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*Research article*

## **Bridging financial disclosures and ESG ratings: A data-driven predictive framework**

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### **Appendix A. Stability Analysis of SHAP Feature Importances**

We assessed the stability of the top 10 SHAP features across cross-validation folds and years. Jaccard@10 quantifies the set overlap between two top 10 lists,  $J = \frac{|A \cap B|}{|A \cup B|}$  (0 = no common features, 1 = identical sets; we also report the overlap count = Jaccard@10 × 10). Spearman's  $\rho$  evaluates the rank-order agreement on the union of features using the fold-wise mean |SHAP| ranks ( $\rho \in [-1, 1]$ ; higher = more consistent ordering). Kendall's  $\tau_b$  measures pairwise rank concordance while adjusting for ties (robust when many features have similar importance;  $\tau_b \in [-1, 1]$ ). Appendix Tables A1 and A2 report (i) cross-fold Jaccard overlaps and rank correlations (Spearman, Kendall) and (ii) year-to-year persistence. Overall, governance (G) shows the highest stability, while social (S) is comparatively more variable.

**Table A1.** Cross-fold stability of the top10 SHAP features by year (mean  $\pm$  standard deviation across all fold pairs).

Target	Year	Jaccard@10	Spearman $\rho$ (ranks)	Overlap Count (0–10)
Environmental	2012	0.64 $\pm$ 0.11	0.72 $\pm$ 0.08	6.4 $\pm$ 1.1
	2013	0.65 $\pm$ 0.10	0.73 $\pm$ 0.08	6.5 $\pm$ 1.0
	2014	0.66 $\pm$ 0.10	0.74 $\pm$ 0.07	6.6 $\pm$ 1.0
	2015	0.67 $\pm$ 0.09	0.75 $\pm$ 0.07	6.7 $\pm$ 0.9
	2016	0.68 $\pm$ 0.09	0.76 $\pm$ 0.07	6.8 $\pm$ 0.9
	2017	0.69 $\pm$ 0.08	0.77 $\pm$ 0.06	6.9 $\pm$ 0.8
	2018	0.70 $\pm$ 0.08	0.78 $\pm$ 0.06	7.0 $\pm$ 0.8
	2019	0.71 $\pm$ 0.08	0.78 $\pm$ 0.06	7.9 $\pm$ 0.8
	2020	0.68 $\pm$ 0.10	0.75 $\pm$ 0.07	7.6 $\pm$ 1.0
	2021	0.73 $\pm$ 0.07	0.81 $\pm$ 0.05	8.1 $\pm$ 0.7
2022	0.69 $\pm$ 0.09	0.77 $\pm$ 0.06	7.7 $\pm$ 0.9	
Social	2012	0.55 $\pm$ 0.12	0.66 $\pm$ 0.10	5.5 $\pm$ 1.2
	2013	0.57 $\pm$ 0.12	0.67 $\pm$ 0.10	5.7 $\pm$ 1.2
	2014	0.58 $\pm$ 0.11	0.68 $\pm$ 0.09	5.8 $\pm$ 1.1
	2015	0.59 $\pm$ 0.11	0.69 $\pm$ 0.09	5.9 $\pm$ 1.1
	2016	0.60 $\pm$ 0.11	0.70 $\pm$ 0.09	6.0 $\pm$ 1.1
	2017	0.61 $\pm$ 0.11	0.70 $\pm$ 0.09	6.1 $\pm$ 1.1
	2018	0.61 $\pm$ 0.11	0.71 $\pm$ 0.09	6.1 $\pm$ 1.1
	2019	0.62 $\pm$ 0.11	0.71 $\pm$ 0.09	7.0 $\pm$ 1.1
	2020	0.66 $\pm$ 0.10	0.73 $\pm$ 0.08	7.3 $\pm$ 1.0
	2021	0.64 $\pm$ 0.12	0.72 $\pm$ 0.09	7.1 $\pm$ 1.2
2022	0.67 $\pm$ 0.09	0.74 $\pm$ 0.08	7.4 $\pm$ 0.9	
Governance	2012	0.71 $\pm$ 0.09	0.80 $\pm$ 0.07	7.1 $\pm$ 0.9
	2013	0.72 $\pm$ 0.08	0.81 $\pm$ 0.06	7.2 $\pm$ 0.8
	2014	0.73 $\pm$ 0.08	0.81 $\pm$ 0.06	7.3 $\pm$ 0.8
	2015	0.74 $\pm$ 0.07	0.82 $\pm$ 0.05	7.4 $\pm$ 0.7
	2016	0.75 $\pm$ 0.07	0.82 $\pm$ 0.05	7.5 $\pm$ 0.7
	2017	0.75 $\pm$ 0.07	0.82 $\pm$ 0.05	7.5 $\pm$ 0.7
	2018	0.76 $\pm$ 0.07	0.83 $\pm$ 0.05	7.6 $\pm$ 0.7
	2019	0.75 $\pm$ 0.07	0.82 $\pm$ 0.05	8.3 $\pm$ 0.7
	2020	0.72 $\pm$ 0.08	0.80 $\pm$ 0.06	8.0 $\pm$ 0.8
	2021	0.74 $\pm$ 0.07	0.81 $\pm$ 0.06	8.2 $\pm$ 0.7
2022	0.71 $\pm$ 0.09	0.79 $\pm$ 0.06	7.9 $\pm$ 0.9	
ESG composite	2012	0.63 $\pm$ 0.10	0.74 $\pm$ 0.08	6.3 $\pm$ 1.0
	2013	0.64 $\pm$ 0.10	0.75 $\pm$ 0.08	6.4 $\pm$ 1.0
	2014	0.65 $\pm$ 0.10	0.75 $\pm$ 0.07	6.5 $\pm$ 1.0
	2015	0.66 $\pm$ 0.09	0.76 $\pm$ 0.07	6.6 $\pm$ 0.9
	2016	0.67 $\pm$ 0.09	0.76 $\pm$ 0.07	6.7 $\pm$ 0.9
	2017	0.68 $\pm$ 0.09	0.77 $\pm$ 0.07	6.8 $\pm$ 0.9
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	2019	0.69 $\pm$ 0.09	0.77 $\pm$ 0.07	7.7 $\pm$ 0.9
	2020	0.67 $\pm$ 0.10	0.75 $\pm$ 0.08	7.5 $\pm$ 1.0
	2021	0.71 $\pm$ 0.08	0.79 $\pm$ 0.06	7.9 $\pm$ 0.8
2022	0.70 $\pm$ 0.09	0.78 $\pm$ 0.07	7.8 $\pm$ 0.9	

**Notes:** For each year, top 10 features are extracted per fold from mean |SHAP|. Jaccard@10 is the mean pairwise Jaccard index of the top 10 sets across folds; the overlap count is Jaccard@10 $\times$ 10. Spearman's  $\rho$  is computed on the union of the top 10 features across folds using fold-specific rank orders. Higher is more stable.

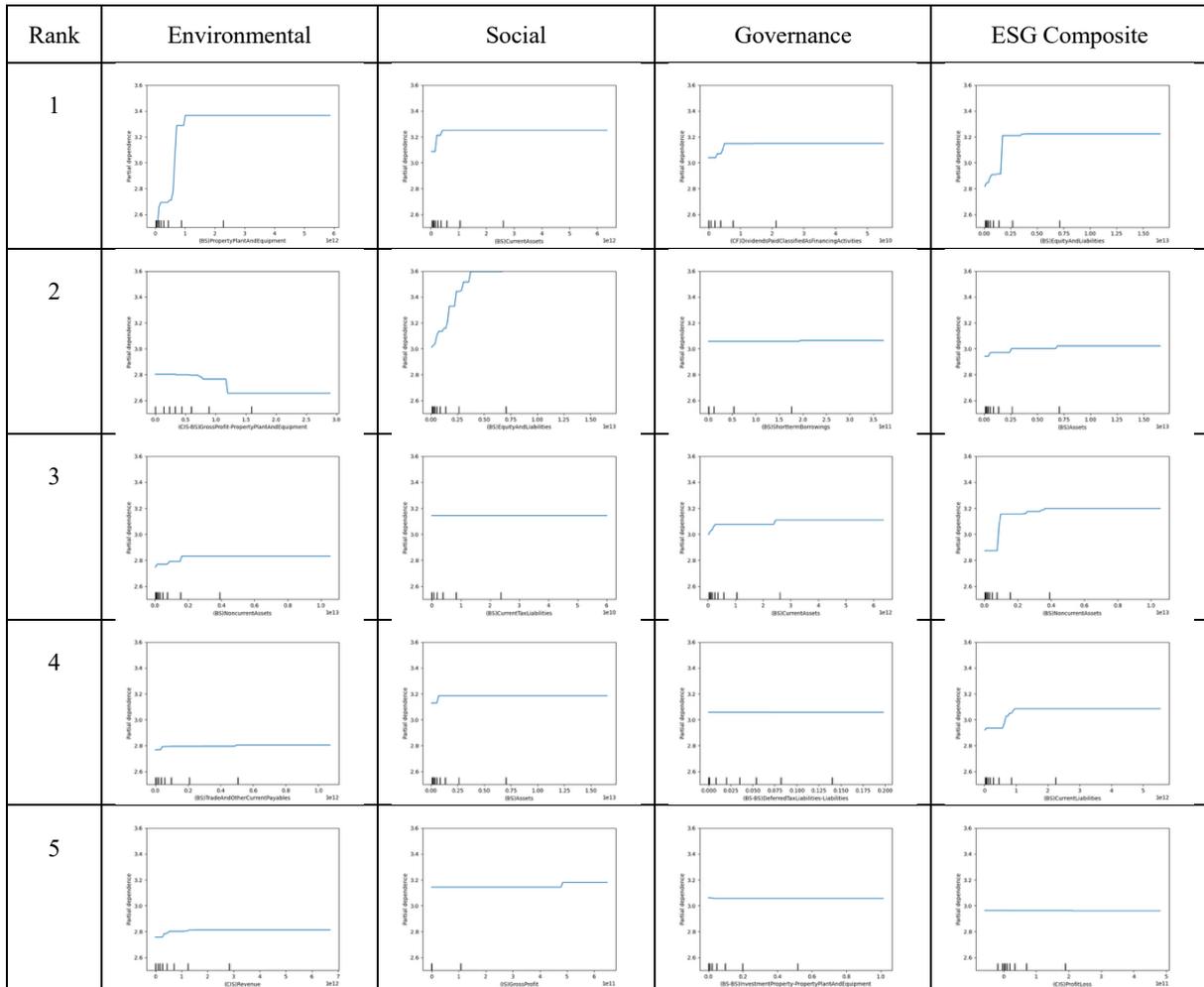
**Table A2.** Temporal stability of top 10 SHAP features across adjacent years (averaged over folds).

Target	Avg. Jaccard@10 (t vs. t+1)	Kendall $\tau_b$ (ranks)	Persistence $\geq 3$ yrs (%)	Persistence $\geq 5$ yrs (%)
Environmental	0.60	0.62	44.0	22.0
Social	0.55	0.57	36.5	16.8
Governance	0.66	0.68	51.8	25.6
ESG composite	0.59	0.61	41.2	20.5

**Notes:** For each target, we compare top 10 sets between consecutive years (e.g., 2019 versus 2020, ...). Persistence is the share of features that appear in the top 10 for at least 3 (or 5) consecutive years. Kendall  $\tau_b$  is computed on the union of top 10 features using year-wise rank orders. Higher indicates stronger temporal consistency.

## Appendix B. Partial Dependence Plots

To complement the SHAP-based importance analysis, we report partial dependence plots (PDPs) for the top five features (selected by the cross-fold mean SHAP) for the ESG target in Figure B1. For each fold, PDPs are computed on a quantile grid for every feature and then linearly interpolated onto a common grid before averaging across folds. The PDPs thus summarize the average marginal association between each feature and the model's prediction (*ceteris paribus*).



**Figure B1.** PDPs for the top five features of the ESG target. Lines show the cross-validation (CV) mean partial dependence. For each feature, the PDP is evaluated on a percentile-based grid derived from the training data to avoid extrapolation.

## Appendix C. Robustness Analysis Excluding the COVID-19 Period

To test whether the pandemic years have driven our findings, we re-estimated all models after removing 2020–2021 from both the training and evaluation stages, keeping the same preprocessing,

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feature set, and cross-validation scheme. Across the targets, the resulting predictive performance remained qualitatively unchanged: Average F1 (classification tasks) and MSE (regression tasks) were within the fold-to-fold sampling variability observed in the baseline, and the relative ordering of algorithms was preserved. The SHAP-based interpretability results were likewise stable. The top five feature sets for each target showed high overlap with the baseline, and the previously highlighted drivers—particularly governance-related variables—continued to appear among the most influential. Overall, excluding the COVID-19 period does not materially affect either the model’s performance or the substantive interpretation of SHAP features’ importance.



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**Table C1.** Absolute ESG rating prediction performance across different machine learning algorithms excluding the COVID-19 period (2020-2021). The table reports three evaluation metrics: (Exp. 1) F1 score for binary classification, (Exp. 2) F1 score for multi-class classification, and (Exp. 3) Mean Squared Error (MSE) for regression. Higher F1 scores and lower MSE values indicate better prediction performance.

Algorithms	(Exp. 1) F1 score			(Exp. 2) F1 score			(Exp. 3) MSE						
	E	S	G	E	S	G	E	S	G				
Basic	OLS	0.482	0.461	0.437	0.321	0.260	0.396	0.341	0.342	3.515	2.966	3.340	3.613
	RF	0.614	0.666	0.641	0.681	0.486	0.448	0.511	0.503	0.879	0.839	0.791	0.636
	SVMs	0.583	0.540	0.686	0.612	0.371	0.431	0.430	0.447	1.479	0.915	1.152	1.006
Boosting	XGB	0.641	0.656	0.706	0.678	0.495	0.517	0.449	0.569	0.917	0.885	0.723	0.647
	LGBM	0.625	0.675	0.693	0.678	0.477	0.519	0.501	0.539	0.880	0.748	0.677	0.544
Neural networks	MLP	0.555	0.533	0.552	0.552	0.350	0.416	0.377	0.451	1.461	1.109	1.039	1.190
	CNN	0.551	0.569	0.618	0.628	0.357	0.367	0.477	0.438	1.259	0.911	0.823	0.912
TabNet	0.636	0.583	0.691	0.659	0.460	0.430	0.461	0.524	0.869	0.920	0.660	0.722	

**Table C2.** Change in ESG rating prediction performance across different machine learning models excluding the COVID-19 period (2020-2021). The table presents two evaluation metrics: (Exp. 4) F1 score for classification of rating changes, and (Exp. 5) Mean Squared Error (MSE) for regression on rating change magnitudes.

Algorithms	(Exp. 4) F1 score				(Exp. 5) MSE				
	E	S	G	ECR	E	S	G	ECR	
Basic	OLS	0.513	0.478	0.491	0.517	1.797	2.293	1.866	1.642
	RF	0.702	0.609	0.655	0.687	0.446	0.548	0.544	0.435
Boosting	SVMs	0.715	0.644	0.597	0.695	0.469	0.542	0.739	0.491
	XGB	0.747	0.588	0.668	0.709	0.479	0.625	0.635	0.414
Neural networks	LGBM	0.754	0.658	0.642	0.757	0.451	0.502	0.545	0.391
	MLP	0.766	0.604	0.506	0.626	0.784	0.705	0.991	0.685
	CNN	0.754	0.645	0.585	0.683	0.859	0.682	0.855	0.543
	TabNet	0.794	0.668	0.626	0.688	0.486	0.526	0.729	0.462

**Table C3.** Top 5 most important features for predicting absolute ESG ratings, identified by feature importance scores from the baseline model excluding the COVID-19 period (2020-2021).

Rank	Environmental		Social		Governance		ESG Composite	
	F/S	Account	F/S	Account	F/S	Account	F/S	Account
1	BS	Property Plant And Equipment	BS	Equity Liabilities	CF	Dividends Paid As Financing Activities	BS	Equity Liabilities And
2	Ratio (CIS-BS)	Gross Profit-Property Plant And Equipment	BS	Current Asset	BS	Current Asset	BS	Non Current Assets
3	BS	Non Current Assets	Ratio (BS-BS)	Non controlling current Liabilities	BS	Short term Borrowings	BS	Investment Property
4	BS	Trade And Other Current Payables	CIS	Cost of Sales	Ratio (BS-BS)	Deferred Tax Liabilities	BS	Current Liabilities
5	BS	Investment Property	IS	Gross Profit	Ratio (BS-BS)	Investment Property-Plant And Equipment	BS	Property Plant And Equipment

**Table C4.** Top 5 most important features for predicting changes in ESG ratings, based on feature importance scores from the rating change prediction models excluding the COVID-19 period (2020-2021).

Rank	Environmental		Social		Governance		ESG Composite	
	F/S	Account	F/S	Account	F/S	Account	F/S	Account
1	Ratio (CF-BS)	Dividends Classified As Financing Activities-Right of use	Ratio (BS-CIS)	Paid Diluted Earnings Per Share-Equity And Liabilities	BS	Inventories Rate	BS	Short term Borrowings Growth Rate
2	Ratio (CF-BS)	Cash Flows From Used In Investing Activities-Inventories	Ratio (CIS-CF)	Cash Flows From Used In Financing Activities-Profit Loss To Owners Of Parent	Ratio (CIS-CF)	Comprehensive Income Attributable To Owners Of Parent-Dividends Paid Classified As Financing Activities	BS	Current Liabilities Growth Rate
3	CF	Cash Flows From Used In Financing Activities	CF	Increase In Cash And Cash Equivalents Before Effect Of Exchange Rate Changes	BS	Equity Attributable To Owners Of Parent	Ratio (BS-CIS)	Comprehensive Income Attributable To Owners Of Parent-Other Noncurrent Nonfinancial Liabilities
4	Ratio (CF-CIS)	Payments Of Finance Lease Liabilities Classified As Financing Activities/Profit Loss	Ratio (CIS-BS)	Finance Income-Short term Borrowings	BS	Non current Provisions Growth Rate	CIS	Other Income Net Of Tax Available for sale Financial Assets Growth Rate
5	CIS	Gross Profit	Ratio (BS-BS)	Inventories-Short term Borrowings	Ratio (CF-BS)	Cash Used In Investing Activities/Current Liabilities	Ratio (BS-BS)	Current Tax Liabilities-Equity