**

# **$U$ as an estimator of $U^*$ : sampling noise in finite samples**

- Discretisation grid c. 0.005 at daily frequency (c. 200 articles)
- cf.  $\text{mean}[\underline{U}] \sim 0.04$ ,  $\text{sample var}[U] \sim 0.01-0.02$
- Estimated noise-to-signal ratio (using model-based partition of  $U$ )
  - Daily: 0.871
  - Weekly: 0.205
  - Monthly: 0.054
  - Annual: 0.006

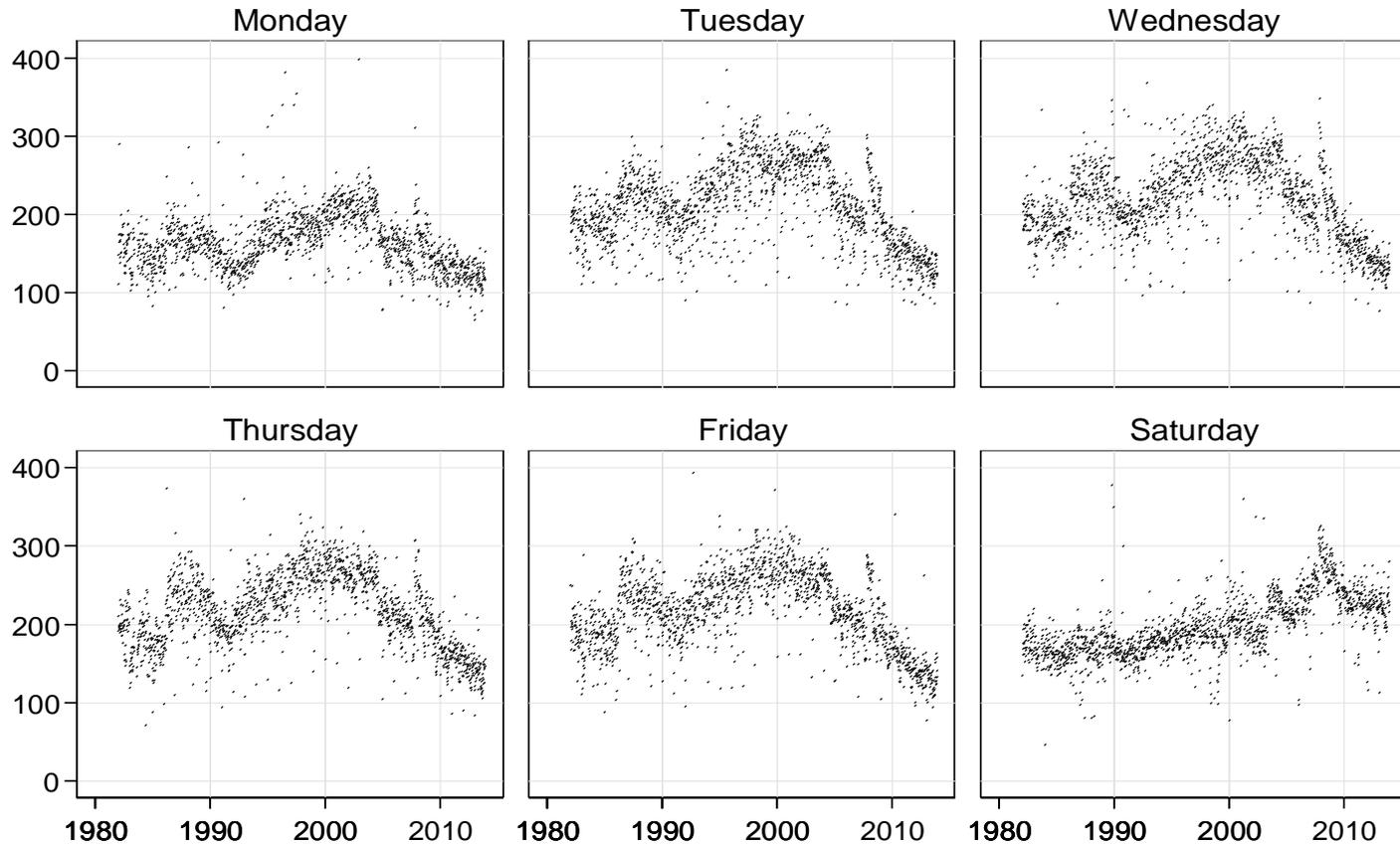
$$\frac{\text{Var}_{U-U^*|n}[U - U^* | n]}{\text{Var}_{U^*|n}[U^* | n]}$$

# Article length: secular variation in the distribution



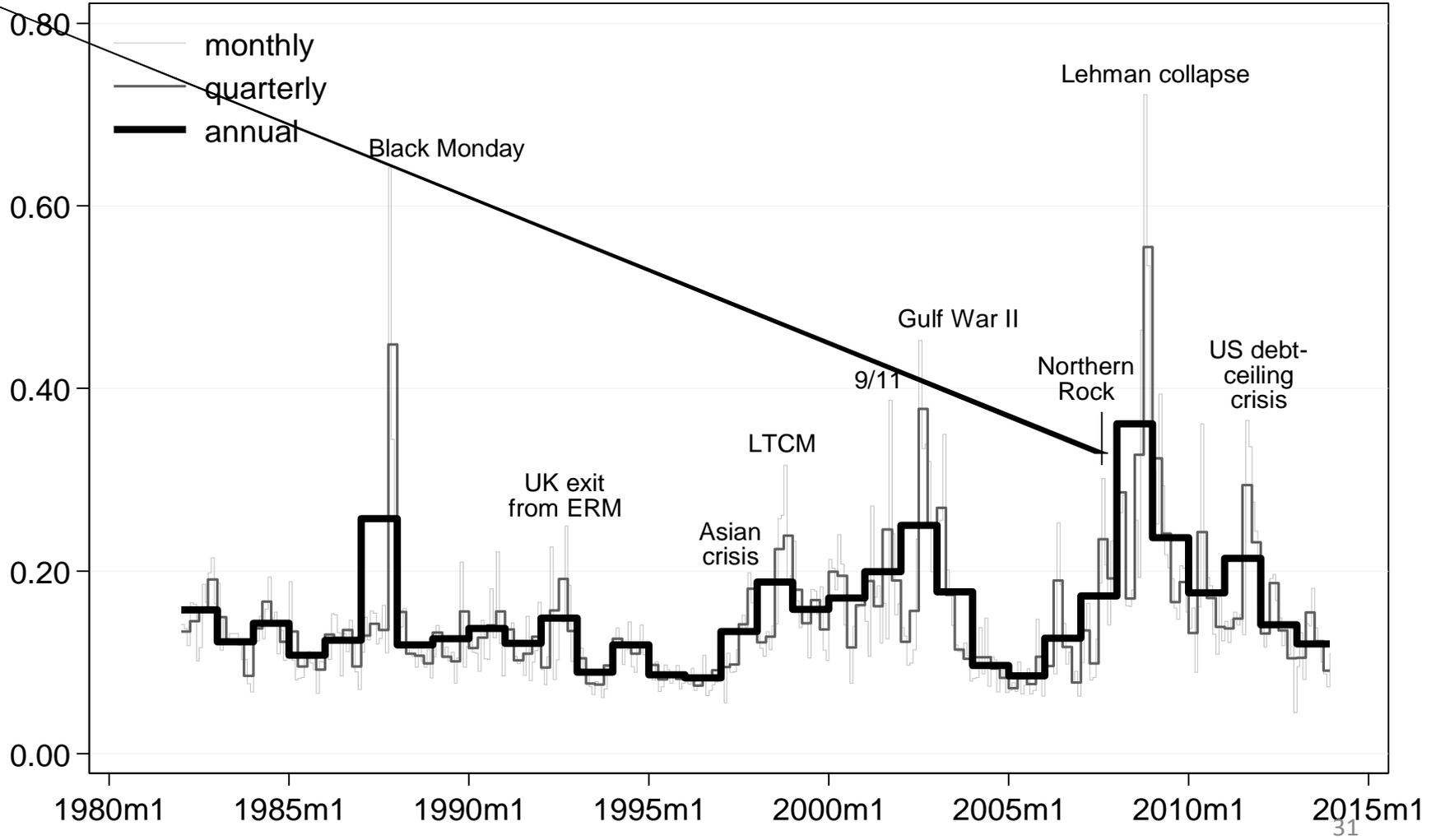
article word count (log scale)

# News volumes: day-of-week seasonality

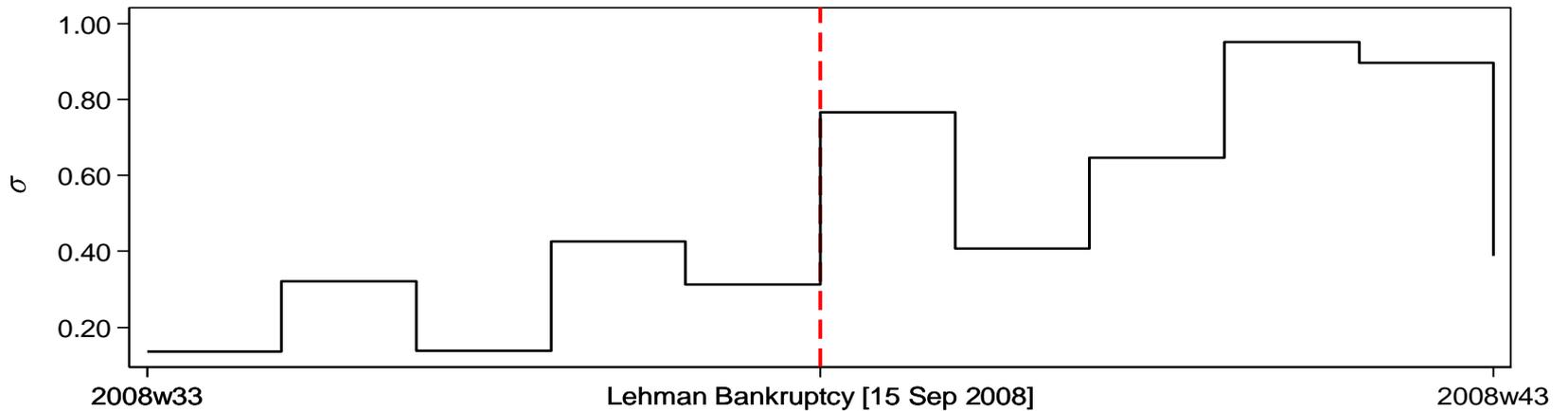
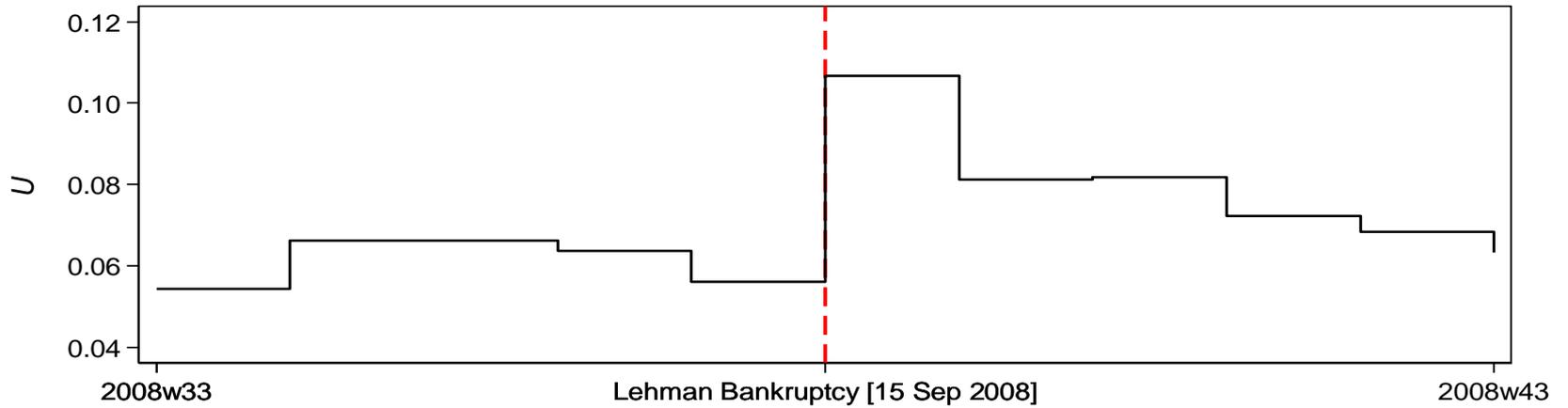


Note: observations with more than 400 articles per days were truncated from the plot.

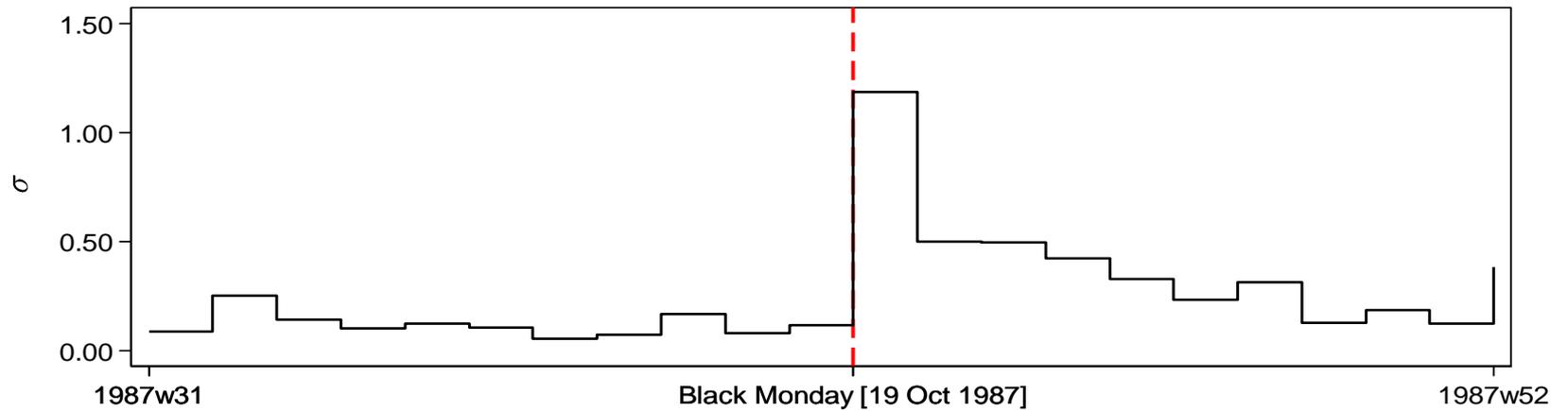
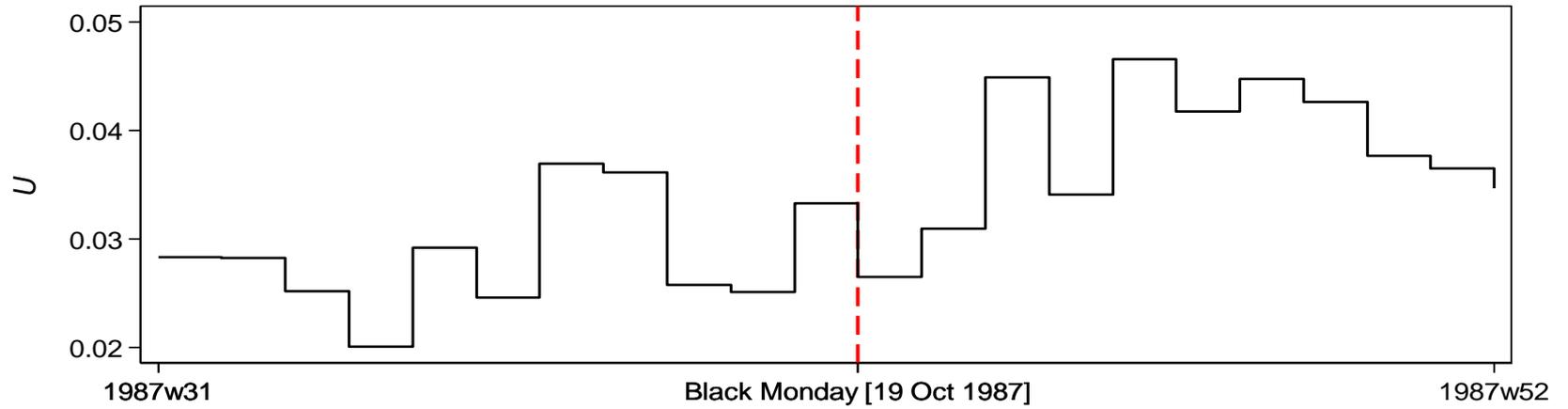
# Stock returns volatility, $\sigma$



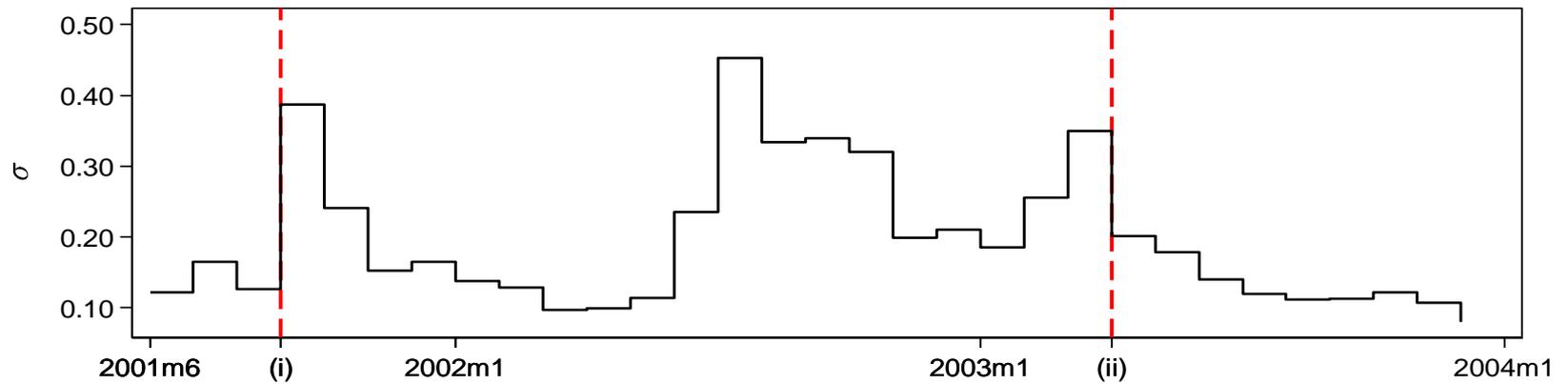
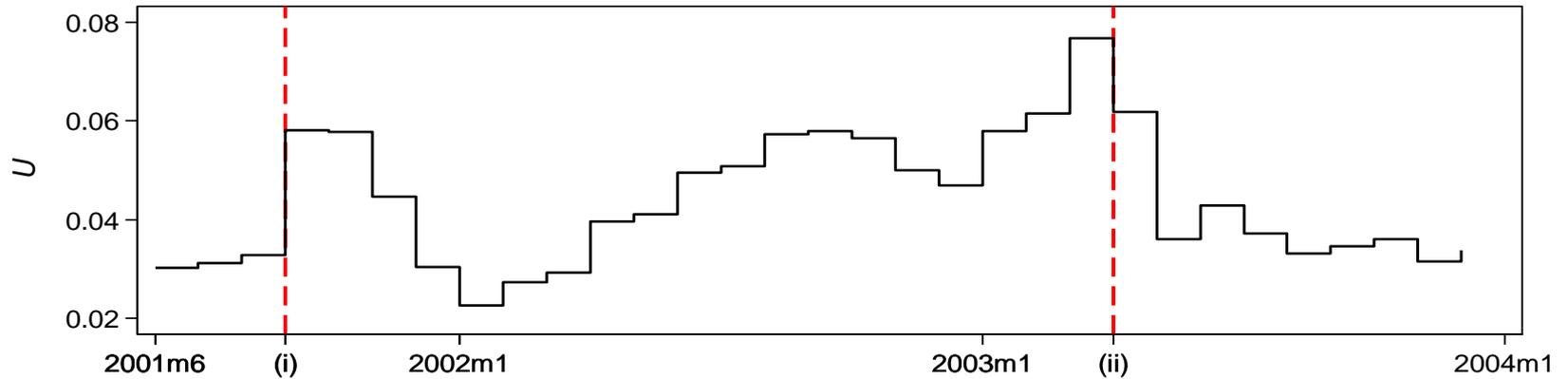
# Lehman collapse 2008



# Black Monday 1987

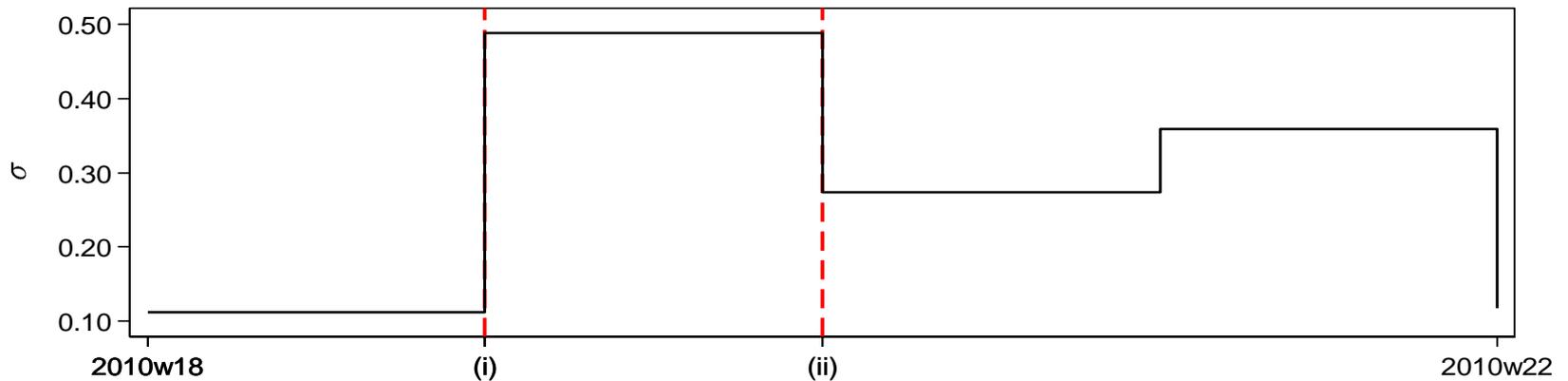
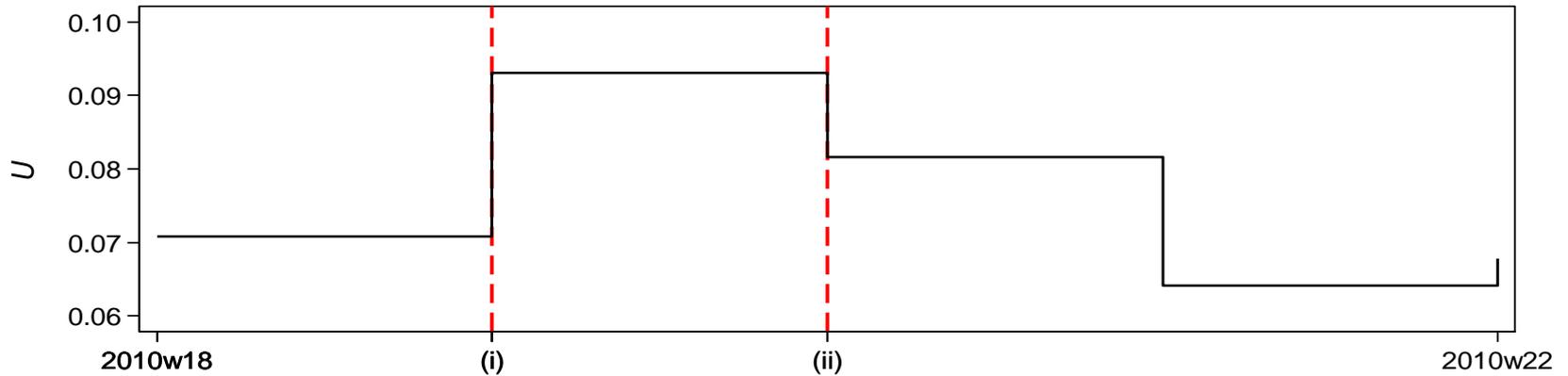


# 9/11 and Iraq War



- (i) Terrorist attacks in US [9 Sep 2001]
- (ii) Iraq War begins (20 Mar 2003)

# UK hung parliament May 2010



- (i) UK national elections result in hung parliament for first time since 1974 [6 May 2010]
- (ii) Coalition government formed [12 May 2010]

# Rank correlation segmented by levels monthly differences, 1984-2012

$L.\sigma$ / $L.u$	< median	$\geq$ median	all
< median	0.108 ↑ +	0.079 ← x ↑ +	0.076 ↑ ++
$\geq$ median	0.406 ↑↑↑	0.306 ↑↑↑	0.331 ↑↑↑
all	0.263 ↑↑↑	0.251 ↑↑↑	0.142 ↑↑↑

Note: significance levels estimated using block bootstrap to account for autocorrelation

# Sign concordance segmented

## monthly differences, 1984-2012

$L.u$ \ $L.\sigma$	< median	$\geq$ median	all
< median	0.539 ↑	0.437 ← ↑ ↑ ↑	0.497 ↑ ↑ ↑
$\geq$ median	0.606 ↑ ↑	0.670 ← ↑ ↑ ↑	0.644 ↑ ↑ ↑
all	0.566 ↑	0.575 ← ↑ ↑	0.571 ↑ ↑ ↑

$ \Delta\sigma $ \ $ \Delta u $	< median	$\geq$ median	all
< median	0.495 ↑	0.500 ← ↑ ↑ ↑	0.497 ↑ ↑ ↑
$\geq$ median	0.600 ↑ +	0.681 ← ↑ ↑ ↑	0.644 ↑ ↑ ↑
all	0.543 ↑	0.598 ← ↑ ↑	0.571 ↑ ↑ ↑

Note: significance levels estimated using block bootstrap to account for autocorrelation