

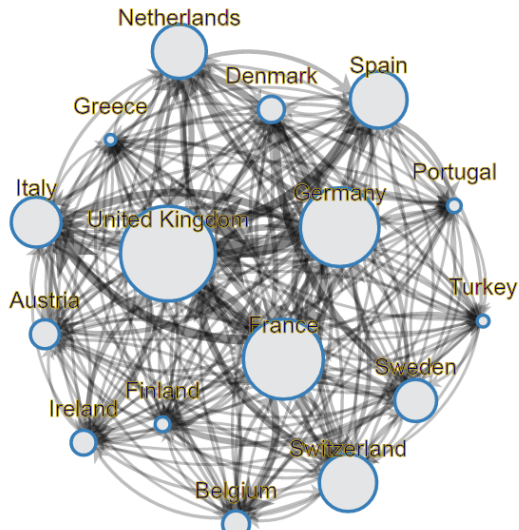


# RiskRank: Measuring interconnected risk

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2000 2001 2002 2003 2004 2005 2006 2007 2008 2009 2010 2011 2012 2013 2014 2015





# Motivation

- ▶ An acute interest in new approaches to assess systemic risk
- ▶ Financial crises triggered by various shocks (unpredictable)...
- ▶ ...but widespread imbalances build-up ex ante (identifiable)
- ▶ Analytics to identify systemic risk at early stages
  - ▶ Cyclical: Build-ups of widespread macro-financial imbalances
  - ▶ Cross-sectional: Interconnected nature of the financial system
- ▶ How to assess vulnerability when risk is interconnected?



# Related literature

## Early-warning models

- ▶ **Univariate:** Kaminsky & Reinhart 1999, El-Shagi et al 2013
- ▶ **Logit:** Lo Duca & Peltonen, 2013, Betz et al 2014
- ▶ **Machine learning:** Holopainen & Sarlin 2015

## Interconnectedness

- ▶ **Estimated:** Poon et al 2005, Billio et al 2011, Diebold&Yilmaz 2014
- ▶ **Real:** Upper & Worms 2004, van Lelyveld & Liedorp 2006, Poledna et al, 2015

## Early-warning models & interconnectedness

- ▶ **Central:** Rose&Spiegel 2009, Minoiu et al 2013, Rancan et al 2015
- ▶ **Pass-through:** Peltonen et al 2015, Hale et al 2015



# Systemic risk aggregation

- ▶ From risk indicators to probability
  - ▶ Signaling: Monitor univariate indicators
  - ▶ Non/linear approaches for combining indicators
  - ▶ Ensemble learning for model aggregation
- ▶ From interlinkages to centrality
  - ▶ In, out & total strength/degree
  - ▶ Betweenness, closeness & eigenvector centrality
  - ▶ DebtRank (Battiston et al 2012)
- ▶ How to combine probabilities and links?



# Contribution

Measuring interconnected risk

- ▶ EWMs aggregate indicators & network measures connectivity
- ▶ We assume a hierarchical & interconnected system of nodes
- ▶ RiskRank: Likelihood & impact in an interconnected system

Key features of RiskRank

- ▶ A general-purpose measure of interconnected risk
- ▶ Allows disentangling individual, direct & indirect effects
- ▶ Allows multiple indirect effects and feedback loops
- ▶ Beyond entities, aggregates upward in the hierarchy



# Aggregation operators

- ▶ Conventional aggregation operators
  - ▶ Min/max: con-/disjunctive operators
  - ▶ Weighted mean: fix trade-off & compensatory
  - ▶ Quadratic/geometric/harmonic/power  $\alpha$  mean
  - ▶ Ordered Weighted Average (OWA) (Yager 1988)
- ▶ Choquet (1953) integral as a general aggregation operator
  - ▶ Includes the above (and more) as special cases
  - ▶ Generalizes to non-linearity and non-additivity
  - ▶ Extends conventional operators with interactions



# Choquet integral

## Definition

Fuzzy measure  $\mu$  on the finite set  $N = \{1, 2, \dots, n\}$  is a set function  $\mu : P(N) \rightarrow [0, 1]$  (where  $P(N)$  is the power set of  $N$ ) satisfying the following two conditions:

- ▶  $\mu(\phi) = 0, \mu(N) = 1$ ;
- ▶ Monotonic, non-decreasing:  $A \subseteq B$  implies that  $\mu(A) \leq \mu(B)$ .

## Definition

Discrete Choquet integral with respect to a monotone measure  $\mu$  is

$$C_{\mu}(x_1, \dots, x_n) = \sum_{i=1}^n (\mu(C_{(i)}) - \mu(C_{(i+1)})) x_{(i)}$$

where  $x_{(i)}$  denotes a permutation of the  $x_i$  values such that  $x_{(1)} \leq x_{(2)} \leq \dots \leq x_{(n)}$  and  $C_{(i)} = \{c_{(i)}, c_{(i+1)}, \dots, c_{(n)}\}$ .





# Additive Choquet integral

The additive Choquet integral is the weighted sum

$$C_{\mu}(x_1, \dots, x_n) = \sum_{i=1}^n \mu(c_{(i)}) x_{(i)}$$

2-additive case covers pairwise interactions and individual effects

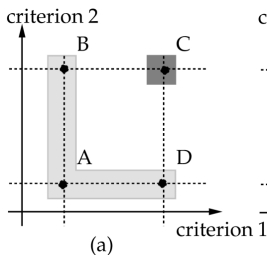
$$C_{\mu}(x_1, \dots, x_n) = \sum_{i=1}^n (v(c_i) - \frac{1}{2} \sum_{j \neq i} I(c_i, c_j)) x_i + \sum_{I(c_i, c_j) > 0} I(c_i, c_j) \min(x_i, x_j) + \sum_{I(c_i, c_j) < 0} |I(c_i, c_j)| \max(x_i, x_j)$$

where  $v(c_i)$  stands for the Shapley-index (average contribution of fixed element  $x_i$  in any subset) and  $I(c_i, c_j) \in [-1, 1]$  for the interaction. This relies on the Möbius transformation of  $\mu$  and that it equals 0 on any subset with cardinality above 2 (Grabisch, 1997)

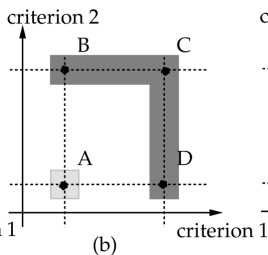


# Case of utility theory

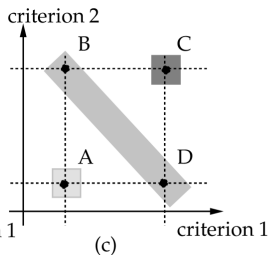
$$C_{\mu}(x_1, \dots, x_n) = \sum_{i=1}^n (v(c_i) - \frac{1}{2} \sum_{j \neq i} I(c_i, c_j)) x_i + \sum_{I(c_i, c_j) > 0} I(c_i, c_j) \min(x_i, x_j) + \sum_{I(c_i, c_j) < 0} |I(c_i, c_j)| \max(x_i, x_j)$$



Positive interaction  
Complements  
 $I(c_i, c_j) > 0$   
**min** operator



Negative interaction  
Substitutes  
 $I(c_i, c_j) < 0$   
**max** operator



No interaction  
Independent  
 $I(c_i, c_j) = 0$   
no operator



# From Choquet to RiskRank

For risk levels  $x_i$  and links  $I(c_i, c_j)$ , 2-additive RiskRank is

$$RR_c = \underbrace{w(c)x_c}_{\text{Individual effect of entity } c} + \underbrace{\sum_{i=1}^n (v(c_i) - \frac{1}{2} \sum_{j \neq i} I(c_i, c_j)) x_i}_{\text{Direct effects of entity } i \text{ on } c} + \underbrace{\sum_i \sum_{j \neq i} I(c_i, c_j) \Pi(x_i, x_j)}_{\text{Indirect effects of } j \text{ via } i \text{ on } c}$$

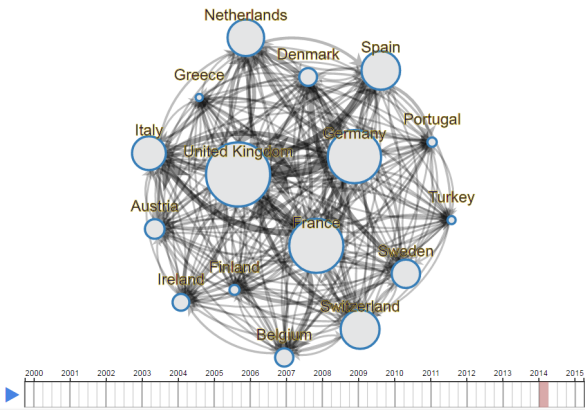
which

- ▶ separates effects
- ▶ allows multiple indirect effects via  $k$ -additivity
- ▶ allows simulated feedback through dynamic iteration



# Real-world example

- ▶ EU early-warning model & BIS exposures
  - ▶ Sample: 15 EU countries, 1980Q1-2015Q1
  - ▶ Domestic risk: Logit model, 14 macro-financial indicators
  - ▶ Linkages: BIS foreign claims, immediate borrower
  - ▶ Output: Country & EU-level measures of interconnected risk





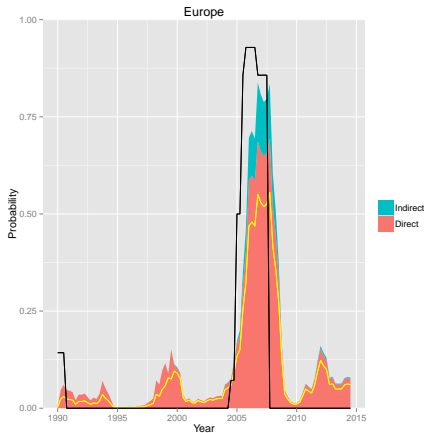
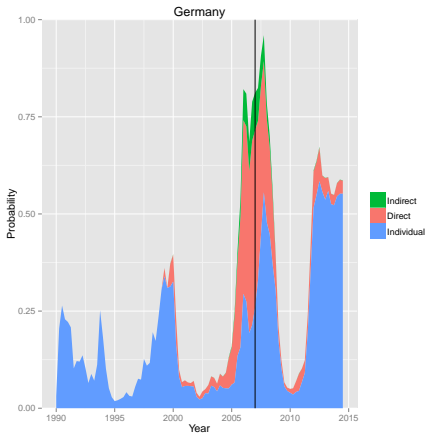
## Real-world example

- ▶  $U_r(\mu = 0.8)$ : Model usefulness with imbalanced preferences
- ▶ AUC: Model usefulness for all  $\mu$

$\mu$	Individual		RiskRank	
	$U_r(\mu)$	AUC	$U_r(\mu)$	AUC
0.0	0 %	0.915	0 %	0.934
0.1	-6 %	0.915	1 %	0.934
0.2	-3 %	0.915	3 %	0.934
0.3	6 %	0.915	14 %	0.934
0.4	12 %	0.915	28 %	0.934
0.5	15 %	0.915	37 %	0.934
0.6	25 %	0.915	47 %	0.934
0.7	44 %	0.915	59 %	0.934
0.8	60 %	0.915	69 %	0.934
0.9	73 %	0.915	78 %	0.934
1.0	0 %	0.915	0 %	0.934

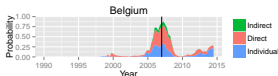


# Real-world examples





# Real-world examples





# Conclusion

- ▶ RiskRank as a measure of systemic risk
  - ▶ EWM: From indicators to probability
  - ▶ Centrality: From interlinkages to centrality
  - ▶ RiskRank aggregates probabilities over links
- ▶ The properties of RiskRank are general in nature
  - ▶ Allows disentangling individual, direct & indirect effects
  - ▶ Allows multiple indirect effects and feedback loops
  - ▶ Beyond entities, aggregates upward in the hierarchy
  - ▶ Could be used for measuring any interconnected risk





Thanks for your attention!