



EUROPEAN CENTRAL BANK

EUROSYSTEM

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Two is company, three's a crowd:


Automated pairing and matching of
two-sided reporting in EMIR
derivatives' data

IFC conference

Are post-crisis statistical initiatives completed?

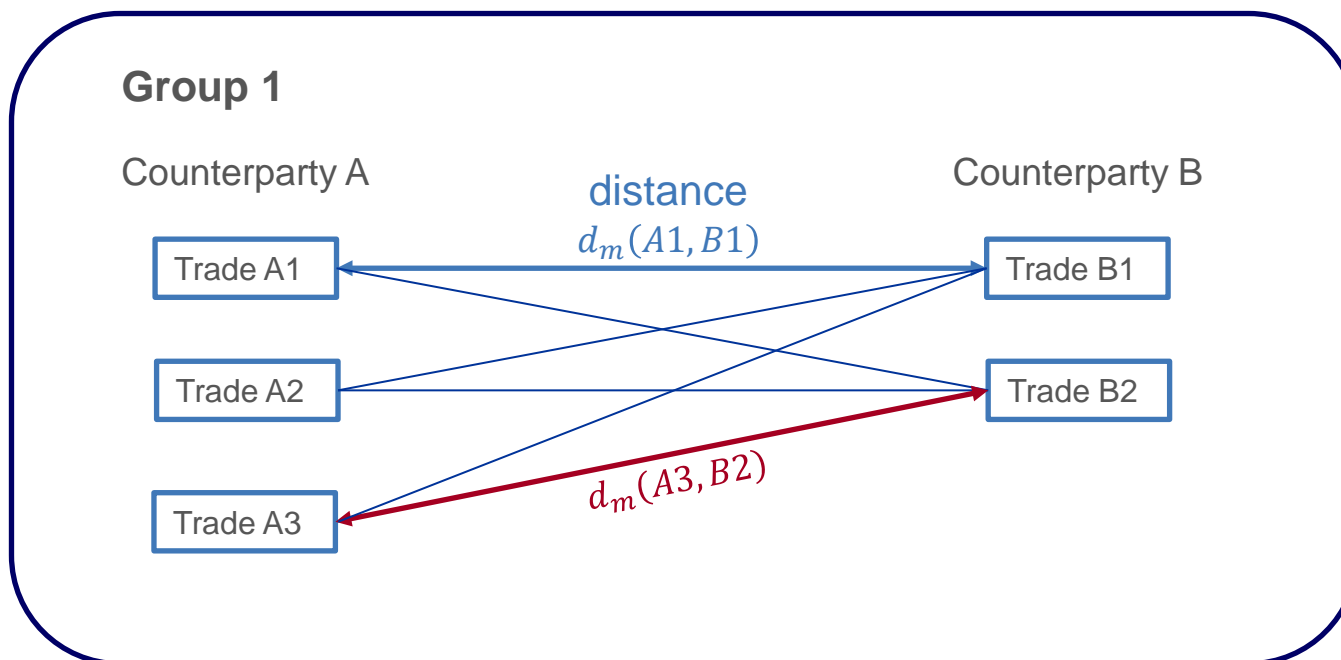
30-31 August 2018, Basel

Motivation: double-sided reporting in EMIR

- EU counterparties report **derivative transactions** to trade repositories
- Separately by both counterparties (**double-sided reporting**)
 - Improves data quality monitoring
 -  Risk of double-counting when analysing and aggregating the data
- **UTI (Unique Transaction Identifier)** to link trades, agreed between counterparties
 - Challenges in implementation (not unique, different UTIs for the same trade)
 - Work on improving pairing and matching (inter-TR reconciliation process)
 - Global initiatives to harmonise UTI between jurisdictions
 - ESMA estimates pairing rate at 87% = but newly reported trades only

Method

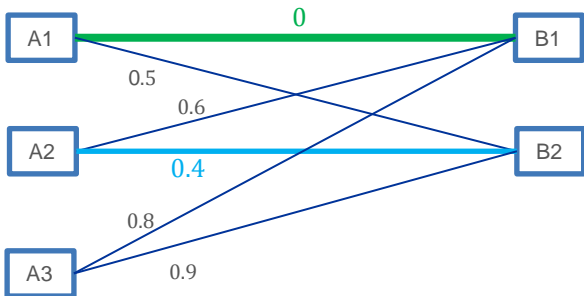
- The trades are split into groups with same values of the **grouping** variables
- The procedure calculates the **matching distance** between each member trade of the group



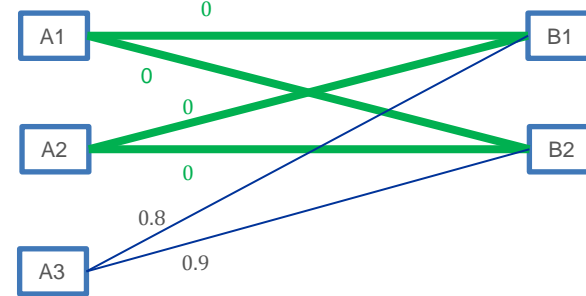
Classification of trades

- Depending on outcome, exclusive categories

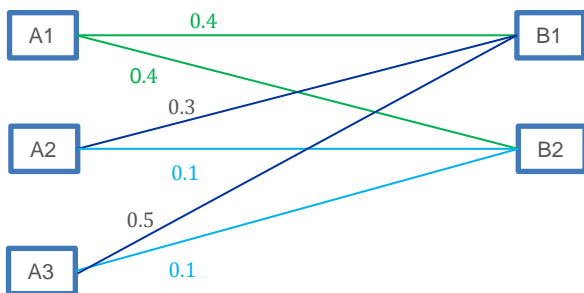
Perfect match / imperfect match



Perfect matching group



Ambiguous



No match



Implementation

- Sample paired with UTI was used to calibrate the parameters
- **Grouping variables:**
 - Counterparties' IDs
 - Asset class
 - Contract type
 - Clearing status
 - Execution date
- **Matching distance weights:** function of fidelity (how good) and specificity (how revealing) of the variable
- **Thresholds of the distance function:** to accept 98% of the observations in the paired sample

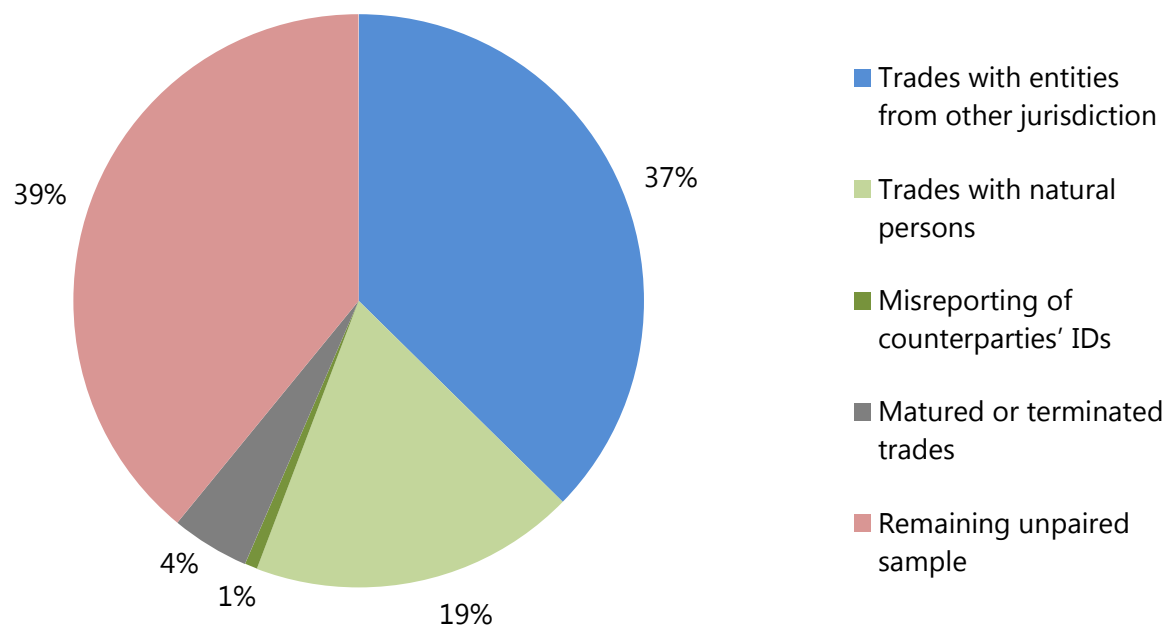
Implementation

- Our procedure has limited impact in the non-paired sample
- Most of these trades don't have any counterpart in their group
→ other reporting issues are at stake

	Paired sub-sample	Non-paired sub-sample
Perfect match	48.74%	0.49%
Imperfect match	28.39%	2.73%
Perfect matching group	12.99%	0.08%
Ambiguous	5.53%	3.61%
No match	4.35%	93.09%

Some things we will never be able to pair

Breakdown of the unpaired sample



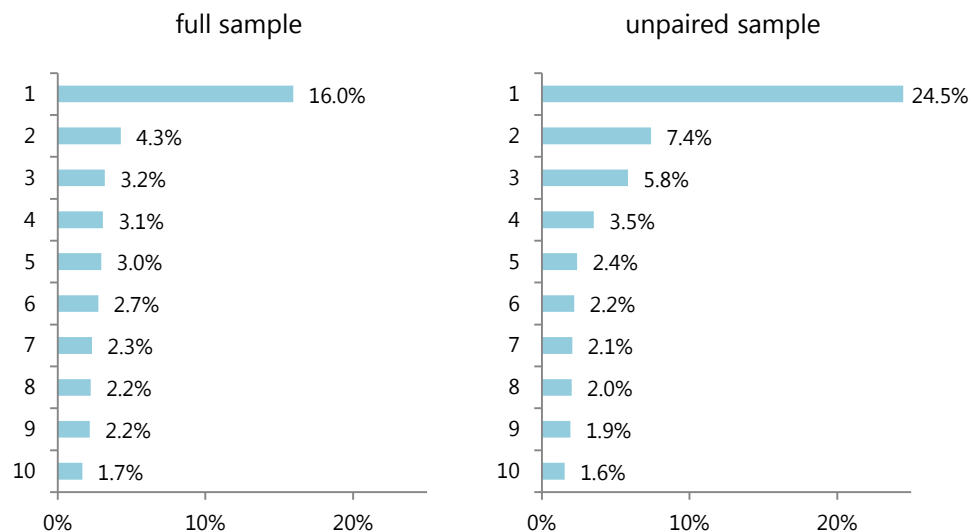
- Remaining 40% of the unpaired sample:
 - may be a result of underreporting
 - may constitute invalid reports

What can't we pair?

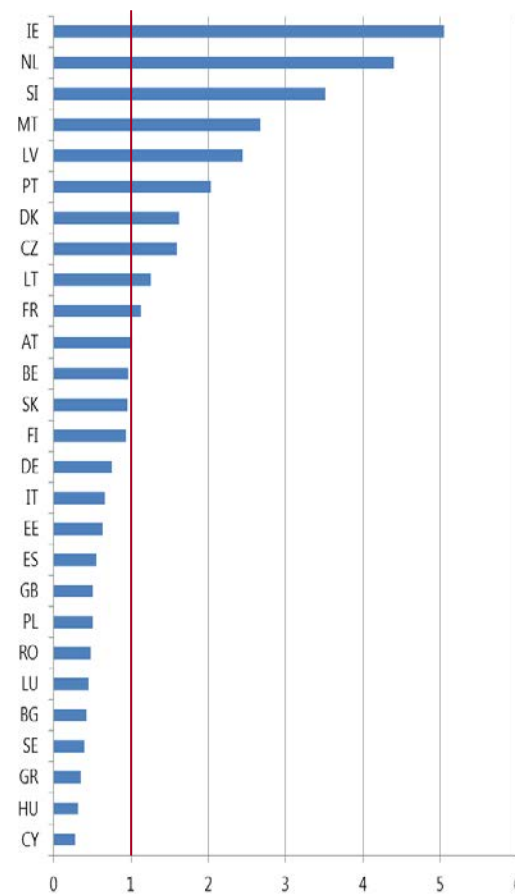
Logit regression - Odds ratios, probability of being paired					
Variable		Odds ratio	Variable		Odds ratio
Asset class			Location of other counterparty		
Commodity		1	Euro area		1
Credit		0.547	Other EU		0.405
Currency		0.508	RoW		0.00676
Equity		0.619	Nature of reporting party		
Interest rate		0.745	CCP		1
Other		0.324	Financial		1.378
Missing		1.371	Non-financial		4.740
Contract type			Other		2.600
Contracts for difference		1	Missing		5.220
Forwards		1.906	Execution date		
Forward rate agreements		1.919	<= 2013		0.247
Futures		0.879	2014-2017		0.403
Option		1.197	2017		0.645
Other		0.420	2018 Q1		1.178
Swap		1.770	2018 Apr-May		1.197
Swaption		1.899	2018 Jun		1
Missing		0.173	2018 Jul		0.599
Clearing Status			> Aug 2018		0.608
No		1	Contract value missing		
Yes		1.720	No		1
Missing		0.439	Yes		0.590
Intra group			Notional amount (log)		
No		1			1.036
Yes		4.228			
Missing		0.669			

What can't we pair?

Share of largest reporters in the unpaired and the full sample



Pairing success by country of the other counterparty (odds ratio)



Conclusions

- Caution is recommended when **making assumptions** about the unpaired sample **to compute aggregates**
- A **significant share of the non-paired sample is difficult to interpret**, and cannot be easily reconciled
- There exist some **clear patterns** between some characteristics of the contracts and **probability of being paired**
- The unpaired sample exhibits **higher concentration** with regards to reporting entities
- A **focused data quality management** process may bring **significant benefits with limited effort**