

Benford testing procedure validation

Andrea Cerasa

**JRC I.3 – Text and Data Mining Unit
Statistics and Information Technologies for Anti-Fraud,
Security and Trade**

Benford's law conference, Stresa, 10-12 July 2019

Objective: study and assess the properties of the Benford testing procedures described in

“Newcomb–Benford law and the detection of frauds in international trade”
*Cerioni, A., Barabesi, L., Cerasa, A., Menegatti, M., & Perrotta, D. (2019).
Proceedings of the National Academy of Sciences, 116(1), 106-115.*

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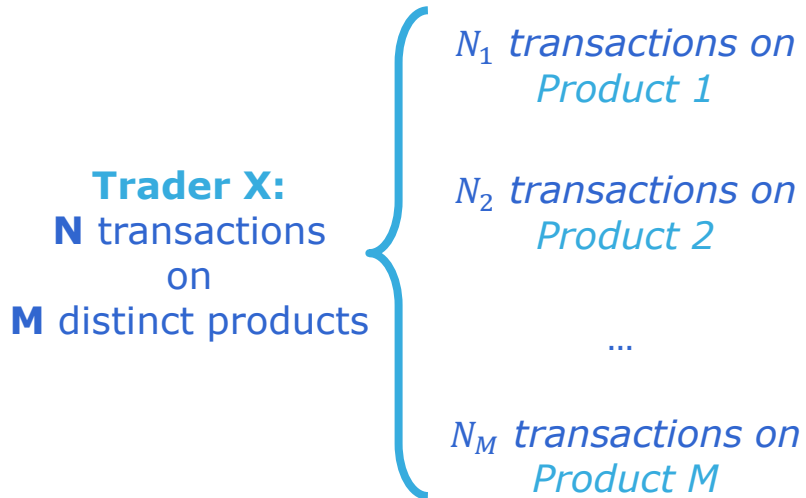
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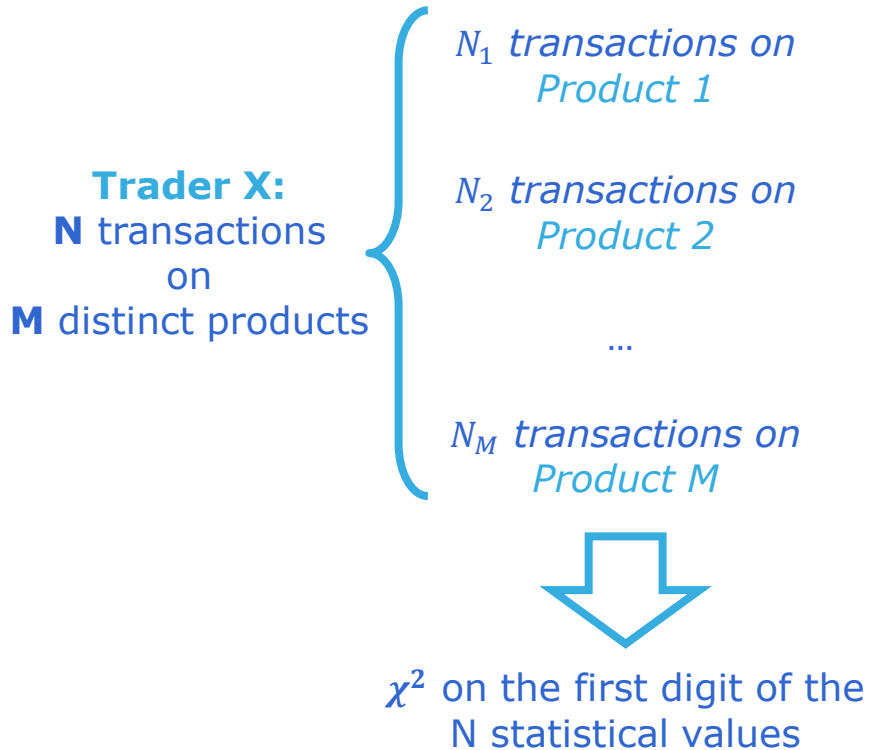
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However, no data were available for a huge validation experiment.

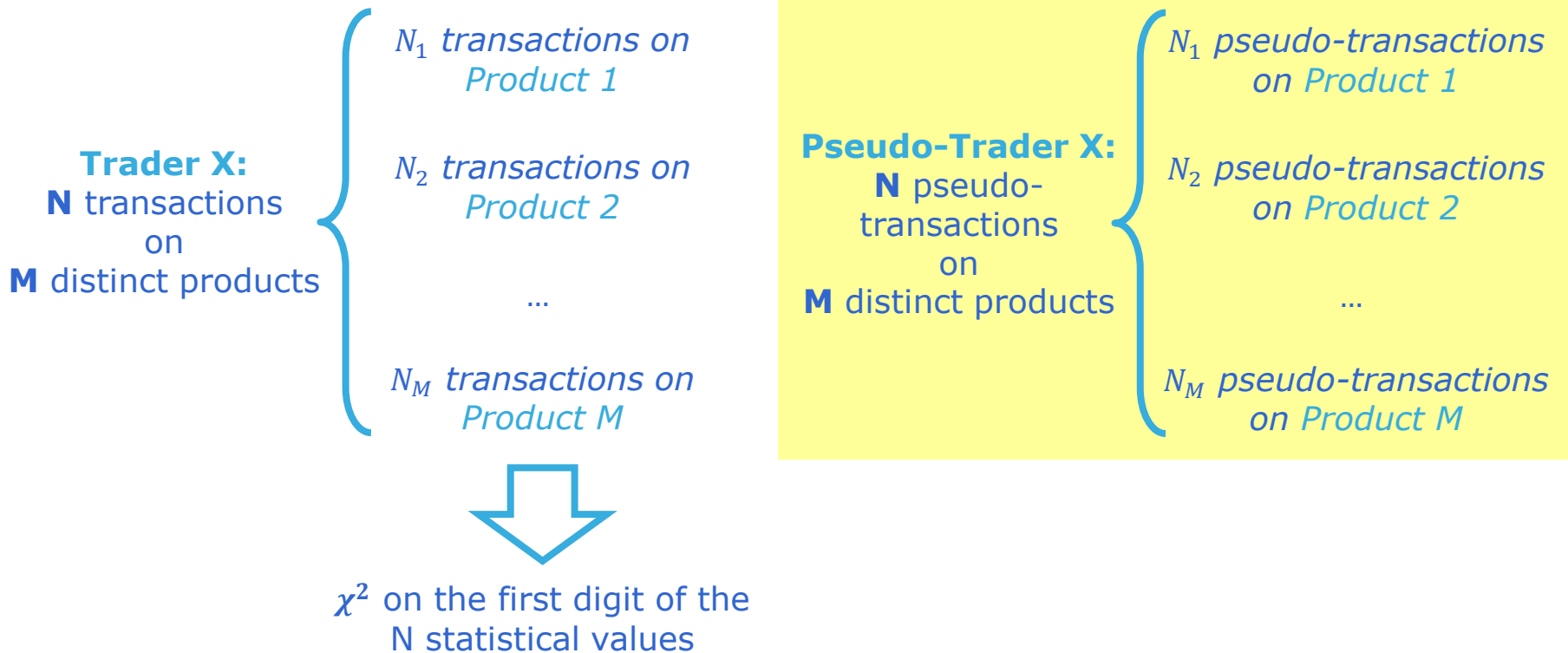
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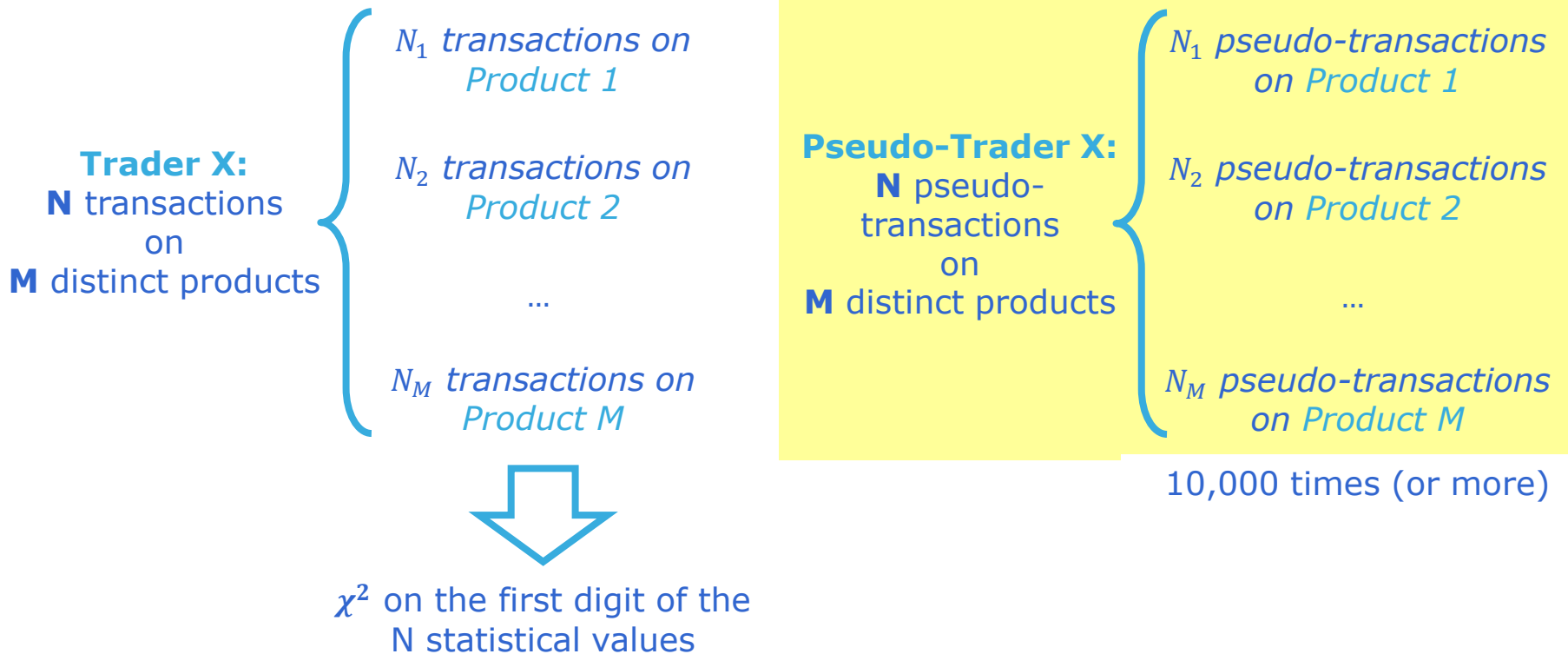
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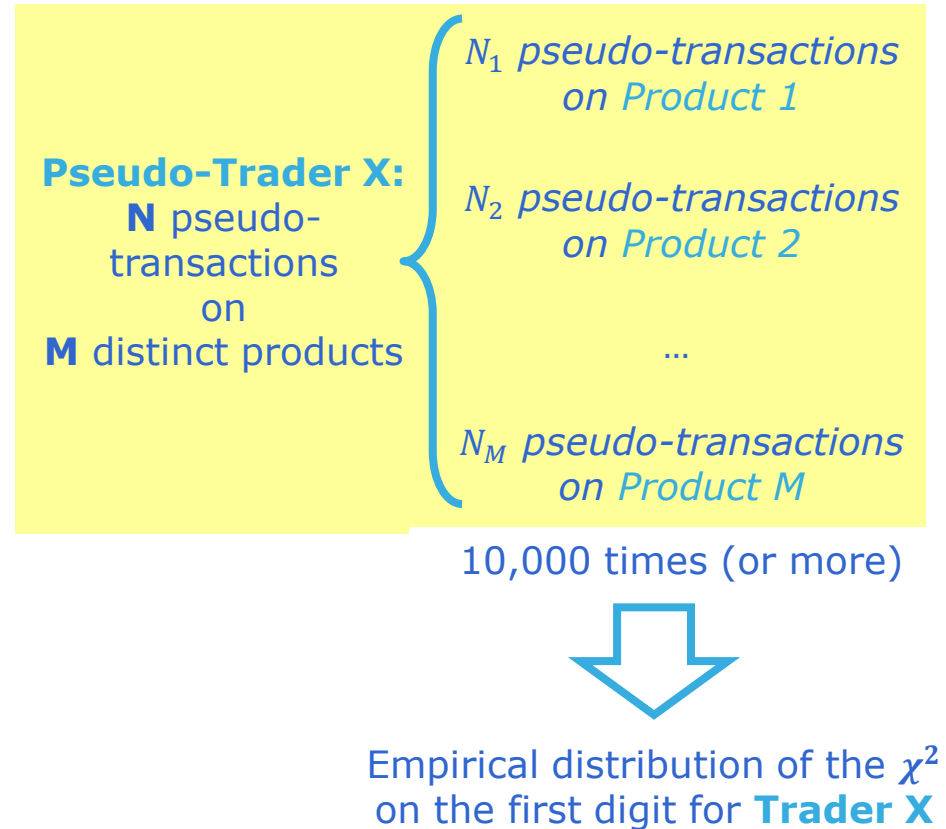
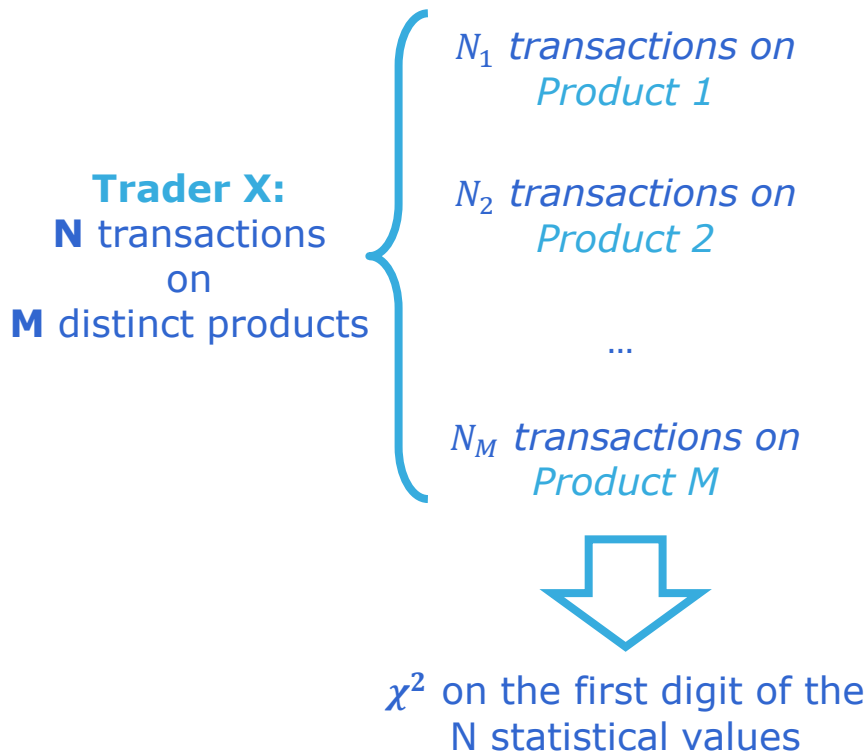
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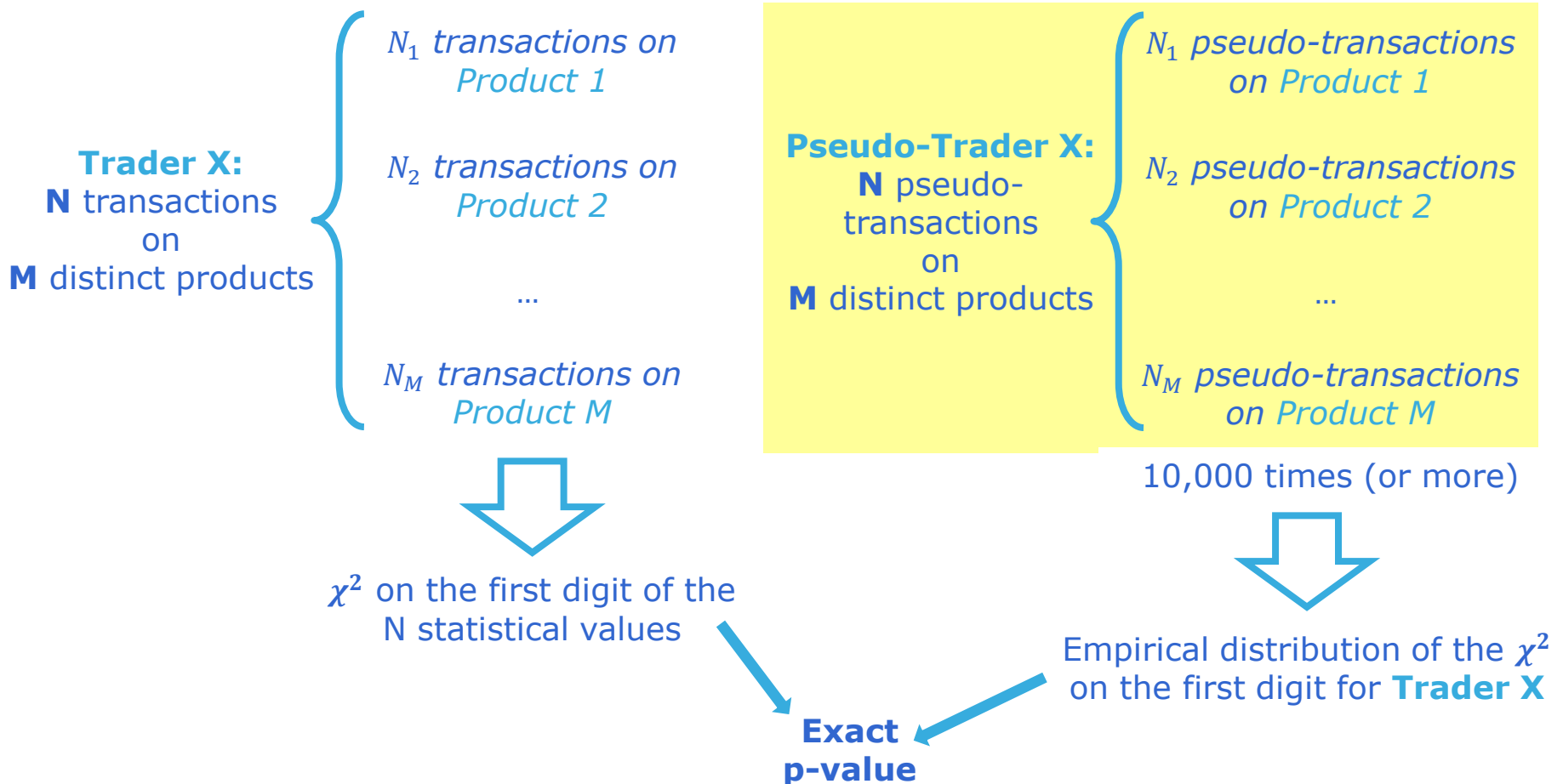
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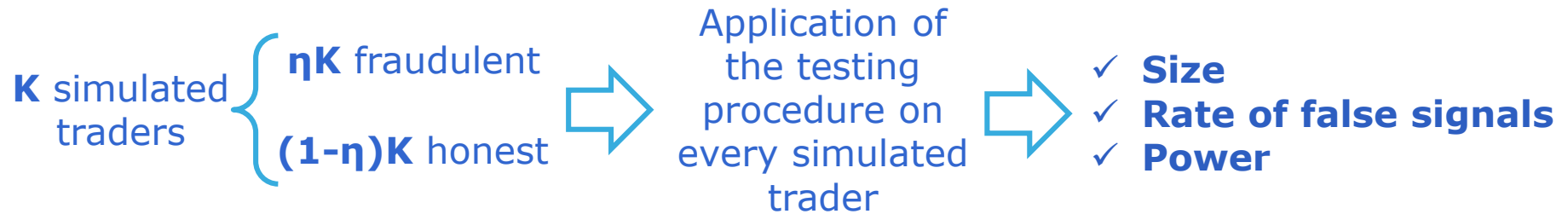
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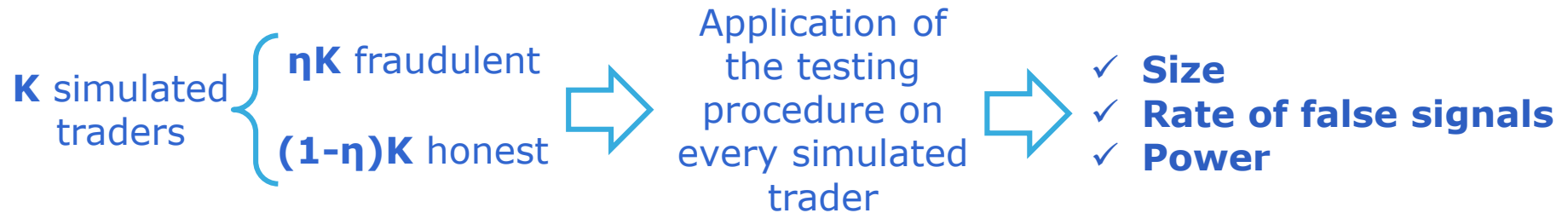
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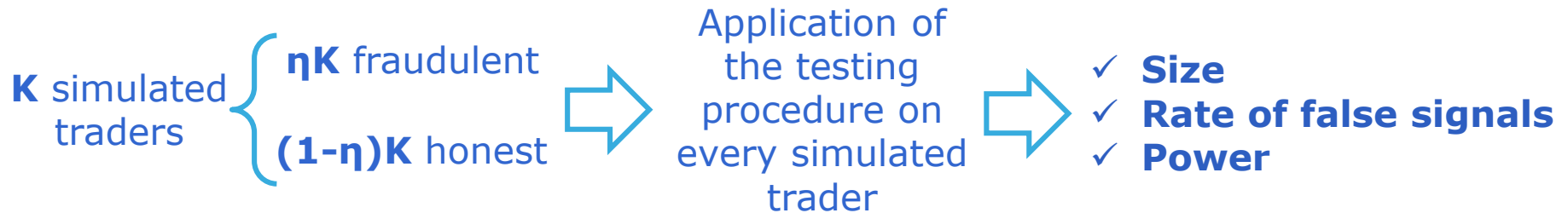


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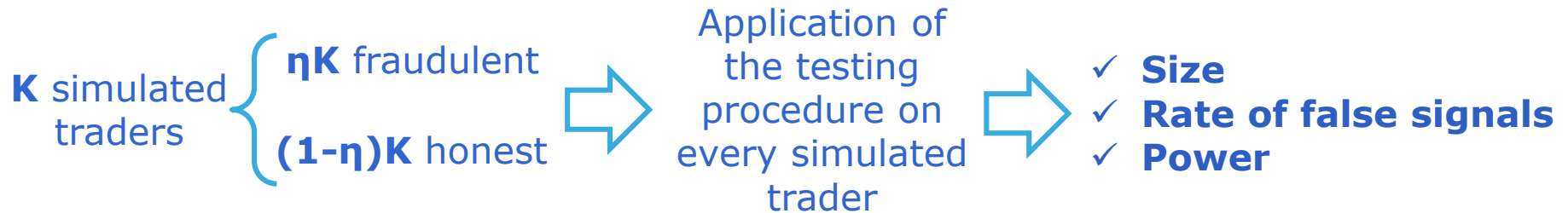
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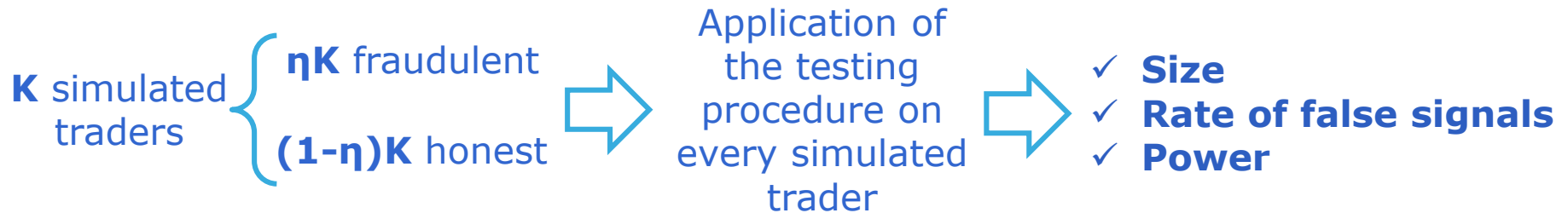
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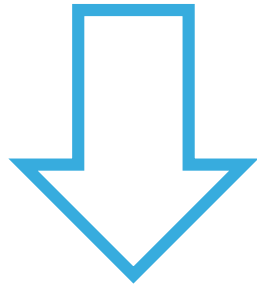
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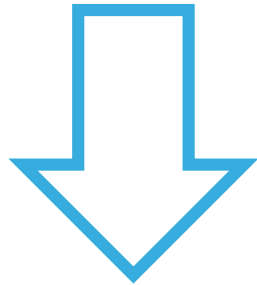


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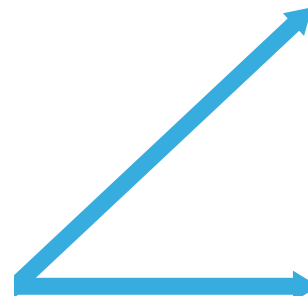
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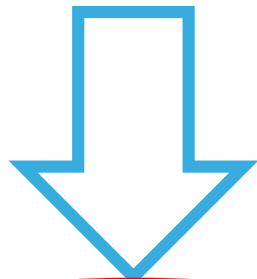
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Our main focus!

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It is difficult to give a reliable definition of an honest trader

Transactions considered	all	9,729,946
Checked companies considered	traders with at least 50 transactions & at least 1 transaction checked	3,856
Definition of honest trader	no fraud detected	3,337
Definition of fraudulent trader	at least 1 fraud detected	519

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Trader A

100 transactions

99 not-checked

1 checked

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Trader A	Trader B
100 transactions	100 transactions
99 not-checked 1 checked	35 not-checked 65 checked
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<i>Regular</i>	<i>All regular</i>	<i>1 under-eval.</i>	<i>50 under-eval.</i>

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Results will then consider different definitions of the honest traders set.

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Improve the reliability of honest traders definition by considering only the traders with a significant fraction (40%) of transactions checked.

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Rate of False Signals: 18.18%
Power: 75.00%

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Size: 8.00%
Rate of False Signals: 18.18%
Power: 75.00%

Nice results but too few traders...

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Focus only on the 55,710 checked records. In this way, it is possible to exactly discriminate honest and fraudulent traders, since all transactions have been checked.

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Checked companies considered	traders with at least 50 transactions	143
Definition of honest trader	no fraud detected	82
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Size: 7.32%

Rate of False Signals: 31.58%

Power: 21.31%

Good results in general.

Low power, but also low fraud intensity (median: 12.5%)

Still high size and high rate of false signals...

Further investigations are required...

Analyzing in details the statistical values declared by the 143 traders, we observed that:

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Therefore we decided to further filter the traders...

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Transactions considered	Only checked	55,710
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Definition of honest trader	no fraud detected	63
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Size: 3.17%
Rate of False Signals: 25.00%
Power: 15.38%

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Size: 3.17%

Rate of False Signals: 25.00%

Power: 15.38%

Size very close to the nominal level (1%)

Improved rate of false signals

Lower power, but also lower fraud intensity (median: 3.9%)

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- When a suitable filter is applied, the size of the testing procedure is close to its nominal level.
- An heuristic filter (i.e. filter 3) guarantees the best results in this sense.
- A refined simulation scheme taking into account similarities and dependences of traders behavior will be object of future studies