

ENGAGE for ESG: a simple solution for ESG reporting

Thursday, 30 October 2025







Today's Speakers



Marco Angheben Head of Business Development & Regulatory Affairs European DataWarehouse ENGAGE Project Coordinator



Raffaella Calabrese Chair of Statistics and Data Science University of Edinburgh Business School

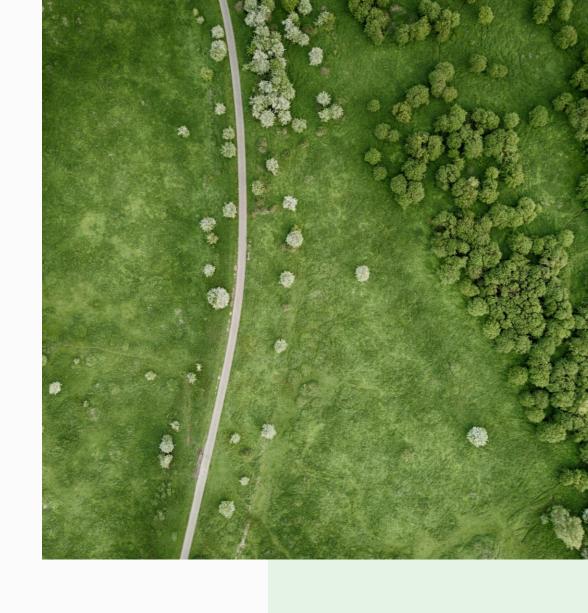


Vincent Mahieu Head of Energy Efficiency & Regulatory Technology Hypoport



Agenda

- I. Welcome and introduction
- II. Climate stress testing for mortgage default probability
- III. The ENGAGE Templates version 1.2 and the ENGAGE Portal
- IV. How to get involved in ENGAGE for ESG





I. Welcome & Introduction

Marco Angheben

European DataWarehouse





ENGAGE for ESG initiative



- Co-funded by the European Union with a LIFE grant
- Six consortium partners from across Europe
- Scope: ENGAGE for ESG aims to provide:
 - A. A simple solution for ESG reporting
 - B. For mortgages and home renovation loans to check alignment with:
 - 1. the **EU Taxonomy Regulation**;
 - 2. in compliance with the Energy Performance of Buildings Directive; and
 - 3. the General Data Protection Regulation.















ENGAGE for ESG: a comprehensive solution

"We do not know what data is needed for EU Taxonomy Alignment"

"What data is needed to gain insights into the renovation potential?" "How does this portfolio compare to other portfolios?"

"Which portfolio has the best renovation performance?" "We have a mandate to fund EU Taxonomy aligned renovation investments but we have a challenge in identifying eligible investments"

Physical risks

Green Asset Ratio calculation

Transition risks management

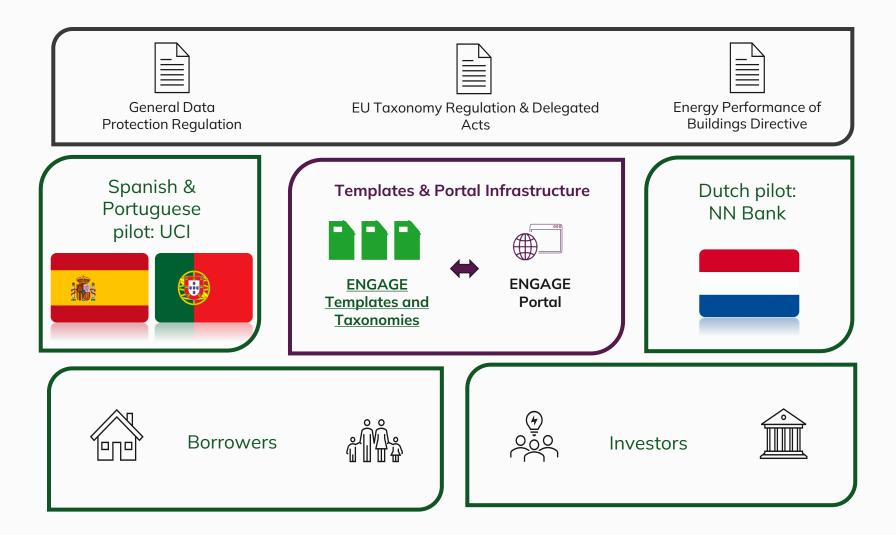




ENGAGE for **ESG** initiative description

Two main pillars:

- A disclosure template for residential real estate data;
- A portal to assess EU Taxonomy alignment.







The ENGAGE Templates: structure

Building block 1 delivered

Building block 2 delivered

Building block 0 • Current ESMA underlying exposures for residential mortgages (Annex 2) and consumer loans (Annex 6)

Building block 1

- Compliance with the EU Taxonomy Regulation for:
- a) Mortgages for the acquisition of new buildings and existing buildings;
- b) Loans for the renovation of existing buildings.

Building block 2

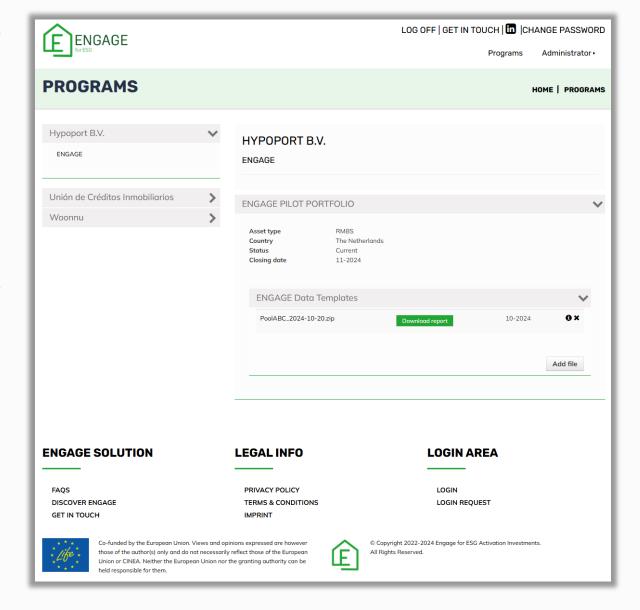
• European Investment Bank information requirements (Group Green Checker) to capture funds from green financing programs for the financing of the renovation wave.





The ENGAGE Portal: high level overview

- The ENGAGE Portal enables lending institutions to upload and assess the alignment of their loan portfolios with the EU Taxonomy.
- The Portal incorporates detailed loan and collateral level checks.
- Access to multiple stakeholders (internal reporting, rating agencies, investors, regulatory supervisors, etc.) upon invitation.
- The ENGAGE Portal allows users to assess multiple portfolios based on the EU Taxonomy.
- The Portal output is twofold:
 - 1. The data quality feedback;
 - The EU Taxonomy alignment report.
- Free access available subject to a legal arrangement.





New ENGAGE user onboarded!

- The Belgium-based social lender Onesto joined the ENGAGE community in October 2025, becoming a user of the ENGAGE Templates and Portal.
- Joris Piette, Project Manager and COO at Onesto, stated: "By working with the ENGAGE Templates and Portal, we gain a crucial puzzle piece in our mission to make sustainable homeownership attainable for low-income first-time buyers. The structured ESG data and Taxonomy insights from ENGAGE allow us to connect impact measurement to real people's housing journeys ensuring that sustainability and affordability go hand in hand."



Belgian Lender Onesto Joins ENGAGE for ESG to Support Green and Affordable Homes

RANKFURT - 21 OCTOBER 2025

The ENGAGE for ESG Consortium is pleased to announce that **Onesto**, a Belgium-based social lender committed to sustainable and affordable housing, has joined ENGAGE for ESG as a trial user.

From October 2025, Onesto will work with the ENGAGE Templates and Portal—innovative tools developed under the EU-funded ENGAGE for ESG project—to assess the Taxonomy alignment of its mortgage and home renovation loan portfolio in a systematic and structured manner. As part of this process, Onesto will begin uploading credit and sustainability information to the ENGAGE Portal, further enriching the ESG data available on the platform.



II. Climate stress testing for mortgage default probability

Raffaella Calabrese

University of Edinburgh



Climate stress testing for mortgages

Luca Zanin - Prometeia

Raffaella Calabrese - University of Edinburgh

Connor Innes Thorburn - Prometeia

Agenda

- Introduction
- Literature review on climate stress testing
- Purpose
- Methodological framework
- Data collection
- Model of credit scoring
- Design of scenarios
- From damage index to climate-adjusted LTV
- Stress test for mortgage PD
- Conclusions



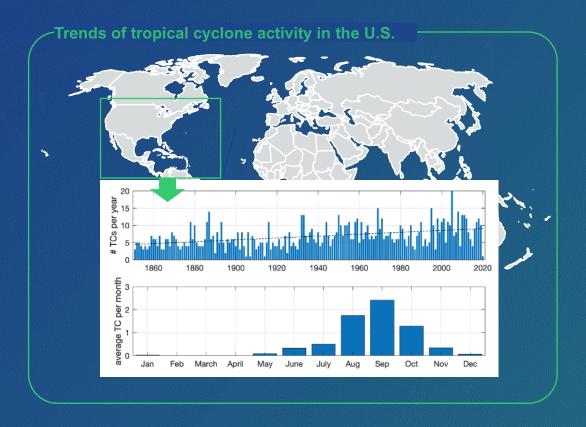
Introduction



Extreme weather events, such as tropical cyclones, floods, and heavy rains, occur more frequently and intensely in the United States than in the past owing to global warming (IPCC, 2023)



Policymakers, central banks, regulatory authorities, and practitioners have expressed concerns about the potential risks that extreme weather events may pose to economic activity and asset values, ultimately threatening financial stability



Stress testing

Stress testing exercises are an important tool for financial institutions to evaluate their portfolio vulnerability to adverse scenarios, especially when outside the range of historical events.

Climate stress testing involves assessing financial institutions' resilience to climate-related risks



Literature review on climate stress testing



Focus

We focus on a climate stress testing framework for mortgage portfolios

The real estate properties used to secure mortgages can be threatened for their location by severe damage from acute weather events with cascading effects on the mortgage default risk.

Some examples from literature and practitioners

Author(s)	Some main features	Some main limitations
Wong and Ho (2023)	The PD and the LGD are estimated by designing predetermined and hypothetical extreme weather events.	The scenarios are not designed using science-based climate or statistical models
Caloia and Jansen (2021)	The design of the stressed flood depth scenario derived from guidance from a Dutch government expert group on flood incidence and related inundation depths.	The scenarios by return periods are not designed using, for example, hydrodynamic models, allowing estimating location-specific flood depths
Federal Reserve Board (2023)	The first exercise of climate stress tests for the six largest U.S. Banks. One of the goals is to evaluate the impacts of extreme events on residential real estate assets of the mortgage portfolio (PD). The FRB recommended that banks consider including tropical cyclone events with a return period of 1-in-100 years and 1-in-200 years in the Northeastern U.S.	No larger return period events are considered No uniform scenarios



Purpose



This study aims to overcome some of these limitations by proposing a methodology for assessing the change in the default probability of mortgage loans affected by extreme weather events (tropical cyclones) with large return periods, such as up to 1-in-1,000 years

Moreover, we aim to address the following two research questions:



The first is whether the default risk in different geographical areas shows heterogeneous exposure to tropical cyclones.



The second research question is focused on insurance coverage to understand if and to what extent it contributes to mitigating the effects of physical risk on mortgage defaults.



For the empirical analysis, we apply the suggested climate stress testing framework to a **portfolio of mortgages** provided by Freddie Mac in **Louisiana**, a U.S. state affected by **devastating hurricanes in the last decades** (e.g., Hurricane Katrina in August 2005; Hurricane Laura in August 2020; and Hurricane Ida in August 2021)



Methodological framework









We designed a climate stress testing framework based on the following steps:



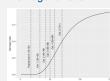
Data source collection and data management

- Data collection (loan-month observations, flood risk index, flood insurance, etc)
- Mapping of historical tropical cyclone events and estimation of the damages using a power function (Emanuel, 2011)

Tropical cyclone tracks



Damage function

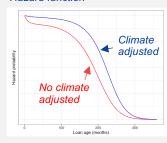




Estimation of a credit scoring model

- Estimation of a dynamic credit scoring model based on survival analysis – including weather events and flood insurance coverage
- Predictive accuracy assessment (out-of-time and out-of-sample)

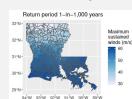
Hazard function





Scenario of damages by return period

 Design of the tropical cyclone scenario by return periods



Wind speed scenario

 Estimation of the index of damage by return periods (proxy for the property damage fraction in the zone)



Stress testing

- Climate damage-adjusted LTV
- Evaluation of the stressed PD under extreme weather events scenarios with or without flood insurance coverage

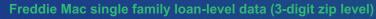
PD - stressed scenario





Data collection

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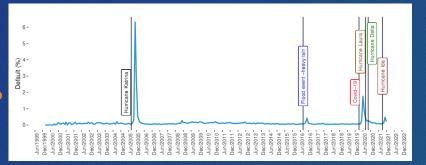








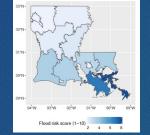




National Oceanic and Atmospheric Administration



First Street Foundation



National Flood Insurance Program (NFIP)



We aggregate the metrics of weather events, damage index, flood risk, and insurance coverage from 5-digit to 3-digit zip code level. We consider using the #properties as a measure for weighing.

Bureau of Labor Statistics



Federal Housing Finance Agency



Housing price index at the three-digit zip code level



Model of credit scoring









Methodology

The dependent variable of a credit scoring model is

$$Y_{it} = \begin{cases} 1 & Y_{it}^* > 0 \\ 0 & \text{otherwise} \end{cases}$$

mortgage default based on the 90+ day delinquency

The model is defined as

$$Y_{it}^* = \alpha + \phi(d)^\top \beta_0 + \boldsymbol{z}_{iot}^\top \beta_1 + \boldsymbol{x}_{it-k}^\top \beta_2 + \boldsymbol{w}_{it-k}^\top \beta_3 + \boldsymbol{c}_{it-k}^\top \beta_4 + \epsilon_{it}$$

Hazard function

Static and timevarying variables Weather and insurance variables

We estimate the model using a logit link function (Bellotti and Crook, 2013)

Results from survival models

	Base	model	Base model+C	limate-related varia
Variables	Estimates	Std. Error	Estimates	Std. Error
Intercept	-4.9471***	0.3179	-6.4038***	0.3199
Loan age	-0.0061	0.0069	-0.0086	0.0069
Loan age (squared)	0.0001**	0.0000	0.0001*	0.0000
Loan age (log)	1.3245***	0.2442	0.9445***	0.2433
oan age (log squared)	-0.1599**	0.0639	-0.0578	0.0638
Occupacy type (Base: Primary residence)				
Investment property	0.2286***	0.0418	0.2396***	0.0415
Second home	-0.0331	0.0697	-0.0436	0.0699
Nr. of units	0.3259***	0.0297	0.2280***	0.0308
Mortgage characteristics				
FICO score	-0.0102***	0.0001	-0.0092***	0.0002
$LTV_{t=3}$	0.0149***	0.0008	0.0196***	0.0008
Spread $IR_{t=3}$	0.0267**	0.0126	0.0951***	0.0127
DTI	0.0185***	0.0009	0.0177***	0.0009
Loan motivation (Base: Purchase)				
Refinance - cash out	0.2900***	0.0257	0.2979***	0.0259
Refinance - no cash out	0.1692***	0.0251	0.1791***	0.0252
Nr. of borrowers	-0.4601***	0.0206	-0.4707***	0.0207
Macroeconomics + insurance				
Unemployment rate _{t-3}	0.2685***	0.0045	0.2856***	0.0046
FItakeup _{t=3}			-0.0229***	0.0006
$ \overline{ \text{Tropical cyclones damages} + \text{major flood event} + \text{flood risk} } $				
Flood event 2016_{t-3}			2.0452***	0.0863
Damage index _{t=3}			0.1404***	0.0026
Flood risk			0.2270***	0.0049
Deviance explained (%)	6.43		9.02	



Model of credit scoring – predictive accuracy









Out-of-time

				MUMBER SAVE	100000000000000000000000000000000000000	estr teatres
Model	Base model+	AUC	Н	KS	F-score	ACC
Survival model	No climate-related variables	0.6912	0.1016	0.2971	0.0055	77.0132
	Climate-related variables	0.7117	0.1280	0.3263	0.0063	81.0094
Binomial model	No climate-related variables	0.6736	0.0885	0.2729	0.0051	75.2346
(link = Logit)	Climate-related variables	0.6997	0.1164	0.3127	0.0060	79.0430
Binomial model	No climate-related variables	0.6744	0.0896	0.2770	0.0051	74.6558
(link = Probit)	Climate-related variables	0.7005	0.1175	0.3103	0.0058	77.0727

The training sample encompasses loans observed from 2000 to 2020. The test sample looks one year ahead (2021)

Out-of-sample

	40				
Base model+	AUC	Н	KS	F-score	ACC
No alimete related veriables	0.7620	0.1056	0.3055	0.0068	69.4748
Climate-related variables	0.7967	0.1950 0.2576	0.3933 0.4414	0.0008	72.2825
No climate-related variables	0.7559	0.1864	0.3826	0.0066	69.1860
Climate-related variables	0.7869	0.2472	0.4258	0.0074	72.0425
X 10 . 1 . 1 . 1 . 1	0.000	0.4050	0.0000	0.000	00.0550
1.0 011111110 1010100 10111010100	0				68.3752
Climate-related variables	0.7902	0.2539	0.4308	0.0073	71.3284
	No climate-related variables Climate-related variables No climate-related variables	No climate-related variables 0.7630 Climate-related variables 0.7967 No climate-related variables 0.7559 Climate-related variables 0.7869 No climate-related variables 0.7565	No climate-related variables 0.7630 0.1956 Climate-related variables 0.7967 0.2576 No climate-related variables 0.7559 0.1864 Climate-related variables 0.7869 0.2472 No climate-related variables 0.7565 0.1870	No climate-related variables 0.7630 0.1956 0.3955 Climate-related variables 0.7967 0.2576 0.4414 No climate-related variables 0.7559 0.1864 0.3826 Climate-related variables 0.7869 0.2472 0.4258 No climate-related variables 0.7565 0.1870 0.3833	No climate-related variables 0.7630 0.1956 0.3955 0.0068 Climate-related variables 0.7967 0.2576 0.4414 0.0076 No climate-related variables 0.7559 0.1864 0.3826 0.0066 Climate-related variables 0.7869 0.2472 0.4258 0.0074 No climate-related variables 0.7565 0.1870 0.3833 0.0065

The average predictive accuracy measures are derived from five-fold out-of-sample cross-validation, with each test sample size constituting one-fifth of the ID loans

- Models incorporating climate-related variables (extreme events, risk exposure, and mitigation risk measures) outperform those that exclude them.
- The survival model outperforms the logit and probit models without duration time.
- These findings not only validate the existing literature on the enhanced predictive accuracy of credit risk when integrating climate-related variables but also underscore the growing preference for survival models over traditional scoring models (such as logit and probit models)



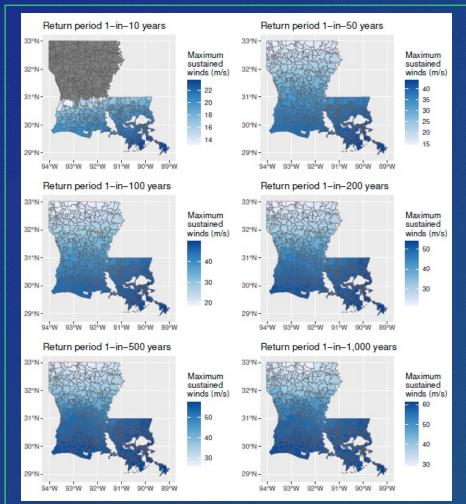
Design of scenarios – tropical cyclone wind speed











- We considered using the Synthetic Tropical cyclOne geneRation Model (STORM) proposed by Bloemendaal et al. (2020) to generate wind speed scenarios for tropical cyclones.
- As elaborated in Meiler et al. (2022), the scenarios obtained from STORM incorporate a greater number of high-impact events (specifically, Hurricanes categories 3-5) compared to alternative solutions (e.g., CLImate ADAptation (CLIMADA) platform and the Columbia HAZard model). This unique feature enables us to generate the most severe scenarios of tropical cyclones, facilitating a thorough evaluation of their impact on mortgage defaults.
- Coastal zones have emerged as the areas most affected by severe tropical cyclones



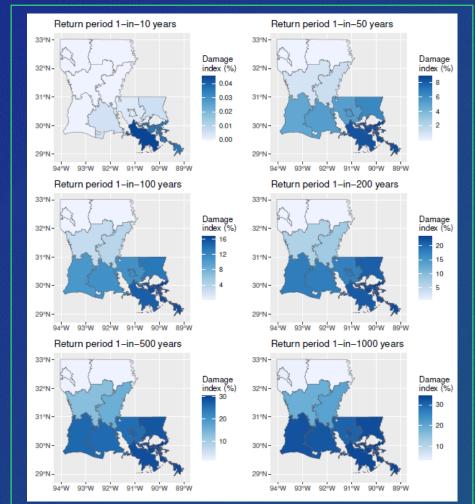
Design of scenarios – damage index











- The damage index is at the five-digit ZIP code level. As the information on mortgages is provided at the three-digit ZIP code level, we compute a weighted average to obtain the wide damage index at the three-digit ZIP code level using the number of properties is provided by First Street Foundation
- Consistently with the wind speed scenarios, the most substantial damage is expected in coastal areas. This occurs when hurricanes intensify by absorbing heat from warmer ocean waters via airsea heat fluxes, making landfall on the mainland with heightened energy levels



From damage index to climate-adjusted LTV



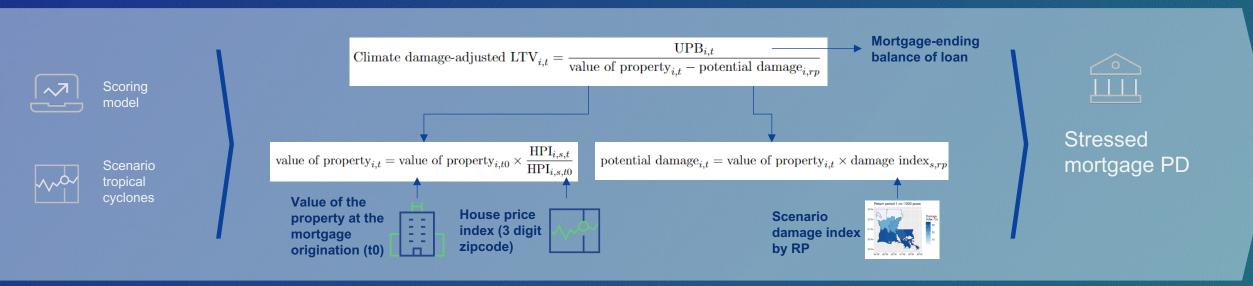






Damages from tropical cyclones, encompassing wind speed, heavy rain/flooding, and storm surges, affect the value of the impacted properties.

The decline in property value contributes to a rise in the loan-to-value (LTV) ratio, consequently elevating the probability of default on mortgages. For a damage index with return period *rp*, we estimate a simplified climate damage-adjusted LTV ratio associated with each mortgage *i* as follows:





Stress test for mortgage PD









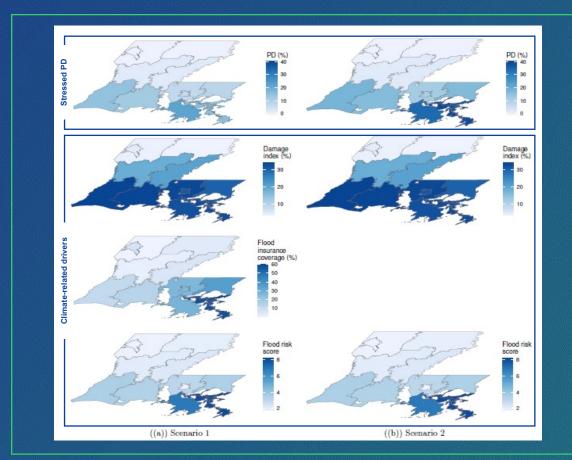
	Scenarios predicted PD (%)						
	with current (%) of insurance coverage on flood risk						
Scenario	Mean	$P50^{th}$	$P95^{th}$	$P97^{th}$	$P99^{th}$		
Baseline	0.0782	0.0574	0.2091	0.2491	0.3507		
RP 1-in-10 years	0.0784	0.0575	0.2097	0.2497	0.3520		
RP 1-in-50 years	0.2055	0.1354	0.6106	0.7527	1.0979		
RP 1-in-100 years	0.5858	0.3565	1.8987	2.3584	3.4035		
RP 1-in-200 years	1.6604	1.0700	5.3277	6.5459	9.1542		
RP 1-in-500 years	4.9574	3.4922	15.0860	18.0101	23.8237		
RP 1 -in- $1,000$ years	9.9810	7.7341	28.1505	32.1224	39.9288		

Scenarios predicted PD (%)

without	an	insurance	coverage	on	flood	risk
WIGHOUG	COLL	III di di di di	coverage	OII	noou	LIDIX

	without an insurance coverage on flood risk						
Scenario	Mean	$P50^{th}$	$P95^{th}$	$P97^{th}$	$P99^{th}$		
RP 1-in-10 years	0.1760	0.1129	0.5324	0.6648	0.9774		
RP 1-in-50 years	0.5306	0.2565	1.9356	2.4174	3.5373		
RP 1-in-100 years	1.5782	0.6815	6.1391	7.6390	10.9140		
RP 1-in-200 years	4.2422	2.0653	15.9236	19.2984	26.1142		
RP 1-in-500 years	10.9998	6.7615	36.5067	42.1965	51.5627		
RP 1-in-1,000 years	19.3382	14.8431	54.2235	59.8746	68.3952		

Scenarios of default probability for various return periods of tropical cyclone wind speeds



The mean probability of default is computed at the three-digit zip code level in Scenarios 1 and 2. Scenario 1 includes flood insurance coverage, while Scenario 2 excludes it. The analysis focuses on a tropical cyclone event with a return period of 1-in-1,000 years



Conclusions

We apply the proposed climate stress test framework to a mortgage portfolio in Louisiana, a region frequently affected by major tropical cyclones in the U.S.

Main findings:

- We found that damage from tropical cyclones with larger return periods significantly increased the default probability.
- The effects of tropical cyclones on the default risk of mortgage loans are spatially heterogeneous.
- The findings highlight the crucial role of flood insurance in mitigating default risk in high-risk flooding locations

Implications for a wider group of stakeholders:



- <u>Financial institutions and regulators</u> can use our proposal to identify mortgage loans that are most vulnerable to extreme climate events. This would support lenders in adopting a more reliable approach for prudential risk management.
- Lenders can share their default risk assessment with <u>borrowers</u> who would be able to make more <u>informed decisions on</u>
 the risk of buying a property in risky areas and potentially purchase <u>insurance to mitigate losses</u> from catastrophic
 events.
- <u>Policymakers</u> can consider these findings to <u>allocate public funds</u> in a more effective way to <u>protect homeowners</u> living in high-risk areas.





III. The ENGAGE Templates v1.2 and the ENGAGE Portal

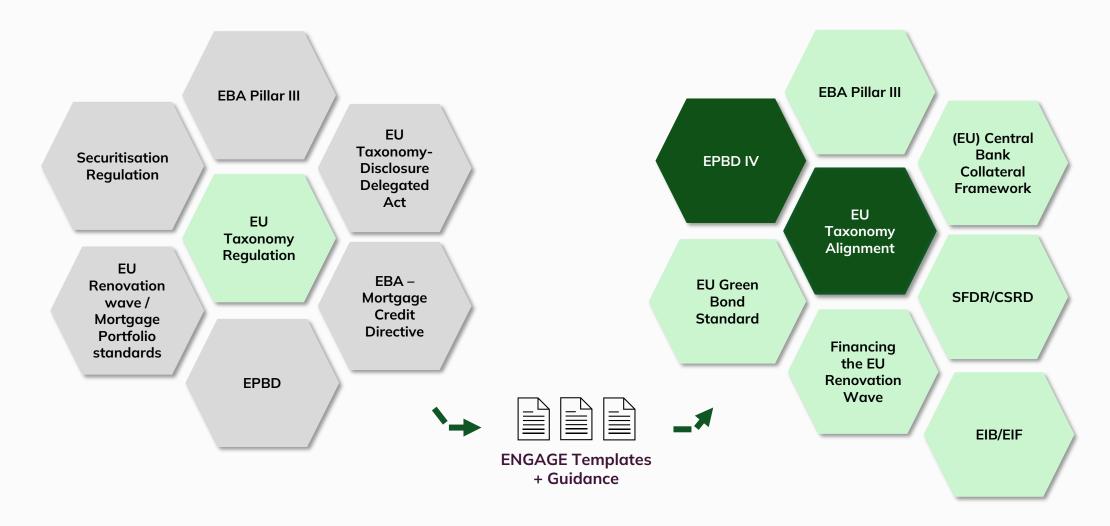
Vincent Mahieu

Hypoport





ENGAGE for ESG regulatory background







The EU Taxonomy - Scope for ENGAGE for ESG

In short:

- The **EU Taxonomy** is a regulation that constitutes the cornerstone of the EU Sustainable Finance Action Plan.
- The EU Taxonomy provides a **common language** and uniform criteria to identify the extent to which economic activities may be considered environmentally sustainable.
- Its aim is to reorientate **capital flows towards lower-emission economic activities** that will help decarbonise the economy.
- The EU hopes to reach its 2030 climate targets and the ultimate goal of net-zero greenhouse gas emissions by 2050.
- The EU Taxonomy defines the minimum criteria that economic activities should comply with in order to be considered environmentally sustainable.



EU Taxonomy Alignment Steps

Do No Significant Harm

Substantial Contribution

Additional Technical Screening Criteria Available

Minimum Safeguards

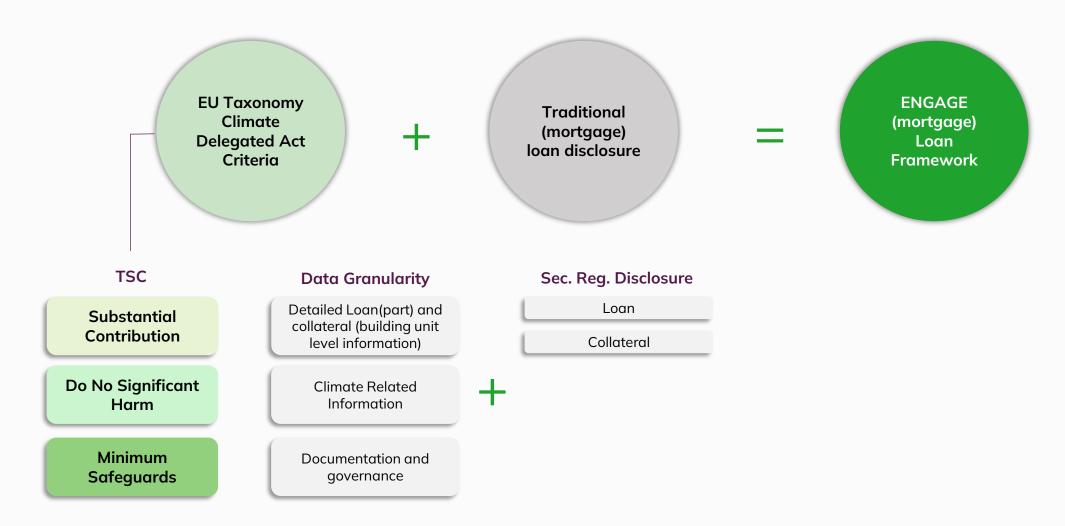
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Sustainable Activity





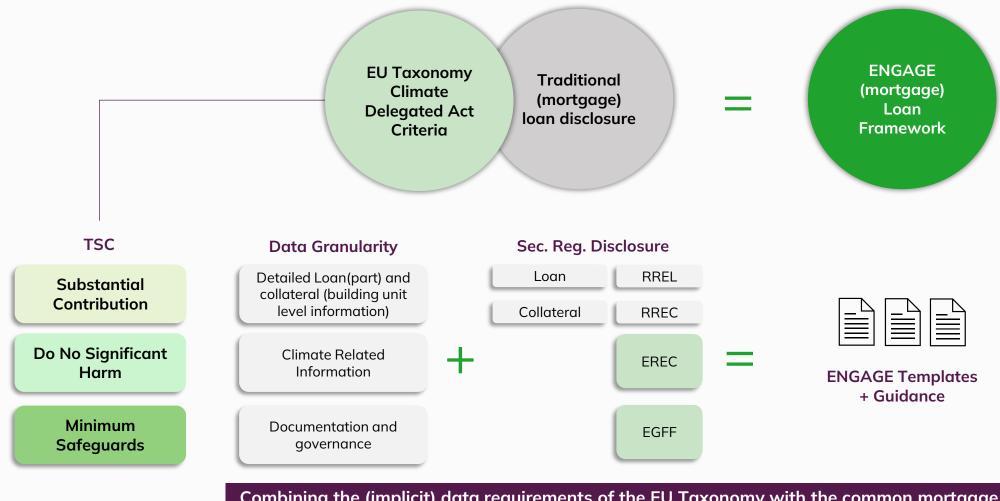
The EU Taxonomy - Scope for ENGAGE for ESG







The EU Taxonomy - Scope for ENGAGE for ESG

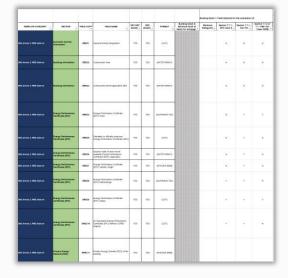






The ENGAGE Templates

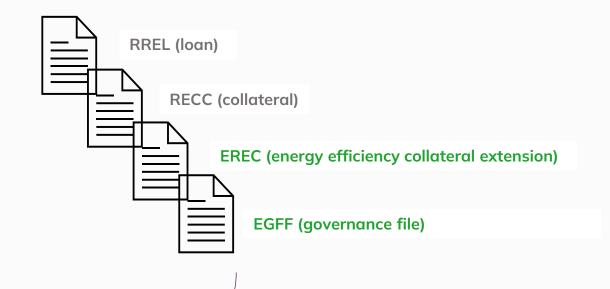
Templates



Guiding Documentation



Sample files



Available upon request





Economic Activities - Scope for ENGAGE



When developing **version 1.2** of the ENGAGE Templates, the ENGAGE Consortium has considered:

- ✓ ENGAGE data framework is an add-on to existing market best practices
- ✓ By identifying common denominators among regulatory disclosure templates, financial institutions can simplify and streamline their (often costly) reporting processes.
- ✓ ENGAGE data framework is scalable and modular.
- ✓ New in ENGAGE Templates v1.2:
 - □ SCC 7.1
 - □ SCC 7.2, 7.3 + 7.6
 - □ CSRD Financed CO² emissions
 - Annex 6 version for unsecured renovation loans
 - **☐** EIB Group Green Checker fields

In the future, the ENGAGE Templates aim to cover relevant elements for EPBD IV and the EU Green Bond Standard





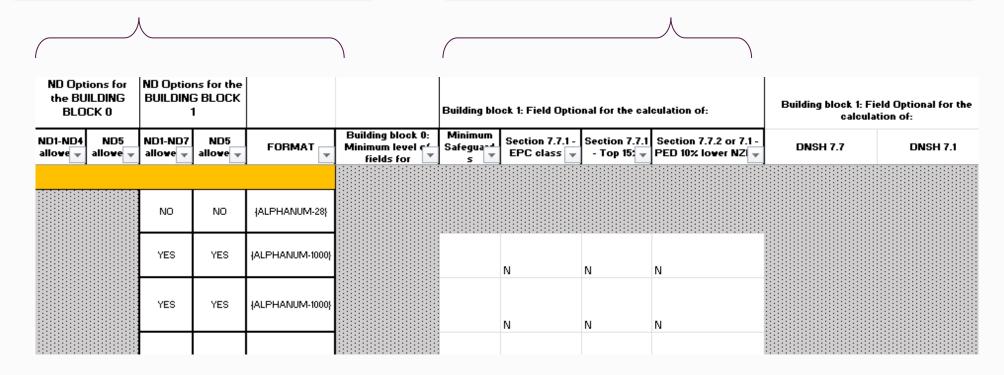
The ENGAGE Templates: how it works

Building Block

- Fields marked with 0 denote the original ESMA disclosure format.
- Fields marked with 1 indicate the ENGAGE
 Add-On fields.

Fields used for Taxonomy TSC assessment

- Fields with **N** are <u>not</u> optional for the TSC assessment for that economic activity.
- Fields with **Y** are not needed per se for the specific check, for that economic activity.







ENGAGE Output: EU Taxonomy Alignment Report

Section	Economic Activity	Subsection		
7.1	Construction of new buildings			
7.2(1) 7.2(2)	Renovation of existing buildings	Major Renovations Reduction of (net) Primary Energy Demand	•	New Constructions
7.3	Installation, maintenance and repair of energy efficiency equipment	Energy bernand		
7.4	Installation, maintenance and repair of charging stations for electric vehicles in buildings (and parking spaces attached to buildings)			
7.5	Installation, maintenance and repair of instruments and devices for measuring, regulation and controlling energy performance of buildings		•	Renovations
7.6	Installation, maintenance and repair of renewable energy technologies			
7.7(1)		Buildings built before 31 December 2020: building has at least an Energy Performance Certificate (EPC) class A.		
7.7(1a)	Acquisition and ownership of buildings	Buildings built before 31 December 2020 - Alternative: building is within Top 15%		Existing/now building
7.7(2)		Buildings built after 31 December 2020		Existing/new building

ENGAGE Output

1. ENGAGE EUT Alignment report

This report provides a comprehensive overview of the portfolio's underlying exposures in the EUT assessment:

- Each (sub) section of the EU Taxonomy's economic activities is broken down into (partial) alignment.
- Offers greater granularity than DDA templates.
- Enables deeper insights into the nuances of the TSC, including SCC and DNSH.

The output is broken down in loans, collateral building (units), and balance.

2. EU Taxonomy Data Quality report

We have also developed an EU Taxonomy Data Quality report for mortgage loans, offering detailed insights into data completeness and portfolio quality.





ENGAGE: EU Taxonomy Alignment Report



Portfolio Date (DD-MM-YY)

Example Portfolio 2024 01-08-2024

Total # of Loans Total # of Building Units

1300 1100

Total Balance €444,600,000.00 TSC passed %

50.94%

Section	Economic Activity	Subsection

		Buildings built before 31
		December 2020: building
7.7(1)	Acquisition and ownership of buildings	has at least an Energy
		Performance Certificate
		(EPC) class A.
		Buildings built before 31
7.7(1a)		December 2020 -
		Alternative: building is
		within Top 15%
7.7(2)		Buildings built after 31
		December 2020

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TSC assessment*

DNSH* (check) passed

	(
# of loans	# of Building Units	Balance	# of loans	# of Building Units	Balance
460	411	€133,400,000.00	414	370	€120,060,000.00
392	301	€113,680,000.00	353	271	€102,370,000.00
23	22	€6,670,000.00	21	20	€6,090,000.00
875	734	€253,750,000.00	788	661	€228,520,000.00

SCC (check passed)

# of loans	# of Building Units	Balance			
411	363	€119,190,000.00			
349	301	€101,210,000.00			
21	22	€6,090,000.00			
781	686	€226,490,000.00			

T	SC Passed (SCC + [ONSH)	
	# of Building Units	Balance	% o
	363	€119,190,000.00	3
	301	€101,210,000.00	3
	22	€6,090,000.00	

		SCC (check pass	ed)	
	% of all loans	% of total Building u	% of total Balance	% of to
0	35.38%	37.36%	30%	31
0	30.15%	27.36%	26%	27
	1.77%	2.00%	2%	1.0



		SCC (check pass	ed)		DNSH (check) pass	ed	TSC Passed (SCC + DNSH)			
	% of all loans	% of total Building u	% of total Balance	% of total loans	% of total Building t	% of total Balance	% of total loans	% of total Building u	% of total Balance	
	35.38%	37.36%	30%	31.85%	33.64%	27%	31.62%	33.00%	27%	
	30.15%	27.36%	26%	27.15%	24.64%	23%	26.85%	27.36%	23%	
	1.77%	2.00%	2%	1.62%	1.82%	1%	1.62%	2.00%	1%	
	67.31%	66.73%	57.07%	60.62%	60.09%	51.40%	60.08%	62.36%	50.94%	





ENGAGE: EU Taxonomy Alignment Report



Portfolio Name Example Portfolio 2024

Portfolio Date (DD-MM-YY) 01-08-2024

Total # of Loans 1300
Total # of Building Units 1100

Total Balance €444,600,000.00

TSC passed % 50.94%

ISC pas	ssed %	50.94%									
							TSC asses	ssment*			
Section Economic Activity Subsection		Subsection		SCC (check passed)		DNSH* (check) passed		TSC Passed (SCC + DNSH)			
			# of loans	# of Building Units	Balance	# of loans	# of Building Units	Balance	# of loans	# of Building Units	Balance
7.7(1)	- Acquisition and ownership of	Buildings built before 31 December 2020: building has at least an Energy Performance Certificate (EPC) class A.	460	411	€133,400,000.00	414	370	€120,060,000.00	411	363	€119,190,000.00
7.7(1a)	buildings a)	Buildings built before 31 December 2020 - Alternative: building is within Top 15%	392	301	€113,680,000.00	353	271	€102,370,000.00	349	301	€101,210,000.00
7.7(2)		Buildings built after 31 December 2020	23	22	€6,670,000.00	21	20	€6,090,000.00	21	22	€6,090,000.00
	Total		875	734	€253,750,000.00	788	661	€228,520,000.00	781	686	€226,490,000.00



ENGAGE: EU Taxonomy Alignment Report



Portfolio Name Example Portfolio 2024

Portfolio Date (DD-MM-YY) 01-08-2024

Total # of Loans 1300
Total # of Building Units 1100

Total Balance €444,600,000.00

TSC passed % 50.94%

TSC pass Total Portfolio

	SCC (check pass	ed)		DNSH (check) pass	ed	T	NSH)		
% of all loans	ıll loans % of total Building u % of total Balance		% of total loans	% of total Building u	% of total Balance	% of total loans	% of total Building ι	% of total Balance	
35.38%	37.36%	30%	31.85%	33.64%	27%	31.62%	33.00%	27%	
30.15%	27.36%	26%	27.15%	24.64%	23%	26.85%	27.36%	23%	
1.77%	2.00%	2%	1.62%	1.82%	1%	1.62%	2.00%	1%	
67.31%	66.73%	57.07%	60.62%	60.09%	51.40%	60.08%	62.36%	50.94%	



Only a limited number of data fields are actually needed!

In this example, of the ENGAGE v1.2 EREC file we see that only a limited number of data fields is needed for the economic activity 7.7 assessment checks:

- In green the minimum data needed for SCC checks
- In purple the DNSH check
- In light blue the PCAF / CO2 financed emissions

	FIELD CODE	CONTENT TO REPORT	Section 7.7.1 - EPC class A	Section 7.7.1 - Top 15%	Section 7.7.2 10% lower NZEB
	EREC1	Unique Identifier			
Identifiers	EREC2	Underlying Exposure Identifier Building block 0			
Identifiers	EREC3	Collateral Identifier Building block 0			
	EREC4	General Activity Designation	Y	Y	Υ
	EREC5	Construction Year	Y	Y	Y
	EREC6	Construction permit application date			
	EREC7	Energy Performance Certificate (EPC) Class	Y		
	EREC8	Estimated or officially produced EPC	Υ		
	EREC9	Issuance date of most recent EPC	Υ		Υ
	EREC10	EPC original Validity	Υ		
	EREC11	EPC Methodology			
Γ	EREC12	Energy Performance Certificate (EPC) Status			
	EREC13	EU-Equivalent EPBD-Regime			
20077	EREC14	Primary Energy Demand (PED) of the building			Υ
SCC 7.7	EREC15	Estimated or officially produced Primary Energy Demand (PED)			
	EREC16	Primary Energy Demand (PED) Based on Reference Building			
	EREC17	Primary Energy Demand (PED) Based on Building or Building unit			
	EREC18	Nearly zero-energy building (NZEB) theshold			Υ
	EREC19	Climatic zone code			
	EREC20	Geographic Region - climatic zone			
	EREC21	Building unit in top 15% indicator		Y	
	EREC22	Top 15% Explanatory Variable			
	EREC23	Top 15% Object Reference Value		Y	
	EREC24	Top 15% Object Treshold Value		Y	
	EREC25	DNSH EO2 Indicator	Υ	Υ	Υ
	EREC26	DNSH EO3 Indicator			
DNSH	EREC27	DNSH EO4 Indicator			
	EREC28	DNSH EO5 Indicator			
	EREC29	DNSH EO6 Indicator			
	EREC30	Balance for renovation of existing buildings			
	EREC31	Balance of IMR of EE equipment			
Renovation	EREC32	Balance of IMR of RE equipment			
SCC	EREC33	Balance of EE equipment			
	EREC34	Balance of RE equipment			
	EREC35	Large new constructions			
	EREC36	Financed CO2-emission	Υ	Y	Υ
Financed	EREC37	Surface Area	Υ	Y	Y
Emissions	EREC38	GHG Data Score	Υ	Y	Y
	EREC39	GHG Emission Factor	Υ	Y	Y

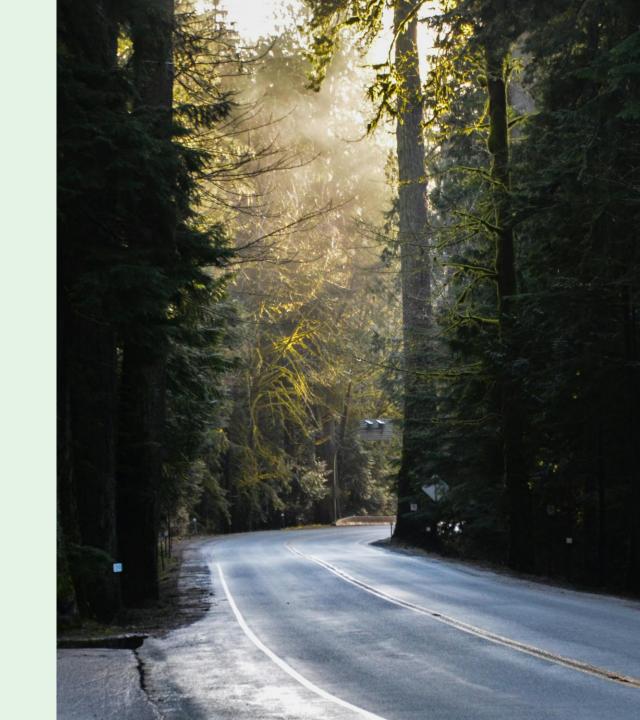




V. How to get involved

Marco Angheben

European DataWarehouse





Free ENGAGE trial period until June 2026



ENGAGE Demonstration

Submission to ENGAGE Portal

Output Reports

- Version 1.2 of the ENGAGE Templates available now through <u>Request for Access to the ENGAGE Templates</u>
- More than **80 institutions** have requested access to the ENGAGE Templates

The ENGAGE Team is available for clarifications on the ENGAGE Templates

- Signature of a <u>standard legal arrangement</u>
- Possibility to submit the ENGAGE sample files various times enriching the information

For each sample file submission two reports are generated: 1) a data quality report; 2) an EU Taxonomy alignment report.





Why become a Trial User via the Portal





Upcoming events

Let's meet in Lisbon and Paris!

European DataWarehouse's 2025 Portuguese Securitisation Event

- Date: Monday, 17 November 2025 at 14:00 (local time)
- Venue: Banco de Portugal, Lisbon, Portugal
- Register <u>here!</u>

European DataWarehouse's 2026 French Securitisation Event

- Date: Friday, 9 January 2026 at 14:30 (local time)
- Venue: Banque de France, Paris, France
- More details to come Stay tuned!









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Project Coordinator:

Marco Angheben marco.angheben@eurodw.eu

ENGAGE General Contact:

engage@eurodw.eu

Communications:

Carla Scarsella carla.scarsella@eurodw.eu