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# Statistical matching for anomaly detection in insurance assets granular reporting



Vittoria La Serra, Emiliano Svezia

Bank of Italy

Statistical Data Collection and Processing Directorate

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- **Data Quality Management** (DQM) in central banks: necessary to ensure high quality in disseminated statistics.
- **Automation** of DQM processes is crucial:
  - to manage the volume of increasingly granular databases
  - to ensure resilience in situations of human resources constraints (pandemic)
- **Machine Learning** (ML) models: emerging to solve DQM issues
- **Proposal:** a record linkage approach using ML models to deal with a DQM issue on insurance granular assets data.

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- **European Insurance Corporations** (ICs) quarterly report to national supervisory authorities and central banks since 2016 (Solvency II Directive).
- **Asset-by-asset information** is provided in template S.06.02 and used for statistical purposes by central banks.
- **Each asset** of an IC is reported with:
  - An identification (ID) code → required to be kept stable and consistent over time
  - A set of qualitative and quantitative features



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- **The DQM issue:** reporting errors in ID codes might occur.
- **Consequences:**
  - two assets from two subsequent quarters are perceived as different when in reality are the same;
  - decrease in quality of IC statistics to be compiled and disseminated.
- **The goal:** to build a model that is capable of identifying pairs of assets that do not share the same ID code but actually refer to the same asset.



## A record linkage approach

- Select two datasets containing **assets from two subsequent quarters**  $Q_t$  and  $Q_{t+1}$ .
- Same assets are similar on the observed features  $\rightarrow$  build a **comparison matrix** to compare all pairs of assets **reported by the same IC** on observed features (qualitative/quantitative) via distance measures.
- Fit **supervised classification models** on the matrix, where the target variable to predict is the binary status of each pair:  $\{match, non-match\}$ .

Asset codes		Target Status	Distance measures on the observed features			
Quarter $Q_t$	Quarter $Q_{t+1}$		Nominal	Ordinal	Numerical	Textual
Code A	Code A	Match	...	...	...	...
Code A	Code B	Non-match	...	...	...	...
Code B	Code B	Match	...	...	...	...
...	...	...	...	...	...	...

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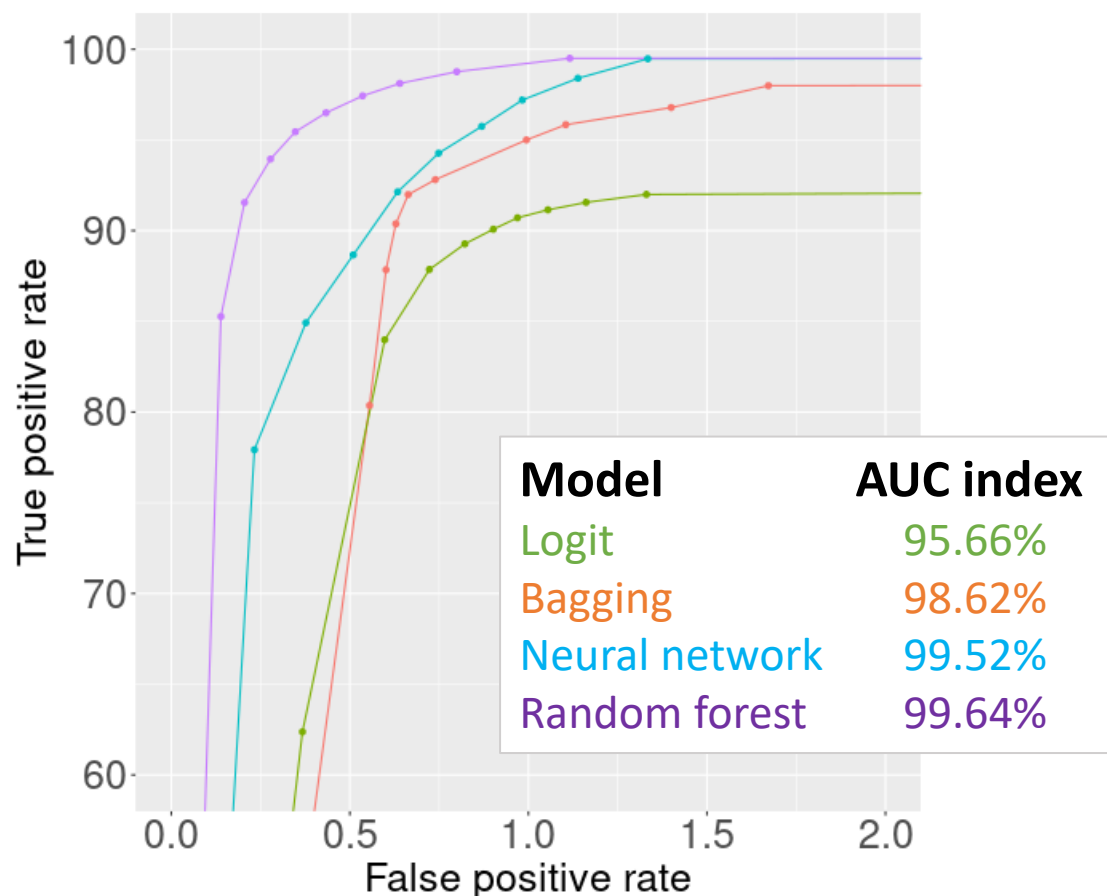
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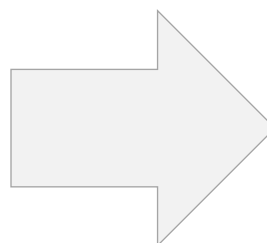
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- From the **Italian database**, assets from two subsequent quarters are selected and compared, building the comparison matrix.
- 70,000 assets reported on average at each quarter → **billions of pairs of assets to compare**.
- Different supervised classification models have been trained and tested:
  - Logit (benchmark), bagging, random forest, neural networks.
  - Fine tuned for different **hyperparameters** combinations (e.g. number of trees, number of hidden layers).
  - Repeatedly **fitted on differently unbalanced datasets** (w.r.t. the target) to ensure **robust results**.



**AVERAGE\* ROC CURVES**

\*Different percentages of unbalance

**TEST RESULTS**  
**for the Random Forest**

- Hypothesizing 5-95% unbalance in the target
- Selecting a probability threshold of 0.2

Balanced accuracy 99% ↑

Correctly classified  
cases of match  
(True positive rate) 98.5% ↑Erroneously classified  
cases of non-match  
(False discovery rate) 9.5% ↓



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## Conclusions

- The proposed methodology returns promising results to reach the goal with high performance.
- An automated method to detect errors in reported ID codes is necessary to ensure high quality of insurance statistics, given: the need for resilience in DQM processes; the volume of IC assets; the impact that such errors have on compiled statistics.

## Further developments

- Improvement in the model training phase: considering all the available Italian data since 2016, not only focusing on two subsequent quarters.
- Evaluation of model performance on different “asset types” (e.g. securities, deposits, loans).
- Monitoring of production results: cross-check with the insurance corporations during a real data production round.







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*Thank you for your attention.*



Vittoria La Serra

[vittoria.laserra@bancaditalia.it](mailto:vittoria.laserra@bancaditalia.it)

Emiliano Svezia

[emiliano.svezia@bancaditalia.it](mailto:emiliano.svezia@bancaditalia.it)