

News-Implied Sovereign Default Risk

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Disclaimer

The views expressed in this presentation are those of the authors and do not necessarily reflect the position of the European Investment Bank.



Assessing sovereign default risk

How to track sovereign default risks if the usual resources are not available?

- No CDS spreads
- No risk ratings
- Debt data is published with a long lag, sometimes with imperfect quality

- Only about 75% of countries worldwide are rated
- 60% of LICs are experiencing debt distress or have high risks

Alternative data sources can help in assessing the current sovereign credit risk situation

IMF Debt sustainability analyses

Local bond yields

Exchange rates

Newspaper articles

In-depth analyses comparable across countries, but with a considerable lag

Not quite informative about risks faced by external creditors

Daily, but public debt only one of the drivers

Real-time, but very highdimensional with low signalto-noise ratio

Broader question: Can we exploit newspaper articles to assess sovereign default risk?



Building a high-frequency proxy for sovereign default risk

Aim of our research: Apply news media text to develop a reliable high-frequency proxy for sovereign default risk that can be computed in real-time for any country without market liquidity constraints

Approach: Use a repository of **international** and **local newspapers**, employ **Natural Language Processing** (NLP) techniques to quantify sovereign default concerns, and assess properties of the measure

The framework does *not* rely on supervised learning which requires an outcome variable This allows us test the performance/additionality of the measure relative to existing proxies

Uses:

- Analyse trends over time in countries lacking high-frequency default risk indicators
- Analyse wider regional trends (due to straightforward aggregation)
- Conduct proactive monitoring (e.g., Early Warning Signals under IFRS9): real-time information
- Revisit theoretical predictions on the influence of default risk on e.g., equity markets

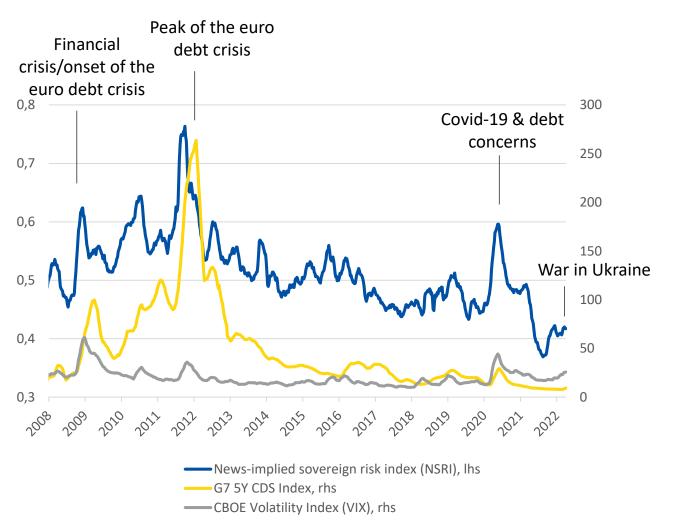


Main results

- Newspaper articles alone can be used to build an informative News-Implied Sovereign Risk Index (NSRI)
- 2. NSRI captures major global risks episodes, including the great financial crisis, subsequent euro debt, and the Covid-19 pandemic
- NSRI co-moves with standard market indicators of global risk such as G7 sovereign CDS spreads and VIX
- NSRI captures country-specific CDS spreads dynamics and provides default risk signals beyond CDS spreads
- 5. NSRI has significant incremental information about future sovereign rating downgrades

Global News-Implied Sovereign Risk Index

Weekly 3-month moving-average



Source: Authors' calculations, Refinitiv Datastream



Case study: NSRI and Russia's invasion of Ukraine

NSRI - Ukraine



NSRI - Russia

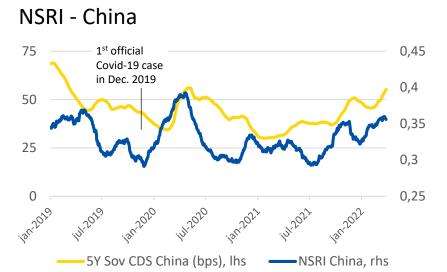


Note: All data 3-month moving averages, last observation 27.03.2022

Source: Authors' calculations, Refinitiv Datastream

- NSRI sends clear warning signals weeks before Russia's invasion of Ukraine on 24 February 2022 for both countries
- The NSRI remains muted compared to CDS spreads and the index's reaction to the 2008 financial crisis
 - Possible explanation: NSRI focuses on news on sovereign risk while the news coverage of the war is centred
 more on political developments with occasional warnings of rating downgrades and Russian default

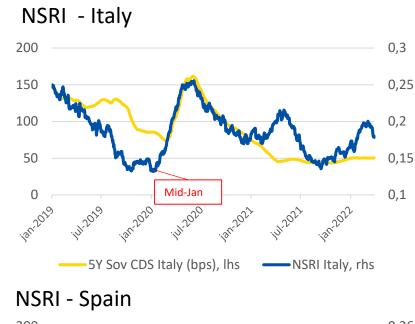
Case study: NSRI and Covid-19

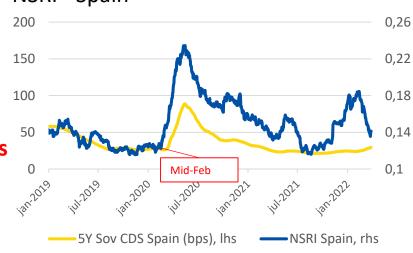


Note: All data 3-month moving averages, last observation 28.03.2022

Source: Authors' calculations, Refinitiv Datastream

- The NSRI quickly reacted to the Covid-19 pandemic, showing a substantial increase in China and other countries from the onset of the crisis
- In Europe, the NSRI responded more rapidly in Italy, the first country of significant transmission, than e.g. Spain NSRI is different across countries
- The NSRI reasonably tracks or leads the evolution of sovereign CDS spreads at the height of the pandemic



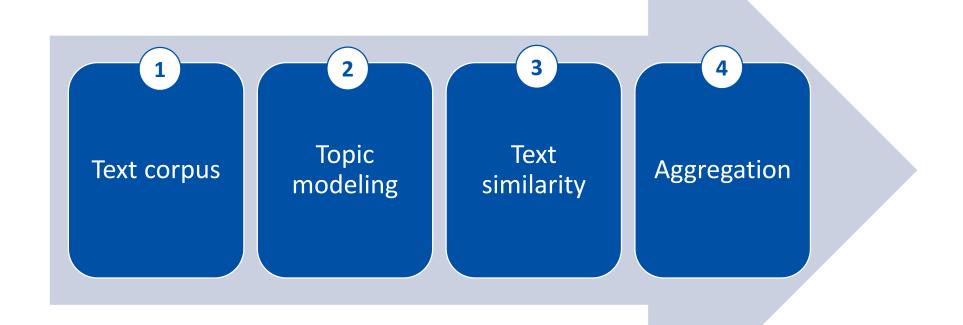




CONSTRUCTING THE NSRI: MEASURING SOVEREIGN RISK USING MACHINE LEARNING



Constructing the NSRI





Constructing the NSRI: 1) Text corpus

Objective: Obtain news covering countries' economic, financial and political developments

Approach: Use the PressReader News API over the period 2006 – 2022

- PressReader provides historical news coverage for many countries, including local and global news publishers
- For each country, retrieve **English language news** that mentions the country's name and one of the key words "economy" and "government" and published either in the country, the US, UK or China
- In total, we retrieved approximately 10 million news articles published by roughly 2,100 publishers covering
 184 countries



Constructing the NSRI: 2) Topic modeling

Objective: Isolate narratives likely to matter for sovereign default risk to reduce noise in the text data

Approach: Use the Latent Dirichlet Allocation (LDA) algorithm to classify news text into a low-level number of topics $T \in \{4, 5, ..., 10\}$

- LDA gives text a hierarchical structure, where documents (news articles) are composed of topics which in turn contain words
- Each document has a probability distribution over latent topics and each topic is defined by a probability distribution over words
- Training the model boils down to deriving the optimal parameters and, more importantly, the number of latent topics T that best fit the data
- Identified nine topics:
 - Economy, Business, Governance I & II, Security, Health, Environment, Tourism, and Miscellaneous



Constructing the NSRI: 3) Text similarity

Objective: Quantify the level of sovereign default concern in each news article about a given country

Approach: Use the **Cosine Similarity algorithm** to compute how similar an article's language is to that of typical texts that signal heightened default risk

- Text similarity allows the use of unsupervised learning techniques that do not depend on market data
- Text similarity algorithms are simple, have low computational cost, and suits our setting where supervised learning seems problematic



Constructing the NSRI: 3) Text similarity (ctd)

Cosine Similarity (CS) is calculated as the inner product of two documents' vectors divided by the product of the vectors' Euclidean lengths

$$CS = \cos(\mathbf{u}, \mathbf{v}) = \frac{\mathbf{u} \cdot \mathbf{v}}{|\mathbf{u}| \times |\mathbf{v}|} \in [0, 1]$$

where, **u** is the vector of an arbitrary news article **v** is the vector of a representative sovereign default risk document

u, v are based on term frequency—inverse document frequency (tf-idf) vectorization — mitigates spurious association between document vectors

Specify v as in studies that construct representative text in other domains (e.g., Hassan et al., 2019, Engle et al., 2020):

Compile a training library of 448 documents centered on worsening sovereign risk and default concerns:

- Textbooks on sovereign default
- Sovereign credit rating downgrade reports by Moody's, Standard & Poor's, and Fitch
- Excerpts from the EIB's sovereign rating rationales for lowly rated sovereigns
- Country risk analysis reports by the Economist Intelligence Unit
- Debt sustainability analysis reports by the IMF/World Bank
- News articles on sovereign defaults and debt problems

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Constructing the NSRI: 4) Aggregation

Objective: Aggregate the information obtained from individual news articles by country and period

Approach: Average article-level CS scores at a daily frequency using only news articles in **sovereign risk-relevant topics**:

$$NSRI_{j,t} = \frac{1}{n_{j,t}^r} \sum_{i}^{n_{j,t}^r} CS_{i,j,t}$$

where $CS_{i,j,t}$ is the CS score of article i published on day t about country j

 $n_{i,t}^r$ is the number of news articles in the sovereign risk-relevant topics published about country j on day t

Sovereign risk-relevant topics are "Economy", "Business", "Governance" and "Security"



NSRI: VALIDATION



Constructing the NSRI: 5) Validation

Objective: Confirm **good separation of noise from likely relevant narratives**

Approach: Inspect topics and terms

S/n	Topic 1 Environment	Topic 2 Security	Topic 3 Miscellaneous	Topic 4 Governance I	Topic 5 Health	Topic 6 Business	Topic 7 Governance II	Topic 8 Tourism	Topic 9 Economy
1	oil	police	woman	right	health	business	president	water	market
2	energy	attack	world	law	pandemic	development	minister	food	bank
3	climate	kill	school	party	virus	project	leader	flight	growth
4	climate change	rebel	family	political	medical	technology	prime minister	tourist	economy
5	fuel	force	team	people	coronavirus	service	war	passenger	rate
6	coal	investigation	sports	court	vaccine	company	peace	tourism	investor
7	mining	intelligence	game	public	covid	sector	millitary	airline	tax
8	oil gas	report	cup	election	hospital	industry	sanction	ship	financial
9	emission	security	play	vote	disease	cooperation	opposition	airport	trade
10	electricity	arrest	love	corruption	patient	programme	administration	hotel	economic
11	oil price	government	film	government	outbreak	opportunity	state	road	debt
12	carbon	terrorist	book	act	infection	digital	meeting	island	price
13	power	millitary	friend	bill	lockdown	international	deal	transport	stock
14	renewable	millitant	football	freedom	people	local	political	travel	investment
15	natural gas	violence	player	state	doctor	partnership	nation	traffic	income
16	solar	army	young	rule	death	innovation	conflict	sea	currency
17	natural	civilian	life	voter	treatment	system	diplomatic	park	export
18	petroleum	crime	home	legal	health care	economic	international	vessel	sale
19	crude	protest	old	democracy	worker	infrastructure	negotiation	boat	gdp
20	gas	soldier	child	president	test	country	ally	route	business
% of News	4%	12%	12%	15%	6%	11%	15%	7%	17%

Security + Governance I + Governance II + Business + Economy ≈ 70% of the news corpus



Constructing the NSRI: 5) Validation ctd



Terms with the highest tf-idf scores in our Sovereign Default Risk Library Font size is proportional to term weight



Constructing the NSRI: 5) Validation ctd2

News-level Topics and CS Scores vs. News Sentiment

News Article Title	(1) CS	(2) Sentiment	(3) Topic
Panel A: News signaling sovereign default concern	S		
Zambia on brink of defaulting on foreign debt	0.81	0.72	Economy
Zambia's default fuels fears of African 'debt tsunami' as Covid impact bites	0.80	0.74	Economy
Argentina defaults yet again, but hopes to get off lightly	0.63	0.94	Economy
Already in default, Argentina hits an impasse with creditors over debt restructuring	0.62	0.82	Governance II
'Painful' downgrades will raise South Africa's borrowing costs, minister says	0.55	0.90	Economy
What the latest rating downgrades mean for the average South African	0.64	0.87	Economy
World Bank says world leaders moving away from debt cancellation for Africa	0.67	0.88	Business
Panel B: News on socioeconomic problems			
Myanmar: Military Coup Kills Fragile Democracy	0.13	0.91	Security
Nigeria's Economy Faces Worst Recession in Four Decades, says New Report	0.23	0.73	Economy
Recession: Nigeria's Economic Crisis Requires a Political Solution	0.31	0.69	Economy
Turkey's Erdogan sacks central bank governor after rate hike	0.11	0.50	Economy
Venezuela crisis: How the political situation escalated	0.08	0.87	Governance I
Chadian President Idriss Deby dies on frontline, rebels vow to keep fighting	0.13	0.78	Governance I
Nigeria's inflation rises again, hit four-year high February	0.12	0.50	Economy
Panel C: Miscellaneous news			
Minister orders intelligence-led operation against violence in South-East	0.03	0.89	Security
Man who drove into News Cafe in Rosebank appears in court	0.03	0.98	Security
Uber pledges to boost safety for SA drivers as accidents rise	0.07	0.67	Health
UK imposes sanctions on Russians, Guptas in first use of anti-corruption law	0.09	0.86	Security
Naomi Osaka withdraws from French Open amid row over press conferences	0.09	0.80	Miscellaneous
Chelsea beat Man City to win Champions League	0.07	0.58	Miscellaneous

CS is markedly different from news sentiment and sufficiently differentiates news signaling elevated sovereign credit risk from those that do not



NSRI: EMPIRICAL RESULTS



NSRI and CDS spreads

	NSRI % change rel. to $t-1$				NSRI %	NSRI % change rel. to 3-month median			
	Contemporaneous		Lag	Lagged		Contemporaneous		Lagged	
Dep. var.: CDS Spread % change	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	
Panel A: Weekly frequency									
NSRI % change	0.399	0.258	-0.055	-0.095	0.391	0.213	-0.132	-0.158	
	(5.05)	(4.39)	(-1.15)	(-2.12)	(5.35)	(3.58)	(-1.91)	(-2.43)	
Controls	NO	YES	NO	YES	NO	YES	NO	YES	
R^{2} (%)	25.45	38.73	25.19	28.71	24.86	38.04	24.64	28.32	
Obs.	14,004	14,004	14,004	14,004	13,779	13,779	13,779	13,779	
Panel B: Monthly frequency									
NSRI % change	3.117	1.713	1.495	1.021	4.804	2.384	1.673	1.100	
	(7.11)	(9.59)	(3.53)	(3.47)	(9.08)	(8.62)	(3.87)	(3.27)	
Controls	NO	YES	NO	YES	NO	YES	NO	YES	
R^{2} (%)	10.81	46.33	8.66	18.99	14.67	47.64	8.97	19.27	
Obs.	3,015	3,015	3,015	3,015	3,073	3,073	3,073	3,073	

$$\Delta CDS_{j,t} = \beta_0 + \beta_1 \Delta NSRI_{j,t-\tau} + \gamma X + \xi_j + \lambda_t + \epsilon_{j,t}$$

 \boldsymbol{X} includes lag of $\Delta CDS_{j,t}$, country-specific news sentiment, stock market return, and volatility; it also includes global macro-financial variables – the world stock market return, implied volatility index (VIX), US economic policy uncertainty Index (EPU), and US economic activity index (ADS)

NSRI is significantly associated with CDS spread contemporaneously and predicts future CDS spread dynamics



NSRI and sovereign rating downgrades

	(1)	(2)	(3)	(4)	(5)	(6)
NSRI	0.114	0.115	0.092	0.092	0.080	0.088
	(6.51)	(6.63)	(4.31)	(4.31)	(3.78)	(3.76)
Sentiment		0.051	0.048	0.047	0.037	0.092
Ret		(3.12)	(1.80) -0.017 (-1.45)	(1.77) -0.023 (-1.23)	(1.38) -0.025 (-1.44)	(3.60) -0.044 (-2.62)
Vol			0.155 (9.46)	0.158 (8.95)	0.115 (5.35)	0.119 (3.98)
World Ret			(====)	0.011 (0.48)	0.031 (1.48)	0.066 (4.08)
EPU				,	0.041 (2.08)	0.044 (1.54)
ADS					0.010 (0.57)	0.012 (0.65)
VIX					0.056 (1.97)	0.051 (1.20)
CDS Spread						0.008
R ² (%) Obs.	16.5 2,605	17.3 2,605	27.5 1,716	27.5 1,716	28.9 1,716	(0.46) 33.5 1,074

$$Downgrade_{j,t} = \beta_0 + \beta_1 NSRI_{j,t-\tau} + \gamma \mathbf{X} + \xi_j + \epsilon_{j,t}$$

One SD \upsigma in average NSRI over the 30-day window ending t-2 predicts \upsigma in the probability of downgrade by 15–22% relative to the fraction of downgrades in the data

NSRI complements CDS spread; useful as a stand-alone high-frequency default risk proxy



NSRI and equity markets I

$$Ret_{j,t} = \beta_0 + \beta_1 \Delta NSRI_{j,t-\tau} + \gamma X + \xi_j + \lambda_t + \epsilon_{j,t}$$

	Pa	nel A: Dolla	r index retu	ırn	Panel E	Panel B: Local-currency index return				
	Contemporaneous		Lag	Lagged		oraneous	Lag	Lagged		
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)		
ΔNSRI	-9.316	-7.120	7.560	6.443	-6.111	-3.978	6.294	5.466		
Controls R ² (%)	(-4.25) NO 11.74	(-3.50) YES 36.78	(2.75) NO 11.72	(2.39) YES 13.40	(-3.14) NO 10.03	(-2.31) YES 32.50	(2.70) NO 10.03	(2.37) YES 11.61		
Obs.	24,101	24,101	24,101	24,101	24,004	24,004	24,004	24,004		

$$Ret_{j,t} = \beta_0 + \beta_1 \Delta NSRI_{j,t-\tau} + \beta_2 \Delta NSRI_{j,t-\tau} \times HighFiscalConstraint_{j,t} + HighFiscalConstraint_{j,t} + \gamma \mathbf{X} + \xi_j + \lambda_t + \epsilon_{j,t}$$

	Contemp	oraneous	Lagged		
Dollar index return	(1)	(2)	(3)	(4)	
ΔNSRI	-3.078	-1.651	12.023	11.863	
	(-1.11)	(-0.68)	(2.61)	(2.61)	
Δ NSRI \times High Fiscal Constr.	-10.005	-9.098	-8.586	-9.958	
	(-2.48)	(-2.44)	(-1.74)	(-2.04)	
High Fiscal Constr.	8.221	6.912	9.242	10.431	
	(1.81)	(1.55)	(2.05)	(2.13)	
Controls	NO NO	YES	`NO ´	YES	
R^{2} (%)	11.74	36.73	11.72	13.41	
Obs.	23,862	23,862	23,862	23,862	

 \boldsymbol{X} includes lag of $Ret_{j,t}$ country-specific news sentiment, and stock market volatility. Also includes global macrofinancial variables — the world stock market return, implied volatility index (VIX), US economic policy uncertainty Index (EPU), and US economic activity index

 $Ret_{i,t}$ is in basis points

Continuous right hand side variables are normalized to unit variance



NSRI and equity markets II

	Pa	nel A: Dolla	ar index retu	ırn	Panel B: Local-currency index return				
	Contemp	Contemporaneous		Lagged		oraneous	Lag	Lagged	
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	
ΔNSRI	-7.093	-9.908	5.287	5.138	-3.770	-5.765	4.129	4.350	
	(-3.07)	(-4.23)	(1.86)	(1.80)	(-1.90)	(-2.89)	(1.72)	(1.80)	
Δ GNSRI	-18.900	-21.554	19.640	16.683	-19.914	-21.741	18.551	15.976	
	(-6.53)	(-7.93)	(6.45)	(5.71)	(-7.41)	(-8.24)	(6.21)	(5.66)	
Ret(t-1)		-41.845		-39.172		-33.777		-31.676	
		(-6.22)		(-5.96)		(-5.36)		(-5.15)	
Sentiment		-18.085		10.578		-13.709		9.520	
		(-3.94)		(2.85)		(-3.25)		(2.70)	
R^2 (%)	11.88	13.27	11.87	13.06	10.23	11.45	10.20	11.26	
Obs.	24,101	24,101	24,101	24,101	24,004	24,004	24,004	24,004	

- $\triangle GNSRI$ is the percentage change of the Global NSRI. Global NSRI is computed as the cross-sectional weighted average of NSRI using last year-end's equity market capitalization as weights
- Investors tend to respond more to default concerns that are more global in nature compared to country-specific concerns global (push) factors are crucial for local asset prices



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Conclusion

- We developed a novel framework for quantifying sovereign default risk in real-time using big data and natural language processing techniques
- We show that news media narratives contain valuable information about countries' riskiness useful for tracking default concerns in real-time
- The NSRI is informative about sovereign CDS spread dynamics and future credit rating downgrades
 - Useful as a high-frequency default risk proxy, especially where daily market-based data or other proxies are not available
 - The NSRI can be integrated as an important early warning indicator of any sovereign risk monitoring framework
- The NSRI allows the measurement of both global risk episodes, such as the Covid-19 pandemic, the GFC and the war in Ukraine, and country-specific risks

