

Venezia - December 3, 2024



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The use of ESG data in SME and credit analysis

WP1 - CRIF Work progress



Agenda

1. Introduction

2. Correlation between ESG and lending
3. ESG profile analysis of the Veneto region SMEs
4. The importance and power of data

Introduction



Reference context

In a **political** and **regulatory context**, where the concept of **sustainability** is becoming **increasingly relevant**, **different projects** have been launched relating to the **integration** of **ESG factors** into **credit parameters**

The **ESG issue** represents an **important challenge** and it is **reflected** in a **regulatory framework context** (in particular: EBA Report on E&S Risks, CE Proposal ESG Rating 2023/314, EU Parliament Amendment 2023/314, CRR Amendments)

In order to include **sustainability drivers** (ESG) in **business decisions**, it is crucial to have **accurate** and **robust** information

Objectives

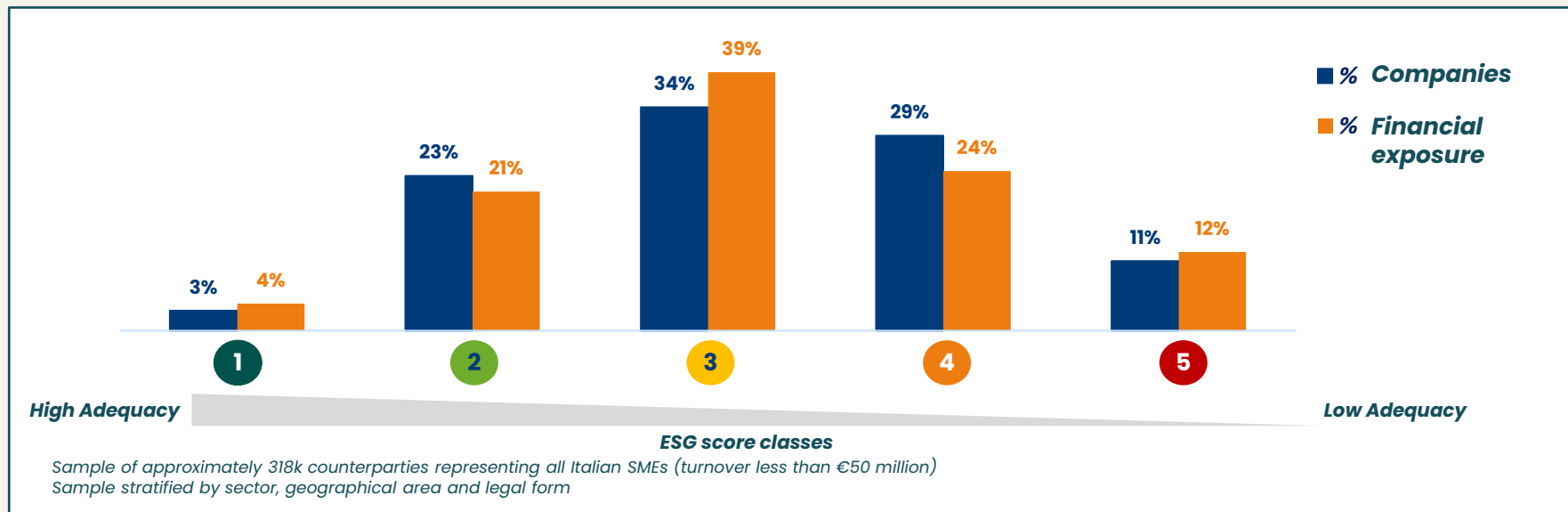
Based on **market evidence** and several **use cases**,  presents its study on the **relationship between ESG and loans**, with a focus on the sustainability profile of **Veneto companies**, highlighting the potential of accurate and granular data. Some of these analyses were also made possible thanks to the collaboration of 

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SMEs stock distribution by ESG score classes

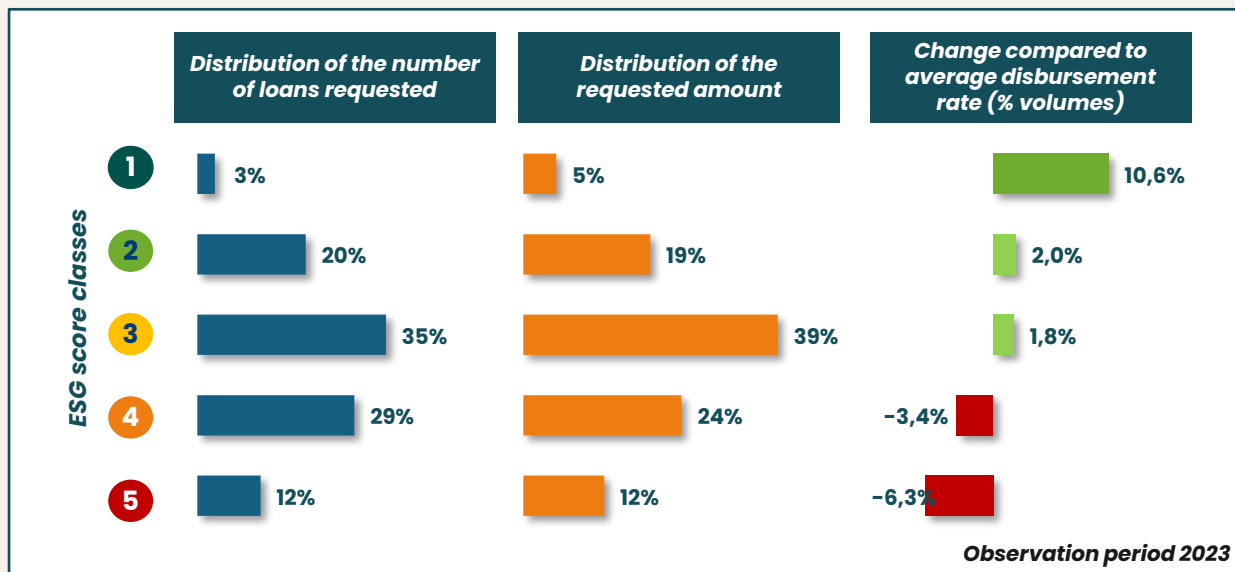
In the **distribution** of the **SMEs analyzed**, approximately **26% of companies** are observed in the **low** and **very low ESG score classes** and **40%** in the **high** and **very high ESG score classes**. Moving from the number of companies to the share of loans disbursed, we observe how the **financial exposures are distributed** in an almost **symmetrical manner**



* These analyses were made possible thanks to the collaboration of RED Risk

Credit policies and ESG score

Observing the **2023 CRIF data** relating to the **disbursement** of **loans** to **Banks** in the Italian system, it is **highlighted** that **companies** with **ESG scores** with **high adequacy** (high or very high score) are more **advantaged** in terms of **access to credit**

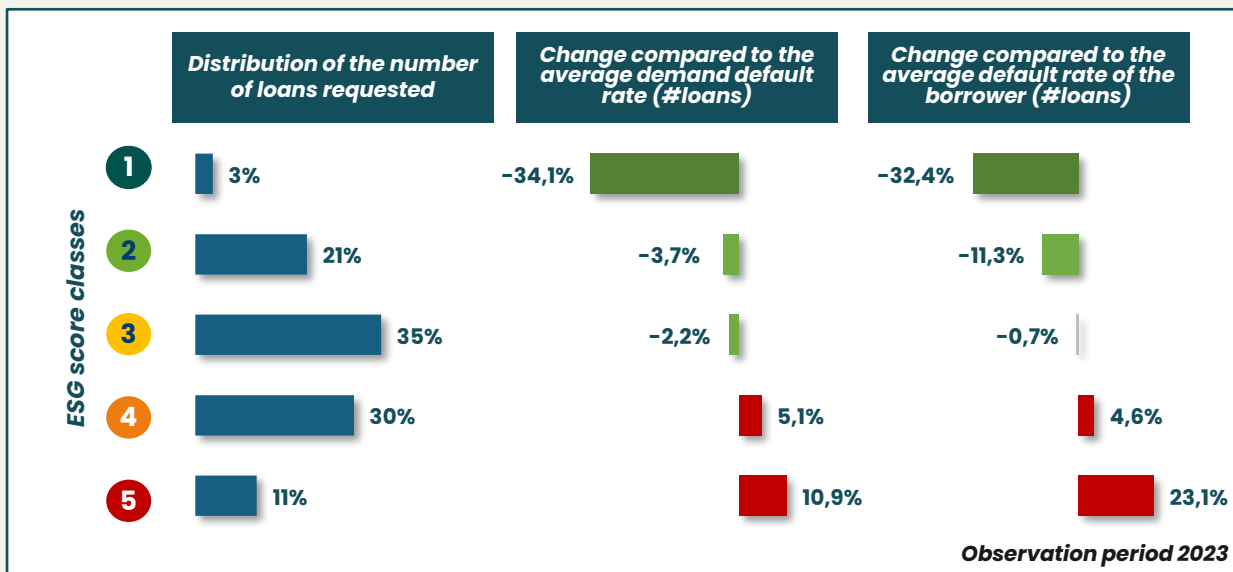


Note

- The **amount requested** is **independent** of the **level of ESG adequacy**
- The **amount paid** is **highly dependent** on the **level of ESG adequacy**

Credit risk and ESG score

Analyzing the **default rates** in the **banking system, companies** with a **high ESG score adequacy** (high or very high score – classes 1 and 2) **appear less risky than the average** of the **sample analyzed**



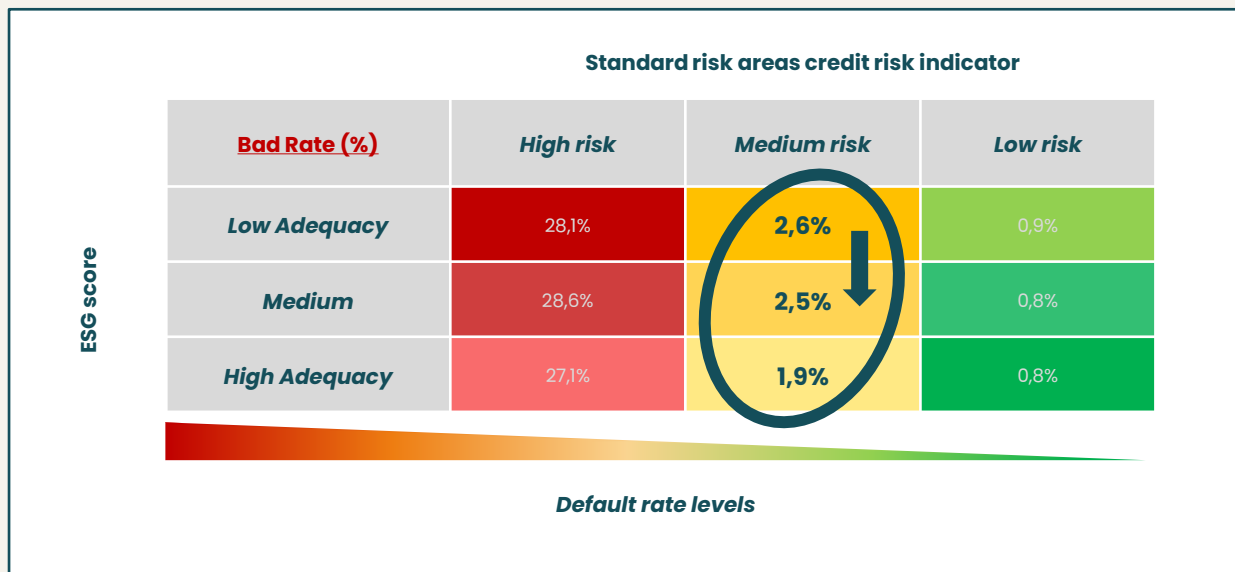
Note

- **ESG score efficiently discriminates the risk level of companies requesting credit**
- **The report is confirmed on the post-acceptance portfolio**

* These analyses were made possible thanks to the collaboration of RED Risk

Correlation between credit risk and ESG score

Below is a **correlation analysis** carried out **between** the **ESG score** and **CRIF credit risk indicator**



Note

- **Strong correlation** between **ESG score levels** and **credit risk**
- The **ESG adequacy levels** enable the **improvement** of the **discrimination capacity** by **streamlining** the **sorting** of the **population in line** with the **credit risk**



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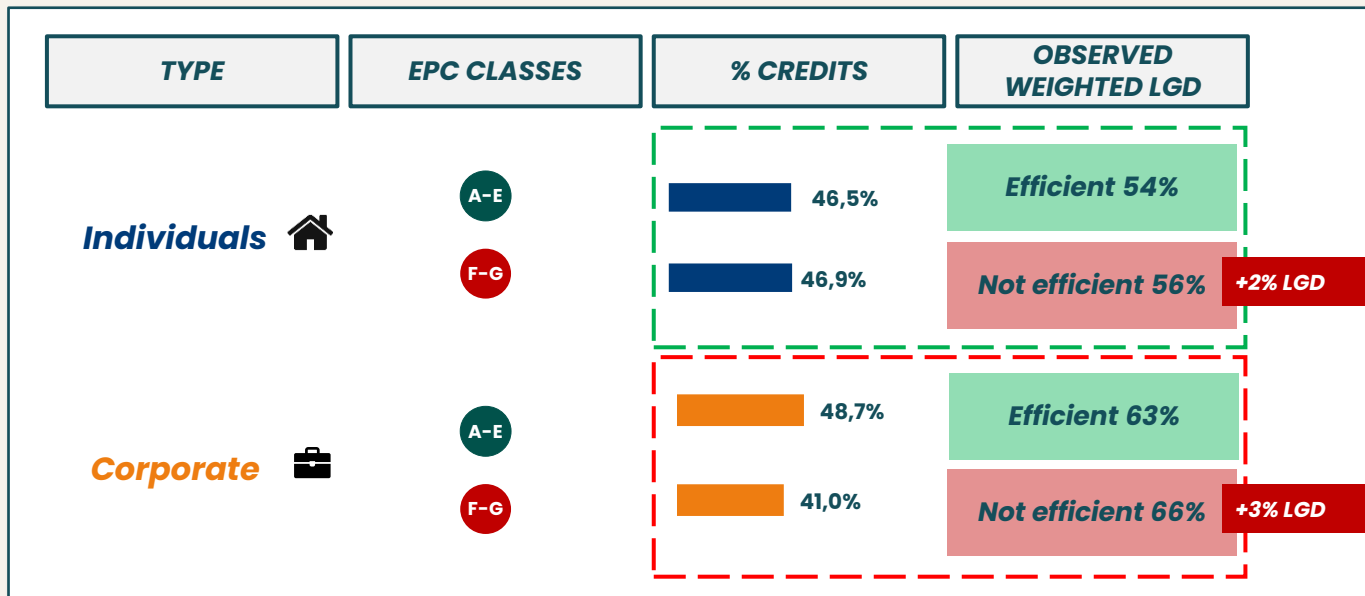
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Some use cases...

Correlation of energy efficiency classes and LGD

USE CASE 1

Among the **counterparties analyzed**, both **individuals** and **businesses**, there is a **strong correlation** between the **levels of energy efficiency (EPC)** and the **estimates of expected losses (LGD)** on the **Bank's NPL portfolio**

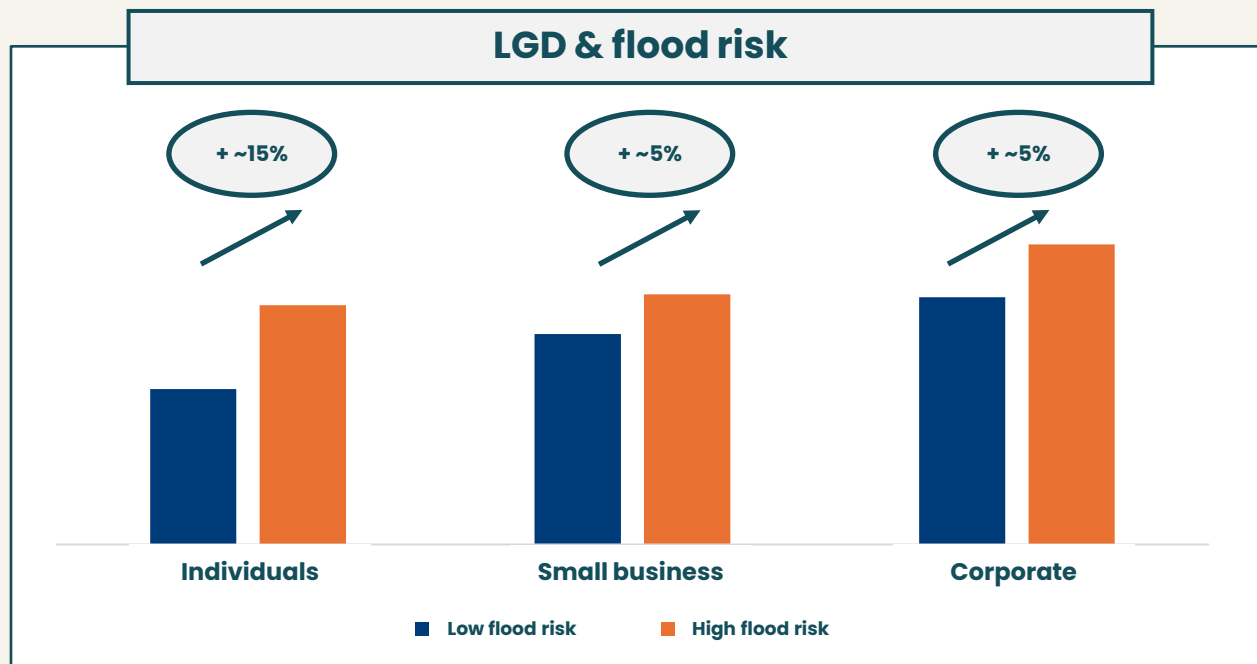


Note

- **Strong correlation** between **estimates of expected losses (LGD) on credits** and the **energy efficiency levels of collateral**
- **Energy-efficient collateral generates low loss estimates** on the portfolio

Correlation of flood risk and LGD

USE CASE 2



Note

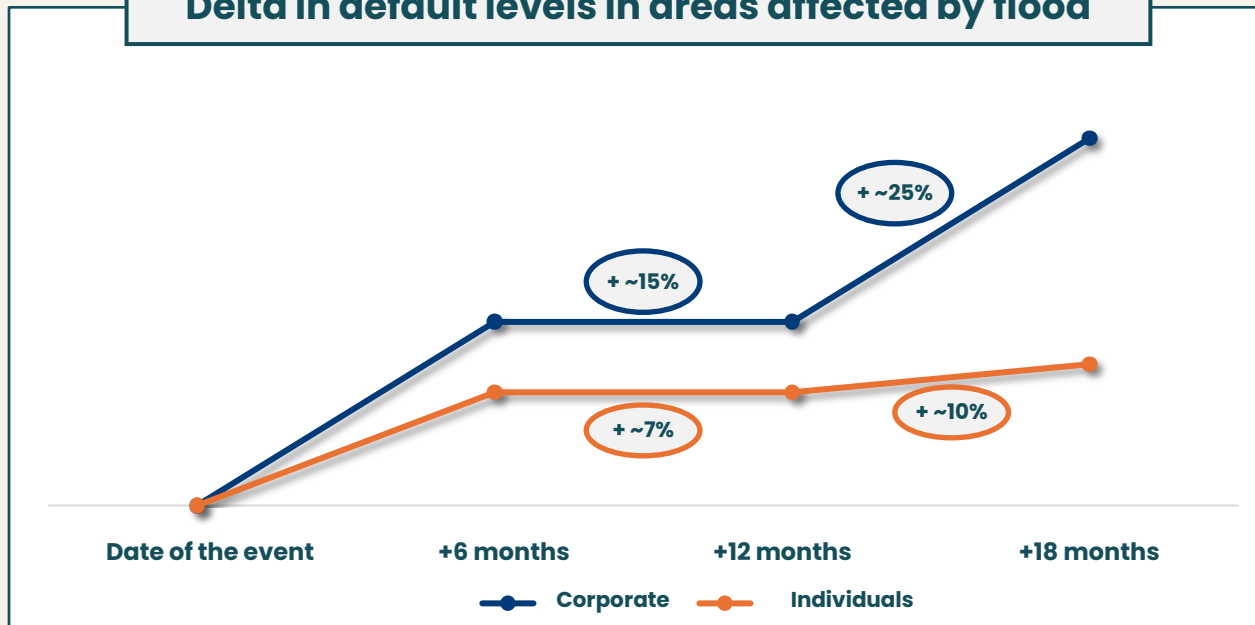
- Exposures with **low physical risk** (flood risk) **generate reduced loss estimates** for the portfolio
- The **correlation** between **ESG risks** and **credit risk** is starting to be **significant**, **enabling integration into regulatory and management LGD parameters**

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Correlation of flood risk and PD

USE CASE 3

Delta in default levels in areas affected by flood



Note

- The **correlation** between **ESG risks** and **credit risk** is starting to be **significant**, enabling integration into regulatory and management PD parameters

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USE CASE 4

Correlation of transition risk and PD

The **discriminatory power** of the **transition risk variable** respect to the **probability of default** of the **Bank's business portfolio** is reported, **comparing** the **impacts** of the **different stress scenarios** on the **Bank's lifetime PDs**



Note

- Counterparties in classes 4-5 present a higher level of default risk
- Strong correlation between transition score levels and the probability of default
- The impact of the delta scenario on PD is minimal in score classes 1, 2 and 3 while it is more intense in bands 4 and 5

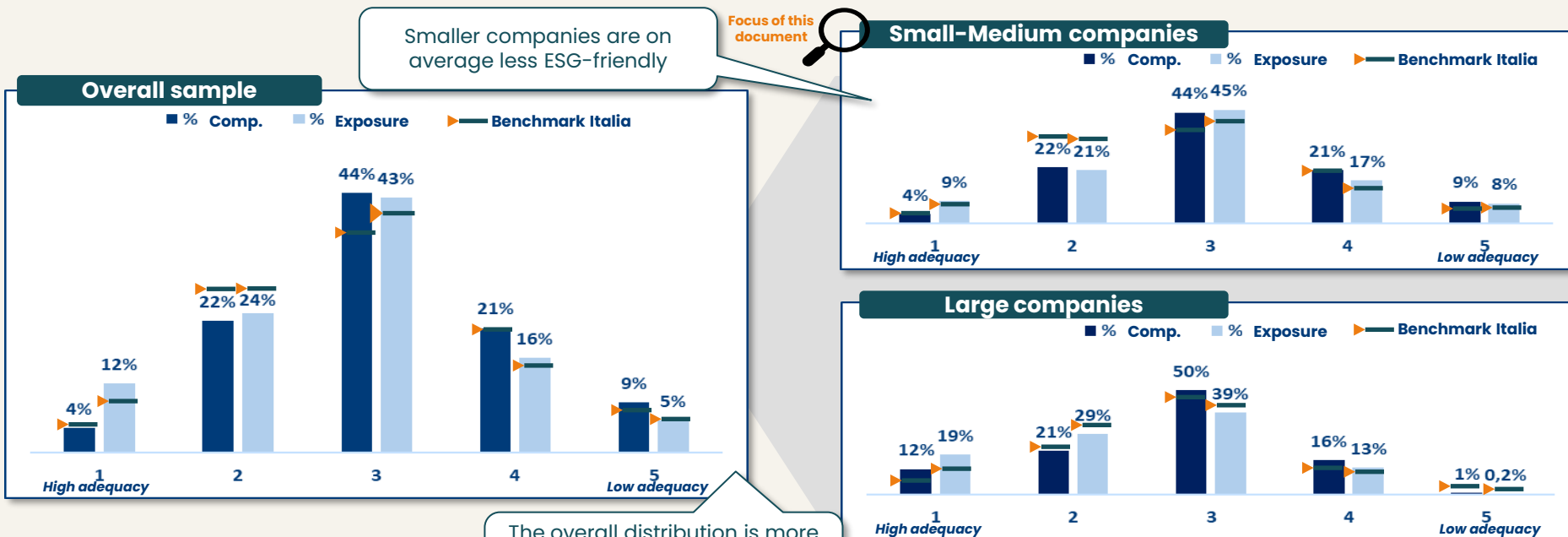
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ESG Score

The distribution of the **ESG score** over the **number of companies** and **financial exposure** value is shown below

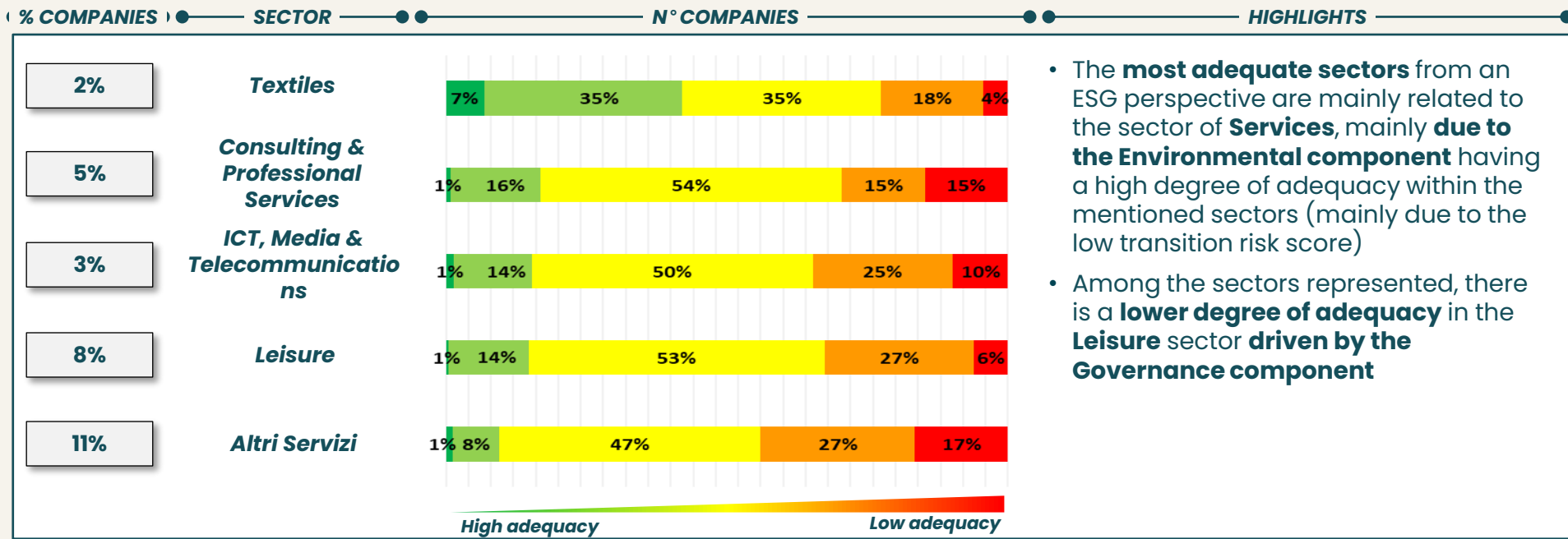


The overall distribution is more influenced by the behaviour of SMEs than large enterprises

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ESG Score – Distribution by Sector

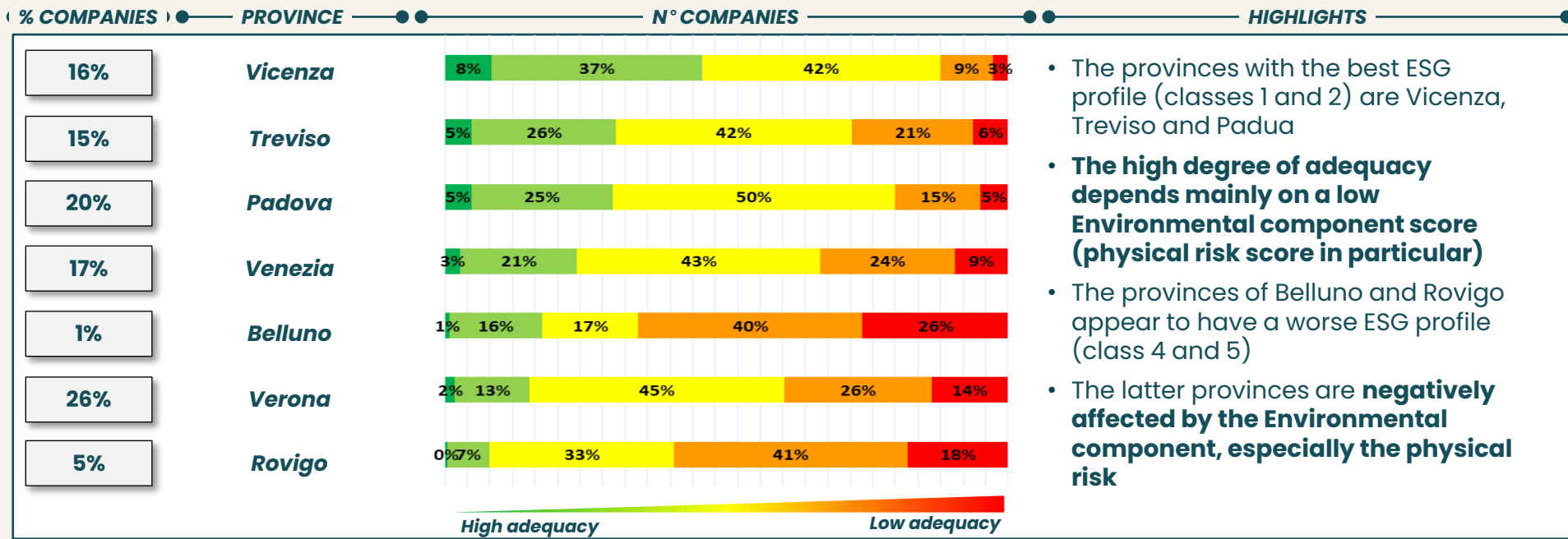
The **ESG score** distribution of the **five sectors with the best adequacy** profile is shown below



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ESG Score – Province Distribution

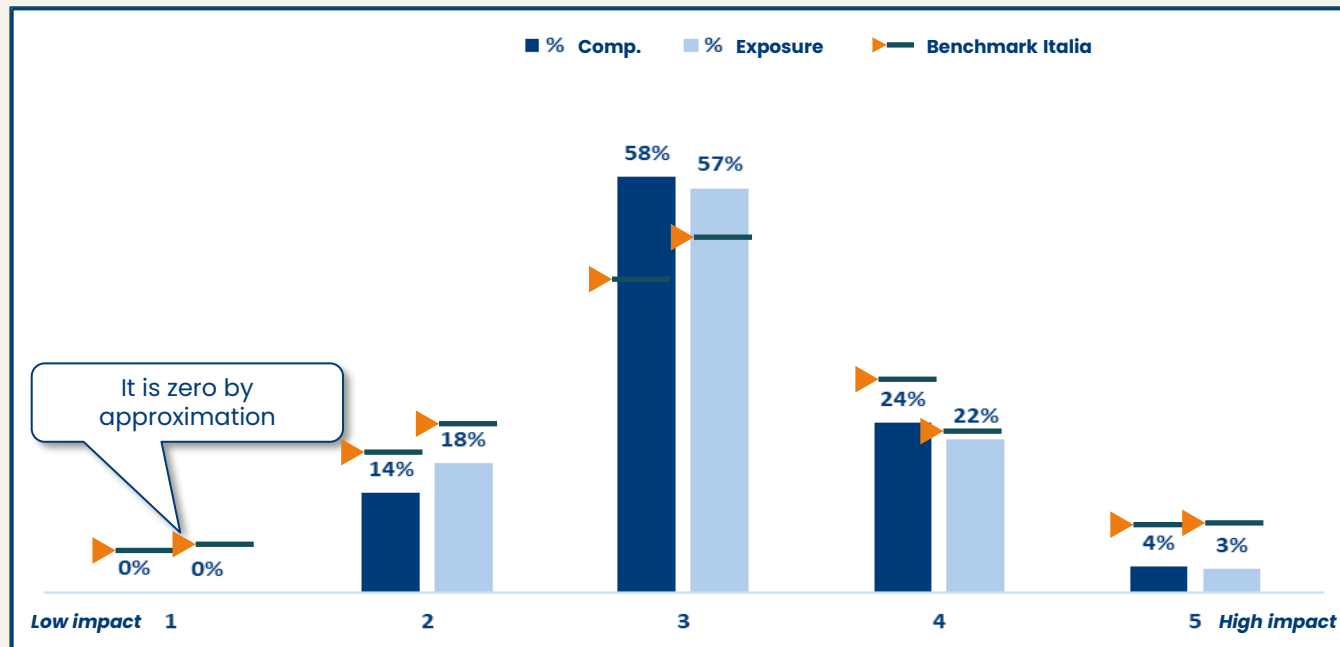
The **ESG score** distribution of the **Veneto provinces** sorted by **highest adequacy** values is shown below



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Physical Risk Score

The **ESG score distribution** of the Veneto provinces **sorted by highest adequacy** values is shown below



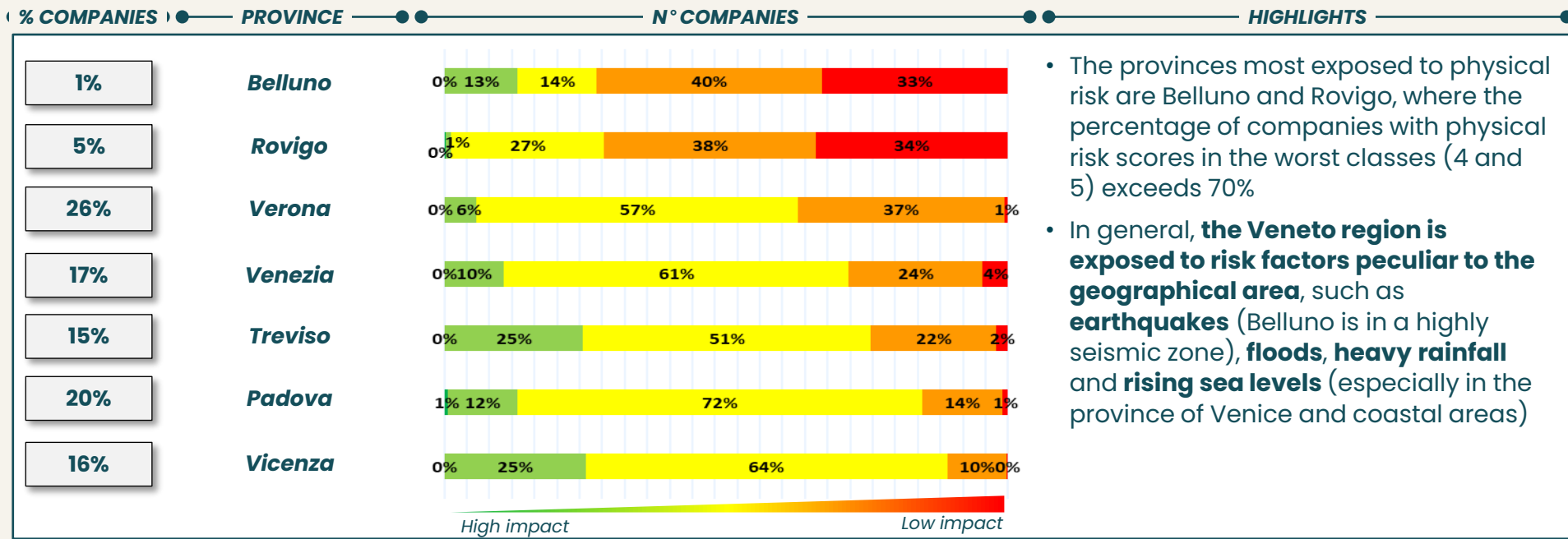
HIGHLIGHTS

- 28% of counterparties is affected by high physical risk (classes 4 and 5)
- **Compared to the Italian market, companies in the Veneto region are less affected by high physical risk, with a higher concentration in the middle class**
- For those companies most exposed to physical risk, a **path to awareness of these components through adaptation activities must be undertaken**, so making the expected structural impacts less severe

* These analyses were made possible thanks to the collaboration of RED Risk

Physical Risk Score – Province distribution

The distribution of the **Physical Risk score** per province **sorted by highest risk** values is shown below

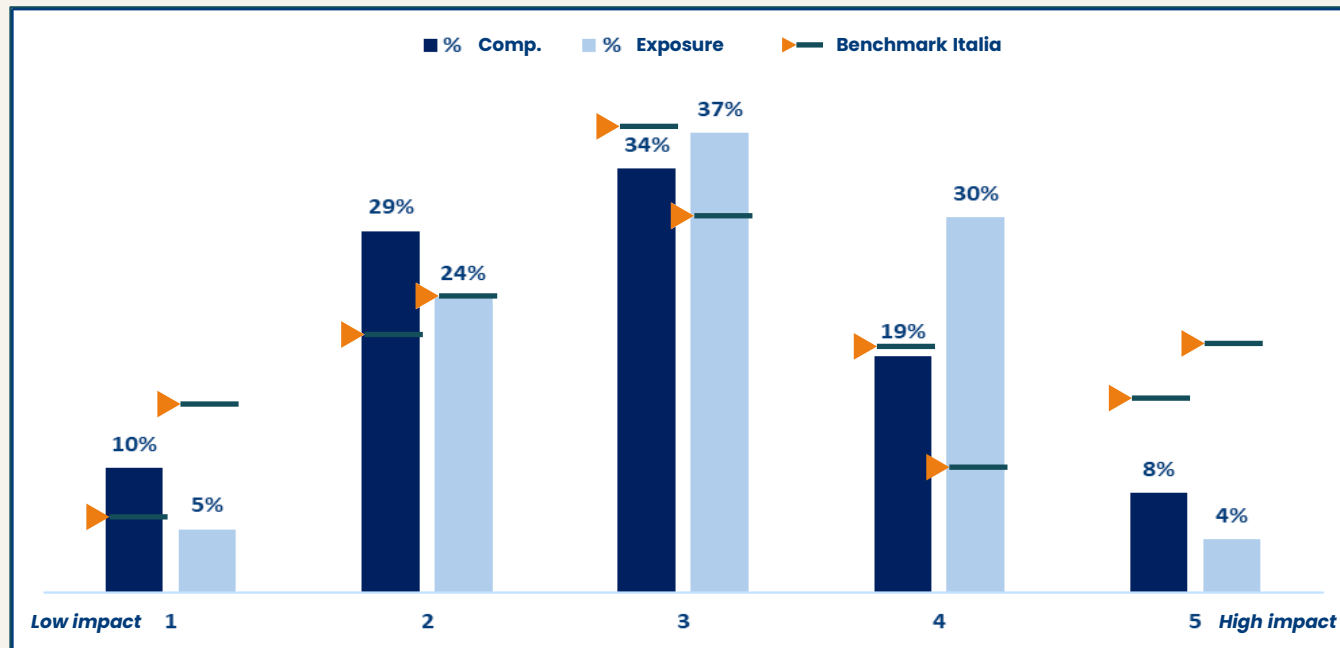


- The provinces most exposed to physical risk are Belluno and Rovigo, where the percentage of companies with physical risk scores in the worst classes (4 and 5) exceeds 70%
- In general, **the Veneto region is exposed to risk factors peculiar to the geographical area**, such as **earthquakes** (Belluno is in a highly seismic zone), **floods, heavy rainfall** and **rising sea levels** (especially in the province of Venice and coastal areas)

* These analyses were made possible thanks to the collaboration of RED Risk

Transition Risk Score

The distribution of **Transition Risk score** over the **number of companies** and **financial exposure** is shown below



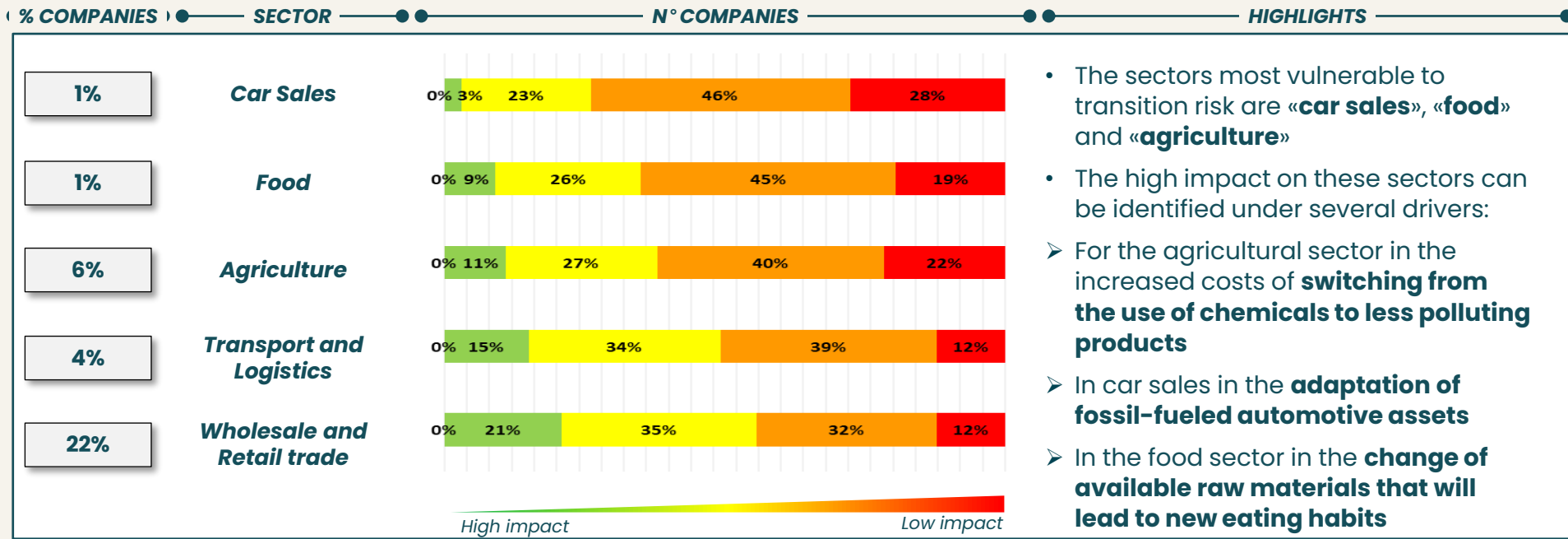
HIGHLIGHTS

- The transition score shows a mostly counterparty and exposure in the lowest impact classes (1,2,3)
- **Compared to the market, Veneto has fewer companies in the highest impact classes (4, 5)**
- As expected, **companies most exposed to transition risk need more investment to cope with the sustainable transition process**
- From the figure, it can be seen that financial debt is one of the ways taken to deal with this type of investment

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Transition Risk Score – Distribution by Sector

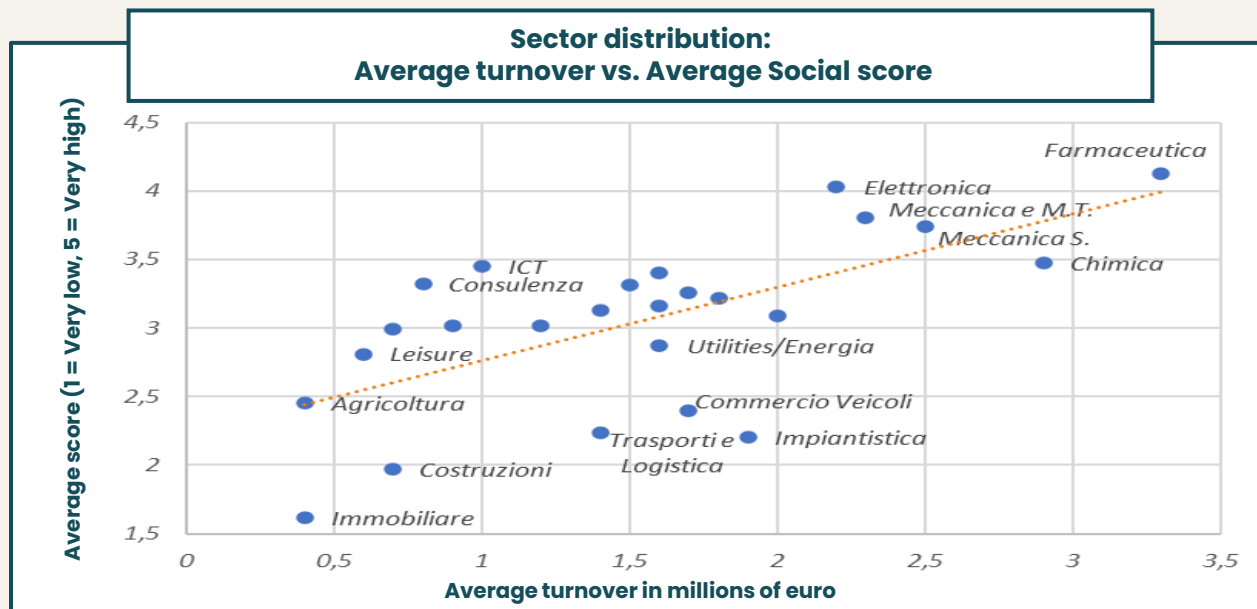
The distribution of **Transition Risk score** of the **five most exposed sectors** is shown below



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Social factor vs. Size

The **Social factor** seems to go **in parallel** with the **size of SMEs**: as the size increases, the focus on social issues increases



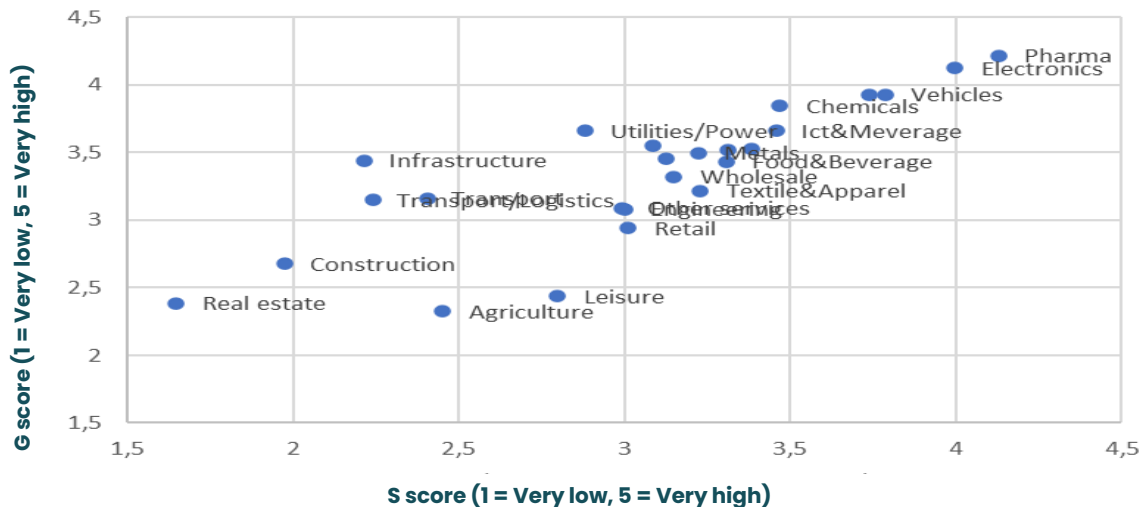
HIGHLIGHTS

- Sectors such as **ICT** (Information and Communications Technology), **Consulting** and **Electronics** are **more socially aware than average**
- Sectors such as **Utilities** and **Transport**, although on average larger, **tend not to have developed targeted actions on social issues**

Governance factor vs. Social factor

The **Governance factor** goes **hand in hand** with the **Social factor**

Sector distribution: S score vs G score



HIGHLIGHTS

- **Companies with higher turnover have more structured governance**
- **Larger companies are often structured to provide external disclosure**, including through legality ratings, certifications and financial declarations on a voluntary basis

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The importance and power of data

In order to make **informed choices and effective actions to include ESG drivers in business decisions**, it is critical **to rely on accurate, robust, and granular data analysis** (as in the examples above).
The following slides provide some explanatory examples of the importance of these features

To contribute to the study of the phenomena and related economic relationships, **CRIF is working to provide** the project with **a complete and granular sample** (including **ESG and Business Information**)

The **sample** will consist of the following

- Approximately **150,000 SMEs**
- **Representative of the Italian framework**
stratified to respect Italian distribution according to the drivers
 - i. Sector (NACE)
 - ii. Geographical Area
 - iii. Turnover Size

Comparison with open data – Physical Risk

Evidence on the **riskiness of the Romagna territory** from other **public** (ISPRA) and **non-public data sources** is reported. The data reported refer to the **municipalities analysed** (Alfonsine, Cesena, Cervia and Faenza)

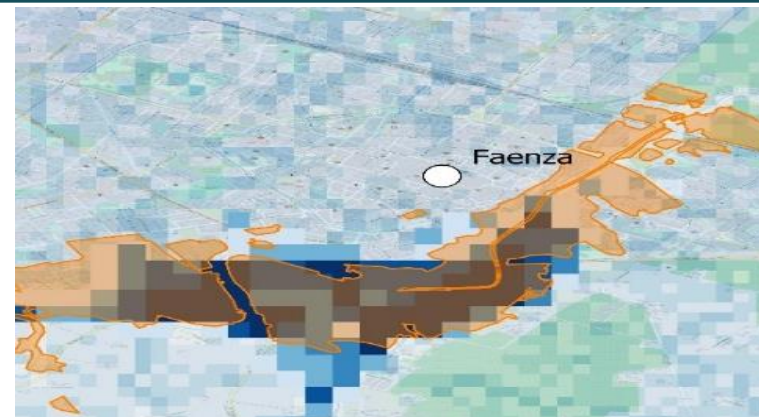
Other providers



Source: [ISPRA data](#)  Medium risk  High risk

~67% properties at High/Medium risk

CRIF Data 90X90

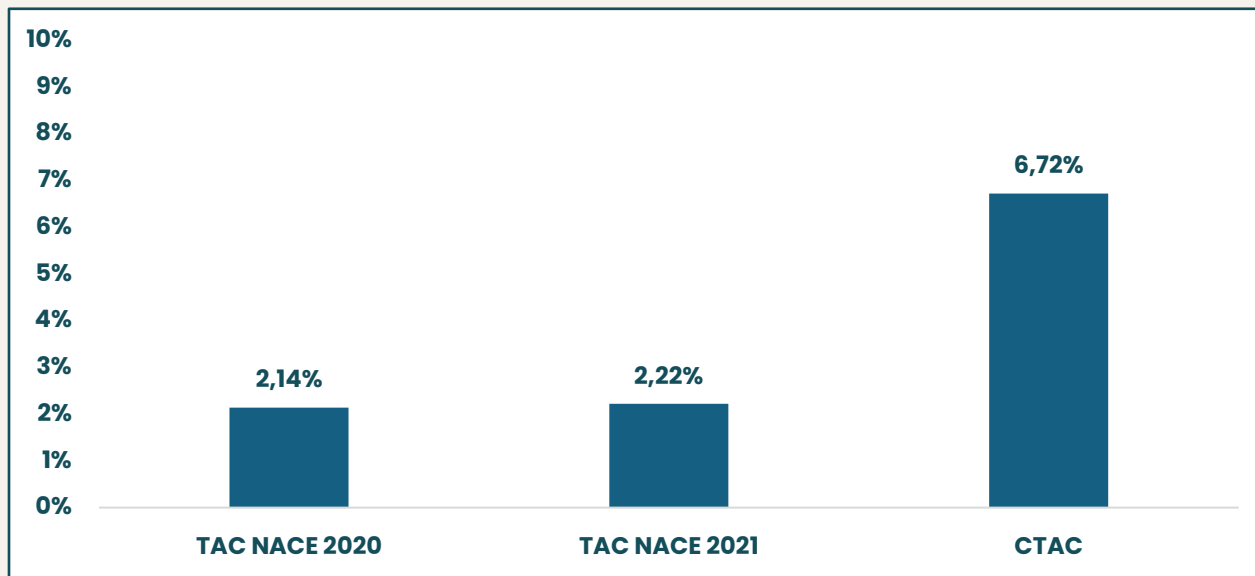


Illustrative image. The % is for the 5 municipalities

~8% properties at High/Medium risk

Comparison with open data – Taxonomic alignment (1/3)

A **comparison** of the **average alignment** calculated on the sample of **Italian SMEs** with the **CRIF TAC** (Estimated Taxonomic Single Name Alignment Value), **TAC NACE 2020 and 2021** is shown below

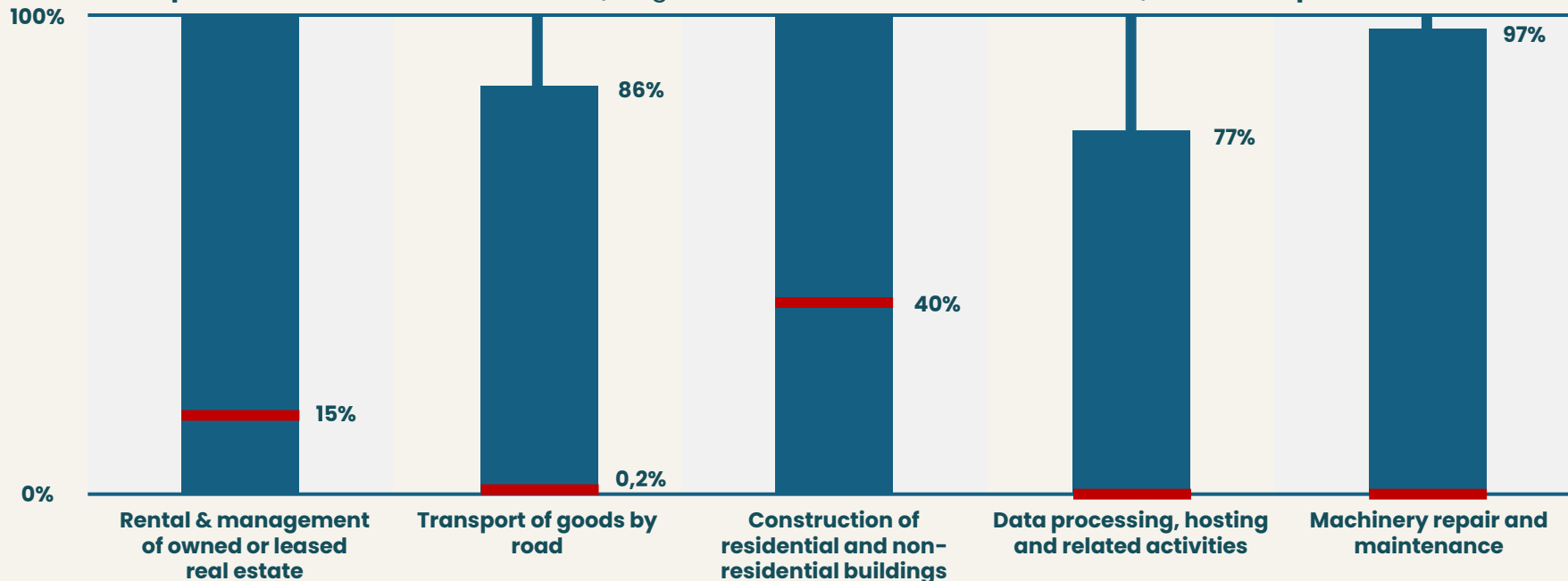


HIGHLIGHTS

- **Applying C-TAC** results in an average alignment of SMEs **three times higher than using the NACE 2020 and 2021 TACNB**: the analysis does not take into account the exposure associated with special purpose technical forms (i.e. mortgages)

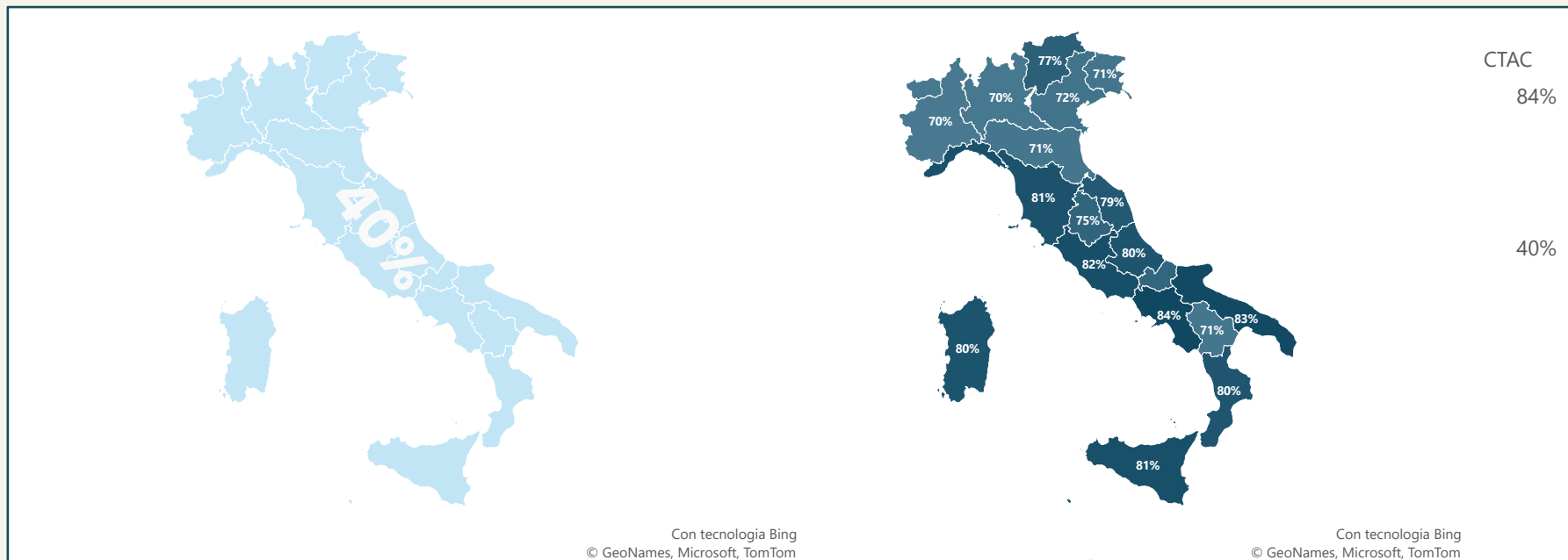
Comparison with open data – Taxonomic alignment (2/3)

A comparison of NACE 2021 and CTAC (range between minimum and maximum) for some specific sectors

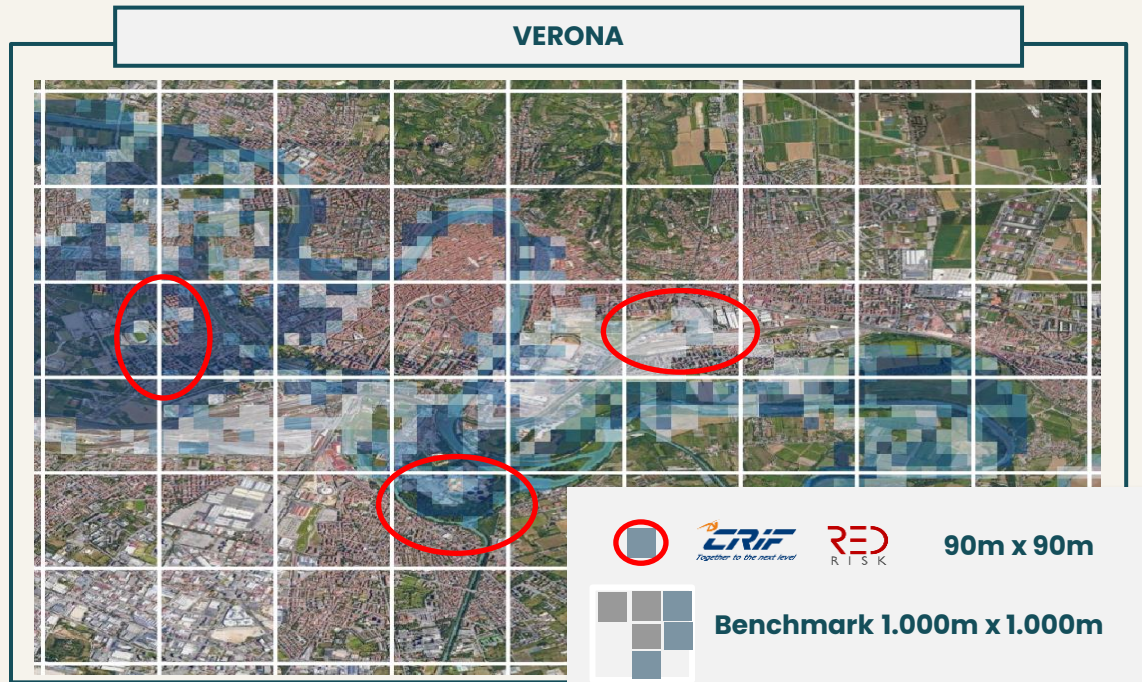


Comparison with open data – Taxonomic alignment (3/3)

Regulatory TAC (left) and CTAC (right) associated with SMEs in the Construction sector in the different Italian regions



Comparison with less granular benchmark – e.g. flood risk (1/2)



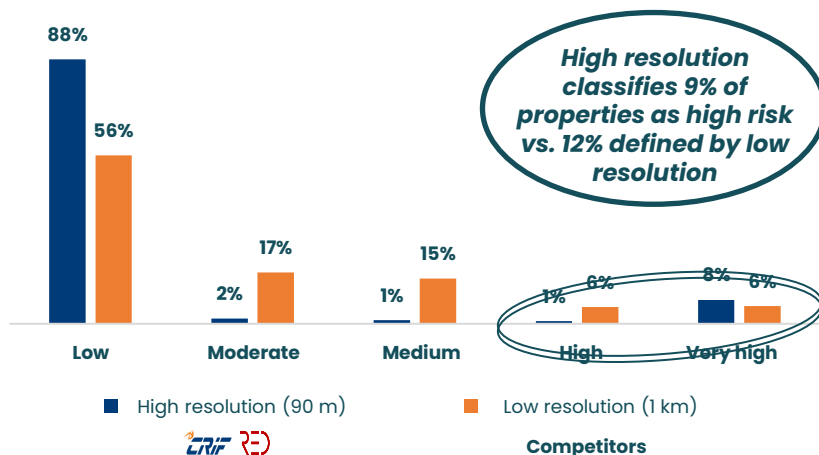
RESOLUTION

The map represents **water heights associated with a flood** with a 100-year return period. The **white pixel** represents **water heights of a few cm** (associated with minor damage to the property), the **dark blue heights of more than 3 m** (associated with major damage, such as damage to the house installations, ...). The 1,000m x 1,000m resolution grid incorporates micro-areas with very heterogeneous risk levels. Moreover, **within the 'pixel' properties tend to be concentrated in the areas of lowest risk**, accentuating the distortion associated with average ratings

Comparison with less granular benchmark – e.g. flood risk (2/2)

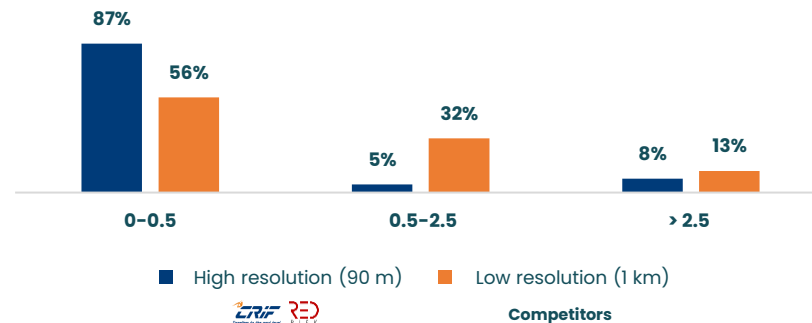
On the analysed portfolio, the application of a high-resolution model is estimated to result in a **reduction in expected losses of between €30k and €60k***.

Distribution by Physical Risk classes



Loss distribution

Normalized average annual losses (per thousand)



Average annual loss 89 € — 2x → 197 €

* These analyses were made possible thanks to the collaboration of RED Risk
*Guaranteed portfolio of about EUR 1m EAD, PD at 0.7% and LGD at about 14%. Simulation conducted by stressing only LGD for the impact of physical risk

Closing remarks

Data quality evolves compliance in «effective» risk management

The ESG profile of SMEs affects credit risk

It is a long journey, but we have shown that there is room for improvement

Venezia – December 3, 2024



Thank you

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