



The nonlinear impact of firms' ESG disclosures on analysts' earnings forecast accuracy

Xuehui Zhang^a, Guoying Mu^b, Fei Han^{b,*}

^a Inner Mongolia University of Finance and Economics, Huhhot, China

^b School of Economics and Management, Inner Mongolia University of Technology, Huhhot, China

ARTICLE INFO

Keywords:

ESG disclosure
Analyst earnings forecast accuracy
U-shaped relation

ABSTRACT

The disclosure of environmental, social, and governance (ESG) information by firms can signal to the capital markets the potential sustainable development capabilities of the company and provide oversight to curb opportunistic management practices. However, the continuous disclosure of ESG information not only leads to excessive investment and increased operational risks but also contributes to information overload in the capital markets. We examine the impact of firms' ESG disclosures on analyst earnings forecast accuracy (AEFA) using a nonlinear approach to capture both sides of ESG effects. Our nonlinear findings suggest that ESG disclosure has a U-shaped effect on AEFA. Specifically, a firm's ESG disclosure enhances AEFA but at a decreasing rate. The favorable effect becomes smaller as firms disclose more ESG, and eventually, increases in ESG disclosure led to worsening AEFA. The conclusion remains unchanged after an array of robustness checks. The mechanism test reveals that earnings management and media attention mediate the U-shaped relation between ESG disclosure and AEFA. Additional analysis shows that the U-shaped relation is more salient for SOEs and non-polluting firms. The findings not only broaden the research on the impact of ESG disclosure on AEFA, but also help investors make sound investment decisions.

1. Introduction

ESG refers to corporate environmental, social, and corporate governance initiatives. Many firms proactively disclose their ESG activities. Moreover, some stakeholders, such as investors, employees, customers, supply chain partners, governments, and other capital market participants, use ESG to assess corporate risks and corporate sustainability performance (Aguilera et al., 2007; Bai & Chang, 2015; Campbell, 2007; Garriga & Mele, 2004) and drive investors and banks to invest in firm with high ESG performance (Riedl & Smeets, 2017; Huang, Bui, et al., 2024). Previous studies examine the factors contributing to ESG disclosure including the role of government (Lin et al., 2021; Weber, 2014) and institutional ownership (Azar et al., 2018; Cheng et al., 2022), among others. Moreover, several studies examine the effect of ESG disclosure on various stakeholders' decisions and corporate performance (Jun et al., 2022; Yuan et al., 2022; Chai et al., 2023; Rau & Yu, 2024; Tang et al., 2024). Interestingly, Edmans (2023) articulates ESG to be irrelevant in guiding long-term investment decisions.

Analysts enhance the efficiency of information transfer (Huang et al., 2018) and improve the operational efficiency of the capital market (Loh

& Stulz, 2018). Although analysts primarily rely on publicly disclosed financial information in the earnings forecasting process (Jackson, 2022), their attention to non-financial information has gradually increased. ESG information disclosed by firms, as a form of non-financial information, provides incremental insights into corporate sustainability and future value, serving as an effective complement to financial information (Dhaliwal et al., 2012). Analysts' attention to ESG information has significantly increased (Baldini et al., 2018; Roger, 2024). Thus, ESG disclosure serves as a critical reference for analysts' earnings forecasting behavior, highlighting the need for further research on the relationship between ESG disclosure and analyst earnings forecast accuracy (AEFA).

The literature, however, underexplores the relation between ESG disclosure and analysts and pays even less attention to the effect of ESG disclosure on AEFA. The exceptions are Luo and Wu (2022), who find that a firm's having a high ESG rating leads to better AEFA in a sample of US firms, and Liu et al. (2024), who document that a large disagreement on ESG ratings from different rating agencies translates into worse AEFA in China. Given the important role of analysts in the capital market (Driskill et al., 2020; Schipper, 1991), there is room to further

* Corresponding author.

E-mail addresses: 20221800351@imut.edu.cn (G. Mu), hanfei@imut.edu.cn (F. Han).

<https://doi.org/10.1016/j.irfa.2025.104332>

Received 29 November 2024; Received in revised form 4 May 2025; Accepted 7 May 2025

Available online 11 May 2025

1057-5219/© 2025 Elsevier Inc. All rights are reserved, including those for text and data mining, AI training, and similar technologies.

investigate the relation between ESG disclosure and AEFA. We further motivate our study from the recent findings of a nonlinear effect of a corporate social network on corporate social responsibility (Zhao & Chan, 2023). In addition, we argue that ESG disclosure has a two-sided effect on the AEFA. In the short run, ESG information disclosed by firms force to strengthen internal governance (Cabaleiro-Cervino & Mendi, 2024; Chubchuwong, 2019; Garavaglia et al., 2024; Park, 2024) and improve the quality of disclosure, which improves AEFA. However, in the long term, excessive ESG disclosure may lead to negative net present value (NPV) projects and elevate business risks (Bartkus et al., 2002). Furthermore, analysts' cognitive resources are inherently scarce and constrained (DellaVigna & Pollet, 2009) and thus excess ESG disclosure is a burden to analysts. Both scenarios suggest an increase in firm opacity and more difficulty for analysts in interpreting firm information when there are more ESG disclosure, leading to a decrease in AEFA.

The examination on the nonlinear relationship between ESG disclosure and AEFA in China is mainly based on the following considerations. First, compared to European and American countries, emerging market countries, such as China, started ESG development relatively late (Zhang et al., 2023), and do not have a strict regulatory system as well as a unified and systematic ESG disclosure standard. Hence, there is a lack of effective verification of the authenticity and credibility of ESG reports issued by companies (Zhou et al., 2021), as well as an imbalance between "quantitative coverage" and "quality assurance" of ESG disclosure. Consequently, analysts have limitations to screen information and analyze ESG data. Hence, it is informative to explore the non-linear impact of ESG disclosure on AEFA in an emerging market such as China. Second, compared to developed markets, the high participation of retail investors in the Chinese stock market suggests that information dissemination exhibit time lags (Wu et al., 2024), suggesting that analysts' reports have become a key source of information that influences investors' decision-making (Xu et al., 2025). Thus, AEFA is a core reference for investors' decision-making. Therefore, our findings offer valuable lessons for other markets with similar high participations of retail investors. Finally, China, as one of the economies with the largest number of listed companies in the world, provides large-scale sample data for this study. The data advantage is helpful to accurately portray the complex nonlinear relationship between ESG disclosure and AEFA.

In light of this, using a sample of Chinese firms from 2007 to 2022, we find that ESG disclosure has a U-shaped association with AEFA. The findings remain intact after a battery of robustness checks. Specifically, when ESG disclosure increases, AEFA increases with a decreasing rate. Then, when the ESG disclosure level rises beyond approximately 82nd percentiles of the distribution, AEFA reaches a turning point from increased AEFA to decreased AEFA. Furthermore, the mechanism test reveals that earnings management and media attention mediate the U-shaped relation between ESG disclosure and AEFA, which corroborates the logic underlying the nonlinear relationship. Additional analysis suggests that the U-shaped relation is more salient for state-owned firms and for firms in non-polluting industries.

We make two contributions to the literature. First, this study enriches the research on the impact of non-financial information on AEFA by shifting the perspective to the "two-sidedness" of non-financial information using the lens of ESG. Several studies examine the impact of financial information (De Franco et al., 2011; Dhaliwal et al., 2012; Jackson, 2022) and non-financial information (e.g., forward-looking information, R&D and innovation information) (Beyhaghi et al., 2023; Gu & Wang, 2005; Vanstraelen et al., 2003) on analysts' forecasting behavior, mostly based on a single dimension, whereas this paper focuses on the "double-edged sword" of ESG disclosure. The U-shaped relation suggests that excessive ESG disclosure is not helpful to analysts in their earnings forecast. We argue that the nonlinear relation between ESG disclosure and AEFA partially explains why previous studies had mixed findings for the effect of ESG disclosure on corporate performance (Baker et al., 2021; Brammer et al., 2006; Rau & Yu, 2024; Yuan et al., 2022).

Second, this study not only extends the logical chain between ESG disclosure and AEFA but also provides new perspectives and insights for the effective operation of capital markets. On the one hand, while most of the previous studies focus on a single and linear internal governance perspective (e.g., information quality and operational risk) to analyze the relationship between the two (Luo & Wu, 2022; Wang & Xing, 2023), our study shows that ESG disclosure indirectly reduces AEFA in a nonlinear way through its effect on firms' internal earnings management as well as external media attention. This finding reveals a more complex and multidimensional interaction mechanism between ESG disclosure and firms' financial performance. On the other hand, in the context of complex capital market information, ESG disclosure should not be regarded as a purely ethical proposition. Instead, it is useful to improve the professionalism and information processing ability of media, analysts, and investors to help them identify the "halo effect traps" that may exist in ESG high scoring companies more prudently, and avoid making wrong investment decisions due to blind pursuit of high ESG score.

The remainder of this paper is organized as follows. The following section describes the literature review and research hypotheses. Section 3 describes the data sources and the model. In Section 4, the results of the empirical investigation are detailed, including the base results, quantile regression, endogeneity test, robustness check, mechanism analysis, and the results of heterogeneity tests. The conclusions and contributions of this study follow in Section 5.

2. Literature review and research hypotheses

There are two strands of literature that carry implications for the impact of ESG disclosure on AEFA. The first strand finds ESG disclosure has a positive effect on AEFA. ESG disclosure sends positive signals of a stable and healthy firm to stakeholders through multiple information channels, establishes a responsible and positive corporate image (Babik & Trendafilova, 2011; Cho et al., 2013; Naqvi et al., 2021; Wong et al., 2021), and thus brings reputational capital to the firm (Cui et al., 2018; Goss & Roberts, 2011). Consequently, ESG disclosure reduces information asymmetry (Flammer, 2013; Ghoul et al., 2011; Naqvi et al., 2021; Yu et al., 2018; Zumente & Lăce, 2021), especially information related to a firm's business risk (Lisin et al., 2022). All else being the same, when firms disclose ESG information, analysts have a larger information set to conduct research that leads to an improved AEFA. From the perspective of firms' external monitoring, firm watchdogs, such as investors and the media, use ESG information disclosed by firms to improve the quality of disclosure (Cabaleiro-Cervino & Mendi, 2024; Chubchuwong, 2019; Garavaglia et al., 2024; Park, 2024). If the disclosed information misrepresents the firm, the firm faces the risk of punishment from regulatory authorities. Therefore, for the sake of sustainable development, firms will try their best to disclose high-quality information. Hence, analysts have a reliable source of earnings forecasting information to reduce the analysts' errors in judgment about the firm's financial information (Shaw, 2003). Overall, with an expanded information set and an improved information environment, analysts produce better AEFA.

Second, ESG disclosure is costly (Jun et al., 2022). ESG investments take a long time to produce tangible benefits (Cai et al., 2016; Sun & Gunia, 2018). Such uncertainty in future returns results in a challenge for analysts to evaluate the net benefits of a firm's ESG investment. Consequently, AEFA worsens. Huang (2022, p. 1983) summarizes three possible relations between ESG disclosure and AEFA. He states:

"First, from a valuation viewpoint, it is not clear how ESG factors are incorporated into earnings forecasts. Second, referencing agency and signaling theories, ESG may reduce information asymmetry between management-agents and shareholder-principals. Third, stakeholder and institutional theories situate ESG in a complex multi-actor and multi-interest environment."

The direct evidence on the relation by Liu et al. (2024) and Luo and Wu (2022) supports a positive effect for the impact of ESG disclosure on

AEFA. However, the depicted relation is a linear one. ESG has multiple actors and multiple interests among various stakeholders (Huang, 2022; Rau & Yu, 2024). The relation between ESG disclosure and AEFA is hardly a linear one.

We argue that analysts find information released in the early stage of ESG disclosure (up to a moderate level of ESG disclosure) to be particularly useful. At this stage, even if managers conduct ESG activities to seek private benefits (Buerthey et al., 2020; Putra et al., 2021), they take a “wait and see” approach due to the monitoring effect of ESG. Such monitoring restrains managers’ opportunistic activities (Eccles et al., 2014; Kim et al., 2012; Lopatta et al., 2016; Ng et al., 2021; Waddock & Graves, 1997), and the additional information provided by ESG disclosure suggests that analysts can fully take advantage of the governance benefits of ESG (Dhaliwal et al., 2012; Yu et al., 2018). All else being the same, when firms disclose from low to a moderate level of ESG information, AEFA improves.

When a firm’s ESG disclosure continues to a high level, it is likely due to the firm’s overinvestment in ESG (Barnea & Rubin, 2010; Becchetti et al., 2013). The high cost and uncertain nature of ESG become non-trivial. According to agency theory, overinvestment in ESG imposes additional costs on firms (Jensen & Meckling, 1976; Khanchel & Lassoued, 2022) and even generates negative net present value (Bartkus et al., 2002). Firms’ investment in negative net present value projects increases operational risk (Godfrey, 2005) and uncertain future operating income. Thus, AEFA decreases when analysts consider a firm facing high operating risk (Becchetti et al., 2013; Jennings et al., 2017). Moreover, continuous, and extensive disclosure of ESG information overloads the market with information. According to the limited attention conceptual framework, analysts’ cognitive resources are limited, and they, like average investors, will be constrained by their limited attention. Hence, analysts are unable to focus on all the information disclosed (DellaVigna & Pollet, 2009; Hirshleifer et al., 2009). Therefore, analysts focus on certain positive information about the firm and downplay negative information. Accordingly, they overestimate a firm’s value, which triggers inaccurate earnings forecasts.

Based on the above analysis, we expect that ESG disclosure will affect AEFA; further, this effect is not linear, but rather, has a U-shaped relation. Accordingly, we propose the hypothesis:

H1. ESG disclosure has a U-shaped relation with analyst earnings forecast accuracy.

3. Data sources and modeling

3.1. Sample selection

We use A-share listed firms in China from 2007 to 2022 as the research sample. The Bloomberg database started to disclose ESG information of Chinese companies in 2006, and we lag the explanatory variables by one period, so this study starts from 2007. We exclude financially distressed firms (i.e., companies in “ST,” “*ST,” or “PT” status), firms belonging to financial industries, analysts’ earnings forecast announcements whose release dates are later than the deadline of firms’ annual reports, and firms with missing values in control variables. Typically, many analysts follow a firm. Thus, we only keep the most recent earnings forecast of each analyst per year. The final sample has 10,372 firm-year observations. The ESG disclosure data are obtained from the Bloomberg database, and the rest of the data are from the China Stock Market and Accounting Research (CSMAR) database. We winsorize all continuous variables at the 1 % and 99 % levels.

3.2. Dependent variable: analyst earnings forecast accuracy (*FERROR1*)

Following Hong and Kacperczyk (2010), Abernathy et al. (2013), and Haw et al. (2015), we construct the AEFA metric and denote it as *FERROR1*.

$$FERROR1_{i,t} = \frac{|FEPS_{i,t} - MEPS_{i,t}|}{Price_{i,t-1}} * 100 \quad (1)$$

where $FEPS_{i,t}$ represents the median of all analysts’ forecasts for firm i in year t , $MEPS_{i,t}$ is the actual EPS of firm i in year t , and $Price_{i,t-1}$ is the opening price of firm i ’s stock at the beginning of year t . A small value of *FERROR1* suggests a better AEFA.

3.3. Empirical models

To investigate whether there is a U-shaped link between ESG disclosure and analysts’ earnings projection bias, we use the following econometric model:

$$FERROR1_{i,t} = \alpha_0 + \alpha_1 ESG_{i,t-1}^2 + \alpha_2 ESG_{i,t-1} + \alpha_3 Controls_{i,t} + Year + Industry + \varepsilon_{i,t} \quad (2)$$

where *ESG* denotes ESG disclosure and *Controls* refer to a set of control variables. Following Berkan et al. (2021) and Christensen et al. (2022), we use firm size (*SIZE*), financial leverage (*LEV*), income growth of a firm (*GROWTH*), operating loss (*LOSS*), book-to-market ratio (*BM*), firm age (*LISTAGE*), board size (*BOARD*), board diligence (*MEETING*), top five shareholders’ shareholding ratio (*TOP5*), analyst attention (*ANALYST*), analyst forecast ranges (*HORIZON*) and the number of analyst forecast revisions (*TIMES*). Table 1 presents detailed definitions of all variables. *Year* and *Industry* represent the year- and industry-fixed effects, respectively, and ε is the residual term. Furthermore, *ESG* is in year $t - 1$ because rating agencies typically release their reports at the end of year t and thus analysts use ESG disclosure at $t - 1$ to conduct their research.

Given that the findings from fixed-effects regression (model (2)) are limited to the sample’s average effect, as a robustness check, we also use a quantile regression (QR) model to investigate the potential impact of ESG disclosure on the bias of analysts’ earnings estimates in various quantiles. We use quantile regression (QR) model (Koenker & Bassett Jr, 1978) to analyze the effects of each explanatory variable on AEFA in different quantiles. It benefits in investigating, accurately, the determinants of AEFA. QR model assists in understanding the effects of each variable by looking at the sign of the coefficient, size and level of significance across the distribution (Wu & Chang, 2022). The QR model is:

Table 1

Variable definitions.

This table presents definitions of all variables.

Symbol	Definition
<i>FERROR1</i>	The absolute value of the difference between the median EPS forecasts by analysts and the actual EPS for the year in a firm, divided by the firm’s stock price at the beginning of the period. We multiply it by 100 for easy interpretation.
<i>ESG</i>	Bloomberg ESG Score/100
<i>SIZE</i>	The natural logarithm of total assets
<i>LEV</i>	The ratio of total liabilities to total assets
<i>GROWTH</i>	(Current year’s operating income/previous year’s operating income) – 1
<i>LOSS</i>	Dummy variable that takes a value equal to one if the firm reports a loss during the quarter
<i>BM</i>	The ratio of market value to book value
<i>LISTAGE</i>	The natural logarithm of the difference of current year and year of listing plus 1
<i>BOARD</i>	The natural logarithm of the number of board members plus 1
<i>MEETING</i>	The natural logarithm of the number of board meetings
<i>TOP5</i>	Shareholding of top five shareholders to total outstanding shares
<i>ANALYST</i>	The natural logarithm of the number of analysts following plus 1
<i>HORIZON</i>	The natural logarithm of the average number of days between the date of release of analysts’ forecasts and the date of disclosure of the company’s financial report
<i>TIMES</i>	The number of times an analyst’s earnings forecast for a company is revised in the current year and averaged out

$$FERROR1_{i,t,q} = \alpha_0 + \alpha_1 ESG_{i,t-1}^2 + \alpha_2 ESG_{i,t-1} + \alpha_3 Controls_{i,t} + Year + Industry + \varepsilon_{i,t} \quad (3)$$

where all variables are defined as in model (2), and q denotes various quantiles. We focus on the results of five quantile regressions at the 10th percentile, the 25th percentile, the 50th percentile, the 75th percentile, and the 90th percentile of AEFA.

4. Empirical results

4.1. Descriptive statistics

Table 2 shows the descriptive statistics of the sample. The mean of *FERROR1* is 2.5424 and the standard deviation is 4.0646, indicating that AEFA is volatile. The mean of *ESG* is 0.2665 with a standard deviation of 0.0953, suggesting that there are differences in ESG disclosure among firms. The descriptive statistics of the other main variables are in line with the literature.

4.2. Baseline regression analysis

4.2.1. Baseline regression

Column (1) of Table 3 reports the results of the baseline regression on the effect of ESG disclosure on AEFA. The coefficient of ESG^2 is 6.4716 and significant at the 5 % level while the coefficient of *ESG* is -6.0058 and significant at the 1 % level. Thus, the effect of ESG disclosure on AEFA exhibits a U-shaped relation, that is, as a firm increases its ESG disclosure, AEFA increases at a decreasing rate. The findings support H1.

Drawing from Lind and Mehlum (2010), we not only find the point of inflection lies within the range of values of ESG, but also that the slope of the curve is significantly negative when the explanatory variable takes a minimum value and is significantly positive when the explanatory variable takes a maximum value. In addition, we analyze the U-shaped relationship using the Stata *Utest* command, and the results confirmed that the U-shaped curve relationship is statistically significant.

Thus, the effect of ESG disclosure is favorable for improving AEFA when ESG increases from a low to a moderately high level. When ESG rises to a high level, AEFA deteriorates.

4.2.2. Quantile regression

Columns (2)–(6) of Table 3 show the regression results for the effect of ESG disclosure on AEFA using a quantile regression method. Specifically, we present the findings at various percentiles of AEFA (10th, 25th, 50th, 75th, and 90th), respectively. We note several interesting results. First, the regression coefficients of *ESG* are negatively significant while those of ESG^2 are positively significant across columns (2) to (6). In addition, the magnitudes of the coefficients of ESG^2 steadily rise at the

five AEFA quantiles, which are 0.5781, 1.4494, 3.9721, 9.8688, and 20.5323, respectively. Similarly, those of *ESG* also steadily rise in absolute values from 0.5153, 1.3739, 3.2680, 7.0997, and 13.7008, respectively, in the five AEFA quantiles. Thus, in general, if we confine to a linear model (no ESG^2 in Eq. (3)), AEFA monotonically improves when a covered firm's ESG disclosure increases. After incorporating ESG^2 in Eq. (3), the improvement of AEFA increases in ESG at a decreasing rate. Second, except for the 10th percent quartile, the coefficients of ESG^2 are significant at the 1 % and 5 % levels, suggesting the nonlinear U-shaped feature is persistent over the distribution of ESG disclosure at the higher quantile of AEFA.

In the last row of Table 3, we calculate the magnitudes of the combined effect of *ESG* and ESG^2 on *FERROR1* using the value of ESG disclosure at various percentiles (10th, 25th, 50th, 75th, and 90th) of *FERROR1*.¹ We notice the values of *FERROR1* are negative at the 10th, 25th, 50th, and 75th percentiles of ESG but move to positive at the 90th percentile. That is, as ESG disclosure increases, AEFA improves until *ESG* hits a threshold point at around the 82nd percentile, and then AEFA worsens.² Fig. 1 fits the results of Table 3 column (1). The profile of AEFA exhibits a U-shaped relation between ESG disclosure and AEFA.

The coefficients of control variables, if significant, carry the expected signs. Consistent with intuition, the coefficients of *LEV*, *LOSS*, *BM*, *MEETING* and *HORIZON* are positively significant while those of *GROWTH*, *LISTAGE*, *BOARD*, *TOP5*, *ANALYST*, and *TIMES* are negatively significant.

4.3. Endogeneity analyses

4.3.1. Instrumental variables approach

Given the possibility of endogeneity in the relation between ESG disclosure and AEFA, we use an instrumental variable (IV) approach to mitigate the concern. We use ESG_{t-2} as an IV. ESG_{t-2} is positively correlated with ESG_{t-1} but analysts prefer to use ESG disclosure that is more recent (at $t-1$) instead of farther away disclosures (at $t-2$). Thus, the IV meets the relevancy and exclusion criteria. We present the results in Table 4. Columns (1) and (2) of the first-stage regression results show that the coefficients of the two IVs (ESG^2_{IV} and ESG_{IV}) are good predictors of ESG^2 and *ESG*. The corresponding weak instrument test *F*-statistics suggest that ESG^2_{IV} and ESG_{IV} are not weak instruments. Then, we obtain the predicted values of ESG^2 and *ESG*. The results in column (3) show that the coefficients of predicted ESG^2 and *ESG* remain qualitatively similar to those in Table 3. Thus, the U-shaped relationship between ESG disclosure and analyst earnings forecast accuracy remains intact using the two-stage least squares method.

4.3.2. Heckman two-stage regression

For robustness, we use Heckman two-stage regression to mitigate potential selectivity bias. Specifically, we group the ESG scores of firms based on whether they disclose their ESG scores to construct a dummy variable, *ESG_DUM* (when a firm has ESG score, it equals 1, and 0 otherwise at $t-1$) as the dependent variable in the selection equation and run a probit regression. We follow Ang et al. (2000) to use a firm's sale expense to revenue ratio (*SALE_EXP*) to proxy the extent of agency conflict in a firm and conjecture that if there is high agency conflict in a firm, it is more likely to disclose ESG information. Thus, *SALE_EXP* helps identify *ESG_DUM*. Then, we recover the inverse Mills ratio (*IMR*) and augment Eq. (2) with it. The results of Table 5 show that the coefficients of ESG^2 and *ESG* are positively and negatively significant, respectively and are consistent with those in Table 3.

Table 2

Summary statistics.

This table presents the summary statistics of the sample. Table 1 presents definitions of all variables.

Variables	N	Mean	Median	Std Dev	Min	Max
<i>FERROR1</i>	10,372	2.5424	1.1280	4.0646	0.0000	35.6268
<i>ESG</i>	10,372	0.2665	0.2630	0.0953	0.0909	0.6182
<i>SIZE</i>	10,372	23.2781	23.1663	1.3511	19.8964	27.9424
<i>LEV</i>	10,372	0.4771	0.4901	0.1927	0.0444	0.9158
<i>GROWTH</i>	10,372	0.1925	0.1316	0.3734	-0.5915	5.0089
<i>LOSS</i>	10,372	0.0582	0.0000	0.2342	0.0000	1.0000
<i>BM</i>	10,372	0.6528	0.6553	0.2758	0.0607	1.3013
<i>LISTAGE</i>	10,372	2.4100	2.5649	0.6642	0.6931	3.4012
<i>BOARD</i>	10,372	2.1933	2.1972	0.2058	1.3863	2.7081
<i>MEETING</i>	10,372	2.2529	2.1972	0.3984	1.0986	3.4657
<i>TOP5</i>	10,372	0.5644	0.5628	0.1596	0.1856	0.9309
<i>ANALYST</i>	10,372	2.1896	2.3979	1.0835	0.0000	4.0943
<i>HORIZON</i>	10,372	6.2965	6.3425	0.2926	4.6085	6.8902
<i>TIMES</i>	10,372	1.2825	1.2143	0.7344	0.0000	4.4300

¹ If we take partial derivative of Eq. (2) with respect to *ESG*, we have $2\alpha_1(ESG) + \alpha_2$. By putting in *ESG* values at various percentiles, we can calculate the point estimate of *FERROR1*.

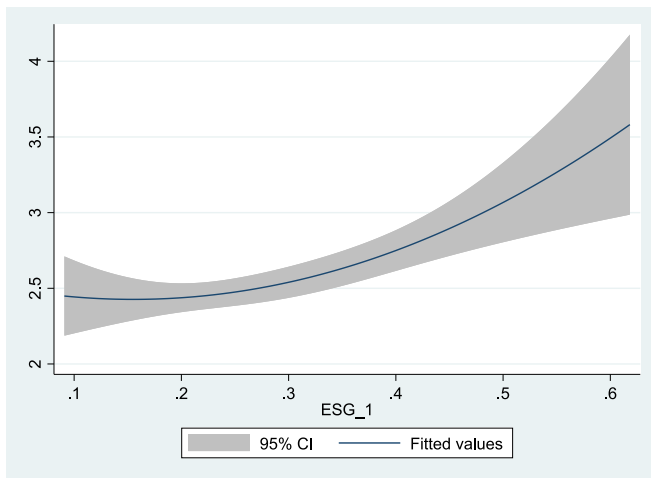
² We also examine Eq. (2) without ESG^2 . The coefficient of *ESG* is -1.7176 .

Table 3

Fixed effect and quantile regression results.

This table presents the results for the impact of ESG on analyst earnings forecast accuracy (*FERRORI*). Column (1) is the fixed effect regression. Columns (2) to (6) present the results of quantile regressions at various percentiles of *FERRORI* (10th, 25th, 50th, 75th, and 90th). The last row presents the point estimates for the combined impact of ESG on *FERRORI* valuated at various percentiles of *FERRORI*. *t*-statistics are in parentheses. Table 1 presents definitions of all variables. ***, **, and * indicate 1 %, 5 %, and 10 % significance, respectively.

Variables	<i>FERRORI</i>					
	(1)	Quantile				
	Fixed effect	10	25	50	75	90
ESG_{t-2}	6.4716** (2.29)	0.5781 (1.38)	1.4494** (2.39)	3.9721*** (3.73)	9.8688*** (4.16)	20.5323*** (3.74)
ESG_{t-1}	-6.0058*** (-3.25)	-0.5153* (-1.72)	-1.3739*** (-3.52)	-3.2680*** (-5.17)	-7.0997*** (-5.39)	-13.7008*** (-4.15)
SIZE	-0.0043 (-0.09)	-0.0093 (-1.10)	-0.0136 (-1.19)	-0.0220 (-1.26)	0.0613 (1.49)	0.2732*** (3.04)
LEV	1.1928*** (5.07)	0.0391 (1.12)	0.1788*** (3.36)	0.5380*** (5.15)	0.9201*** (5.53)	0.9683** (2.52)
GROWTH	-0.5001*** (-5.22)	-0.1150*** (-5.47)	-0.2199*** (-8.60)	-0.3700*** (-6.36)	-0.4367*** (-7.35)	-0.5049*** (-4.99)
LOSS	7.8435*** (25.00)	2.1919*** (10.12)	3.8470*** (18.02)	7.0333*** (13.78)	11.4653*** (19.03)	17.2643 *** (26.12)
BM	2.3753*** (12.51)	0.2475*** (7.50)	0.6394*** (15.45)	1.3901*** (19.15)	2.2833*** (17.66)	3.3503*** (10.59)
LISTAGE	-0.1848*** (-3.66)	-0.0227** (-2.59)	-0.0790*** (-5.64)	-0.1555*** (-6.41)	-0.2499*** (-6.53)	-0.2708** (-3.05)
BOARD	-0.5729*** (-3.47)	-0.0159 (-0.59)	-0.1347*** (-2.96)	-0.3302*** (-4.81)	-0.5923*** (-4.46)	-0.9262*** (-3.36)
MEETING	0.3793*** (4.21)	0.0527*** (3.34)	0.0575*** (2.73)	0.0905** (2.60)	0.0826 (1.25)	0.4083** (2.41)
TOP5	-0.8200*** (-3.59)	-0.1437*** (-3.57)	-0.3329*** (-6.16)	-0.5911*** (-6.08)	-0.9332*** (-4.55)	-1.3756*** (-3.14)
ANALYST	-0.2930*** (-6.79)	-0.0266*** (-3.03)	-0.0652*** (-6.58)	-0.1690*** (-8.62)	-0.4612*** (-12.13)	-0.8279*** (-10.50)
HORIZON	2.9740*** (23.29)	0.3875*** (10.93)	0.8756*** (21.78)	1.6986*** (34.86)	2.7106*** (31.12)	3.6684*** (21.86)
TIMES	-0.0986* (-1.88)	-0.0168** (-1.97)	-0.0283** (-2.33)	-0.0422* (-1.69)	-0.0232 (-0.49)	-0.1637** (-1.99)
Constant	-15.4033*** (-12.58)	-1.9449*** (-8.12)	-4.1165*** (-13.82)	-7.7207*** (-18.20)	-13.2304*** (-14.92)	-20.7570*** (-11.15)
Industry	Yes	Yes	Yes	Yes	Yes	Yes
Year	Yes	Yes	Yes	Yes	Yes	Yes
N	10,372	10,372	10,372	10,372	10,372	10,372
Adj. R ² /Pseudo R ²	0.4420	0.0486	0.1044	0.1815	0.2857	0.3769
F	140.57					
ESG value at the given percentile		0.1529	0.1938	0.2630	0.3145	0.3861
The aggregated effect of ESG and ESG_{t-2}^2 on <i>FERRORI</i>		-0.3385	-0.8121	-1.1787	-0.8922	2.1542

**Fig. 1.** The impact of ESG disclosure on analyst earnings forecast accuracy.

4.4. Robustness tests

4.4.1. Alternative metric of analyst earnings forecast accuracy

For robustness, we use an alternative metric of AEFA. We use the mean of all analysts' most recent EPS forecasts for the year instead of median EPS forecasts to measure AEFA (denoted as *FERROR2*) (Duru & Reeb, 2002). The regression results, as shown in column (1) of Table 6,

Table 4

Instrumental variables approach regression results.

This table presents the results for the impact of ESG on analyst earnings forecast accuracy (*FERRORI*) using an instrumental variable (IV) approach. We use ESG_{t-2} as an IV for ESG_{t-1} . Columns (1) and (2) generate predicted values of ESG_{t-1}^2 and ESG_{t-1} . For brevity, we do not present the coefficients of control variables. *t*-statistics are in parentheses. Table 1 presents definitions of all variables. ***, **, and * indicate 1 %, 5 %, and 10 % significance, respectively.

Variables	First stage		Second stage
	(1)	(2)	(3)
	ESG_{t-1}	ESG_{t-1}^2	<i>FERRORI</i>
ESG_{t-1}^2 (predicted)			10.4653** (2.57)
ESG_{t-1} (predicted)			-9.2088*** (-3.27)
ESG_{t-2}^2	0.4291*** (11.85)	1.1075*** (44.61)	
ESG_{t-2}	0.5776*** (24.21)	-0.1138*** (-8.11)	
Control var.	Yes	Yes	Yes
Constant	-0.0035 (-0.23)	-0.0167* (-1.72)	-19.7675*** (-13.42)
Industry	Yes	Yes	Yes
Year	Yes	Yes	Yes
N	8653	8653	8653
Adj. R ² /Pseudo R ²	0.8835	0.8820	0.4554
F	1075.84	652.85	

Table 5

Heckman two-stage regression results.

This table presents the results for the impact of ESG on analyst earnings forecast accuracy (*FERROR1*) using Heckman two-stage regression. *SALE_EXP* is the ratio of sales expenses to revenue. *ESG_DUM* is a dummy variable with a value of 1 if a firm has ESG score and 0 no such score. *IMR* is the inverse Mills ratio recovered from column (1). For brevity, we do not present the coefficients of control variables. *t*-statistics are in parentheses. Table 1 presents definitions of all variables. ***, **, and * indicate 1 %, 5 %, and 10 % significance, respectively.

Variables	First stage	Second stage
	(1)	(2)
	<i>ESG_DUM</i>	<i>FERROR1</i>
<i>SALE_EXP</i> _{<i>t</i>-1}	0.9840*** (6.73)	
<i>ESG</i> _{<i>t</i>-1} ²		4.4198* (1.65)
<i>ESG</i> _{<i>t</i>-1}		-4.4613** (-2.28)
<i>IMR</i>		-0.0989 (-0.51)
Control var.	Yes	Yes
Constant	-23.8174*** (-44.81)	-14.7389*** (-6.42)
Industry	Yes	Yes
Year	Yes	Yes
<i>N</i>	24,426	10,812
Adj. <i>R</i> ² /Pseudo <i>R</i> ²	0.3830	0.3291
<i>F</i>		94.32

show that the coefficient of *ESG*² remains positively significant at the 10 % level while the coefficient of *ESG* continues to be negatively significant at the 1 % level. Hence, the U-shaped relation between ESG disclosure and AEFA remains.

4.4.2. Alternative metric of ESG disclosure

To control for the potential bias of the measurement error of the explanatory variable, we use ESG data from [Hexun.com](https://www.hexun.com) as an alternative metric for ESG disclosure. The website uses a professional evaluation system that covers the majority of Shanghai and Shenzhen A-share listed firms. The regression results, as shown in column (2) of Table 6, show that the coefficient of *ESG*² continues to be positively significant at the 1 % level, and the coefficient of *ESG* remains negatively significant at the 1 % level, which indicates that the U-shaped relation between ESG disclosure and AEFA remains intact.

Table 6

Robustness test results.

This table presents various robustness results for the impact of ESG on analyst earnings forecast accuracy. In column (1), *FERROR2* = the absolute value of the difference between the average EPS forecasts by analysts and the actual EPS for the year in a firm, divided by the firm's stock price at the beginning of the period multiplied by 100. In column (2), we use ESG data from [Hexun.com](https://www.hexun.com). In column (3), we use a Tobit estimation method. In column (4), we excluded samples with *ANALYST* = 0. For brevity, we do not present the coefficients of control variables. *t*-statistics are in parentheses. Table 1 presents definitions of all variables. ***, **, and * indicate 1 %, 5 %, and 10 % significance, respectively.

Variables	(1)	(2)	(3)	(4)
	<i>FERROR2</i>	<i>FERROR1</i> (Hexun.com ESG data)	<i>FERROR1</i> (Tobit method)	<i>FERROR1</i> (Removal of outliers)
<i>ESG</i> _{<i>t</i>-1} ²	4.5635* (1.65)	0.0003*** (2.84)	6.7581* (2.38)	6.9513** (2.47)
<i>ESG</i> _{<i>t</i>-1}	-5.1147*** (-2.84)	-0.0245*** (-2.87)	-6.1684*** (-3.32)	-5.8756*** (-3.14)
Control var.	Yes	Yes	Yes	Yes
Constant	-15.9733*** (-13.46)	-15.9084*** (-19.36)	-16.7306*** (-13.17)	0.8372** (0.91)
Industry	Yes	Yes	Yes	Yes
Year	Yes	Yes	Yes	Yes
<i>N</i>	10,372	16,968	10,372	9481
Adj. <i>R</i> ² / Pseudo <i>R</i> ²	0.4490	0.4564	0.1053	0.3936

4.4.3. Alternative estimation method

Given that some firms may choose not to disclose ESG information, we use a Tobit model to account for the possible censored sample bias in the sample. Column (3) of Table 6 shows the regression results. The results are qualitatively similar to those in column (1) of Table 3. Support for H1 remains strong.

4.4.4. Removal of outliers

Given the lack of analyst coverage for some firms, we exclude these samples to address possible sample selection bias due to analysts' behavior. Column (4) of Table 6 shows the regression results. The results are qualitatively similar to those in column (1) of Table 3, which support H1.

4.5. Mechanism analysis

4.5.1. ESG disclosure, earnings management, and analysts' earnings forecast accuracy

The monitoring effect of ESG disclosure improves information transparency and increases the possibility of detecting earnings management. Thus, ESG disclosure generally restrains earnings management (Almahrog et al., 2018; Kim et al., 2012; Sun et al., 2024). However, excessive ESG disclosure suggests a firm is concerned about its public image. The firm uses earnings management to paint a positive image of the firm, which lowers AEFA (Bénabou & Tirole, 2010). Taken together, earnings management mediates the U-shaped relation between ESG disclosure and AEFA. That is, when ESG disclosure increases from a low to moderately high level, it restrains earnings management. Accordingly, ESG disclosure has an overall positive effect on AEFA. However, when ESG disclosure is high, the firm uses more earnings management to paint a good image, which makes it challenging for analysts to assess a firm's actual performance (Chaney et al., 2011). In that case, high ESG disclosure contributes to an overall negative effect on AEFA via earnings management. Thus, we argue that earnings management mediates the nonlinear effect of ESG disclosure on AEFA.

In line with Baron and Kenny (1986), we construct a nonlinear mediation effect model. Through a regression analysis, we assess the mediating role of earnings management in H1. The model is specified as follows:

$$AbsREM_{it} = \alpha_0 + \alpha_1 ESG_{it-1}^2 + \alpha_2 ESG_{it-1} + \alpha_3 Controls + Year + Industry + \varepsilon_{it} \quad (4)$$

$$FERROR1_{it} = \alpha_0 + \alpha_1 ESG_{it-1}^2 + \alpha_2 ESG_{it-1} + \alpha_3 AbsDEM_{it} + \alpha_4 Controls + Year + Industry + \varepsilon_{it} \quad (5)$$

where *AbsREM*_{*it*} as the mediating variable, denotes the degree of earnings management for firm *i* in year *t*. Specifically, we construct real activity earnings management indexes from three dimensions of abnormal production cost, abnormal cash flow, and abnormal manipulative expenses, and takes its absolute value (*AbsREM*) to measure the level of real earnings management of enterprises (Cohen et al., 2008; Roychowdhury, 2006). A large value of *AbsREM* suggests a great degree of earnings management. Other variables are the same as those in Eq. (2).

Columns (1) and (2) in Table 6 report the regression results with *AbsREM* as the mediating variable. In column (1) of Table 7, the coefficient of *ESG*² (*ESG*) is positively (negatively) significant at the 5 % (1 %) level. Thus, ESG disclosure and earnings management exhibit a U-shaped relation. That is, ESG disclosure generally restrains earnings management but at a decreasing rate. In column (2) of Table 7, the coefficient of *AbsREM* is positively significant at the 1 % level, while the coefficients of *ESG*² and *ESG* remain significant with the expected signs. These results imply that earnings management plays a mediating role in the relation between ESG disclosure and AEFA.

Table 7

Results of the mediation effect test.

This table presents the results for the mediation effect for the impact of ESG on analyst earnings forecast accuracy. In column (1), we attempted to demonstrate the nonlinear relationship between ESG disclosure and AbsREM. In column (2), we demonstrate that the mediating effect of AbsREM. In column (3), we focus on whether there is a nonlinear relationship between ESG disclosure and MEDIA. In column (4), we demonstrate that the mediating effect of MEDIA. Table 1 presents definitions of all variables. ***, **, and * indicate 1 %, 5 %, and 10 % significance, respectively.

Variables	(1)	(2)	(3)	(4)
	<i>AbsREM</i>	<i>FERROR1</i>	<i>MEDIA</i>	<i>FERROR1</i>
ESG_{t-1}^2	0.2019** (2.28)	5.6319* (1.94)	3.2197*** (5.40)	5.9928** (2.11)
ESG_{t-1}	-0.1893*** (-2.59)	-5.3662*** (-2.77)	-0.8454** (-2.14)	-5.9748*** (-3.22)
<i>AbsREM</i>		1.4684*** (5.59)		
<i>MEDIA</i>				0.1650*** (3.89)
Control var.	Yes	Yes	Yes	Yes
Constant	0.1475*** (4.10)	-15.8491*** (-11.92)	-5.5519*** (-19.94)	-14.6241*** (-11.66)
Industry	Yes	Yes	Yes	Yes
Year	Yes	Yes	Yes	Yes
N	9520	9520	10,254	10,254
Adj. R ² / Pseudo R ²	0.1642	0.4382	0.5470	0.4443

4.5.2. ESG disclosure, media attention, and analysts' earnings forecast accuracy

We argue that as ESG disclosure is a relatively recent phenomenon, and in most cases, firms are not required to disclose their ESG information (Sun et al., 2024).³ These disclosures, if made, lack uniform standards and lead to different ratings (Berg et al., 2022; Christensen et al., 2022). Hence, it is challenging for media to validate the ESG disclosure of a firm due to information clutter and poor comparability, which in turn leads to a decline in media attention. However, as the public's recognition of the ESG concept grows (Sun et al., 2024) and China's regulatory requirements for corporate ESG disclosure improves, disclosures about shareholder governance, risk management, and internal control have received increased media attention (Hahn & Kühnen, 2013). Then, there is a U-shaped relation between ESG disclosure and media attention (*MEDIA*). When ESG disclosure is low (high), media attention shrinks (increases to disproportionately high attention).

However, as media attention increases, at some point, it moves to information overload. An excess of information leads analysts to become distracted and unable to timely react to all public information in the market (DellaVigna & Pollet, 2009; Hirshleifer et al., 2009), resulting in lowered AEFA. To summarize, we propose that media attention mediates the U-shaped relation between ESG disclosure and AEFA in a nonlinear way. That is, ESG disclosure influences media attention via the U-curve effect, which then influences AEFA in a U-shaped relation.

To examine the mediating role of media attention for the effect of ESG disclosure on AEFA, we use Eqs. (4) and (5) but replace *AbsREM* with *MEDIA*, which is the number of media reports of a firm. Following Guo et al. (2018) and Huang, Huang, and Ho (2024), we use the natural logarithm of one plus the number of annual media reports for a firm to measure media attention (*MEDIA*). The data of media reports are obtained from the CNRDS database. A large value of *MEDIA* suggests high media attention. Columns (3) and (4) in Table 7 report the regression results with media attention as the mediating variable. They are qualitatively similar to those in columns (1) and (2). That is, ESG disclosure and media attention

exhibit a U-shaped relation: A low ESG disclosure level contributes to less media attention. As ESG disclosure increases, media attention increases disproportionately. In column (4), the coefficient of *MEDIA* is positively significant at the 1 % level, while the coefficients of ESG^2 and *ESG* remain significant with the expected signs. These results imply that media attention plays a mediating role in the relation between ESG disclosure and AEFA in a nonlinear framework.

4.6. Heterogeneity tests

4.6.1. State ownership

Many firms in China are state-owned. State-owned firms (SOEs) face higher political pressures than non-SOEs, and in addition to economic objectives, they must also bear the duty of achieving job growth and other social objectives (Zhang & You, 2024). We expect that SOEs proactively leverage ESG disclosure in the spirit of serving the societal good. However, SOEs also make ESG disclosure to convey a positive image to their supervising authorities and the public. At the margin, SOEs are more likely than non-SOEs to behave in a way predicted by H1. Accordingly, the nonlinear effect of ESG disclosure on AEFA remains salient. In contrast, non-SOEs consider survival and wealth maximization as their primary objectives (Luo et al., 2024; Zhang & You, 2024). ESG disclosure is not necessarily a priority. Such disclosure from non-SOEs may or may not mean anything to analysts. Accordingly, the nonlinear effect of ESG disclosure on AEFA is weak.

We present the findings in columns (1) and (2) of Table 8. The coefficients of ESG^2 and *ESG* continue to be positively and negatively significant as those in Table 3 in column (1), while the same coefficients are insignificant in column (2).

4.6.2. Industry heterogeneity test

As a developing country, China has many high-pollution industries. Firms in these industries produce heavy pollutants and consume a lot of energy. They naturally face greater legitimacy pressure and public scrutiny. As a strategy response, heavy-polluting firms are motivated to disclose more higher-quality environmental information to minimize fines (Bayoud et al., 2012; Gamerschlag et al., 2011; Patten, 2002; Reverte, 2009). In contrast, firms in non-heavy-polluting industries have less ecological impact, and the intensity of external attention and

Table 8

Results of heterogeneity analysis.

This table presents the results of moderating variables for the impact of ESG on analyst earnings forecast accuracy. In columns (1) and (2), we form subsamples of state-owned (SOEs) and non-state-owned firms (non-SOEs). In columns (3) and (4), we decompose the full samples into subsamples of heavy- and non-heavy-polluting firms. For brevity, we do not present the coefficients of control variables. *t*-statistics are in parentheses. Table 1 presents definitions of all variables. ***, **, and * indicate 1 %, 5 %, and 10 % significance, respectively.

Variables	Property rights		Industry	
	(1)	(2)	(3)	(4)
	SOE	Non-SOE	Heavy polluting	Non-heavy polluting
ESG_{t-1}^2	8.9944** (2.24)	3.0808 (0.77)	3.5320 (0.76)	7.2897** (1.93)
ESG_{t-1}	-7.2927*** (-2.81)	-3.7673 (-1.46)	-3.0574 (-0.93)	-7.1824*** (-3.13)
Control var.	Yes	Yes	Yes	Yes
Constant	-18.8869*** (-10.19)	-16.8305*** (-8.66)	-13.4778*** (-5.98)	-16.5286*** (-11.39)
Industry	Yes	Yes	Yes	Yes
Year	Yes	Yes	Yes	Yes
N	5599	4549	3576	6796
Adj. R ² / Pseudo R ²	0.4642	0.4323	0.4491	0.4328

³ <https://www.forbes.com/sites/jimdeloach/2023/09/13/esg-reporting-is-already-required-heres-what-you-need-to-know/>.

environmental supervision on them is relatively low (Wang et al., 2023). Therefore, non-heavy-polluting firms face little or no environmental scrutiny.

We reexamine Eq. (2) by subsamples of firms in heavy-polluting industries and non-heavy-polluting industries. We classify the full sample into heavily polluting industries and non-heavily polluting industries based on the Guidelines on Industry Classification of Listed Companies revised by the China Securities Regulatory Commission in 2012, the Management List of Industry Classification for Environmental Verification of Listed Companies formulated by the Ministry of Environmental Protection in 2008 (Letter of Environmental Affairs [2008] No. 373), and the Guidelines on Environmental Information Disclosure of Listed Companies (Letter of Environmental Affairs [2010] No. 78). The sample was then divided into two groups for regression analysis. The results in columns (3) and (4) of Table 8 show that the nonlinear relation between ESG disclosure and AEFA remains intact for firms in non-heavy-polluting industries while that for heavy-polluting industries is insignificant.

5. Conclusions and implications

ESG disclosure is a relatively new concept in China through which firms comprehensively consider environmental concerns, social responsibility, and corporate governance in their operations and development strategies. Analysts, who act as important participants in the capital market, play a dual role in researching and disseminating information about covered firms to the public.

We investigate the effects of ESG disclosure on AEFA using a nonlinear framework. Using a sample of A-share Chinese firms from 2007 to 2022, we document that firm ESG disclosure has a U-shaped relation with AEFA. The results remain valid to an array of robustness checks. Additional analysis suggests that earnings management and media attention mediate the nonlinear impact of ESG disclosure on AEFA. Cross-sectional tests reveal the effects are more salient for SOEs and those in non-heavy polluting industries.

Our findings carry several implications. First, it would be beneficial for the government to provide guidance to regulators for formulating ESG disclosure policies. Up to a moderate level of ESG disclosure helps analysts provide accurate earnings forecasts, thereby enhancing capital market development. However, when firms have high ESG disclosure levels, analyst forecast accuracy drops, which harms capital market development. Thus, it would be useful for regulators to develop a complete set of mandatory ESG disclosure policies to encourage ESG disclosure but restrain excessive ESG disclosure.

Second, we demonstrate that analysts need to realize the implications of excessive ESG disclosure. Analysts should remain objective and make reasonable use of ESG information disclosed by firms during the earnings forecasting process to avoid increased earnings forecast bias.

Finally, it would be helpful for investors to evaluate a firms' ESG disclosure and analyst earnings forecasts with an understanding of the various levels of ESG disclosure. Specifically, when investors evaluate firms with low level of ESG, analysts' earnings forecasts are useful because of low AEFA. However, when investors examine firms with high level of ESG, analysts' earnings forecasts are not useful because of high AEFA.

Acknowledgments

Funding National Natural Science Foundation of China Project (No.: 72162028); Project of Basic Scientific Research Business Expenses of Directly Affiliated Universities of Inner Mongolia Autonomous Region (No.: JY20220325); Inner Mongolia Autonomous Region Philosophy and Social Sciences Planning Project (No.: 2024NDC214); Research Start-up Fund Project of Inner Mongolia University of Technology (No.: BS2024044).

Data availability

No

References

- Abernathy, J. L., Herrmann, D., Kang, T., & Krishnan, G. V. (2013). Audit committee financial expertise and properties of analyst earnings forecasts. *Advances in Accounting*, 29(1), 1–11.
- Aguilera, R. V., Rupp, D. E., Williams, C. A., & Ganapathi, J. (2007). Putting the S back in corporate social responsibility: A multilevel theory of social change in organizations. *Academy of Management Review*, 32(3), 836–863.
- Almahrog, Y., Ali Aribi, Z., & Arun, T. (2018). Earnings management and corporate social responsibility: UK evidence. *Journal of Financial Reporting and Accounting*, 16(2), 311–332.
- Ang, J. S., Cole, R. A., & Lin, J. W. (2000). Agency costs and ownership structure. *The Journal of Finance*, 55(1), 81–106.
- Azar, J., Schmalz, M. C., & Tecu, I. (2018). Anticompetitive effects of common ownership. *Journal of Finance*, 73(4), 1513–1565.
- Babiak, K., & Trendafilova, S. (2011). CSR and environmental responsibility: Motives and pressures to adopt green management practices. *Corporate Social Responsibility and Environmental Management*, 18(1), 11–24.
- Bai, X., & Chang, J. (2015). Corporate social responsibility and firm performance: The mediating role of marketing competence and the moderating role of market environment. *Asia-Pacific Journal of Management*, 32(2), 505–530.
- Baker, E. D., Boulton, T. J., Braga-Alves, M. V., & Morey, M. R. (2021). ESG government risk and international IPO underpricing. *Journal of Corporate Finance*, 67, Article 101913.
- Baldini, M., Maso, L. D., Liberatore, G., Mazzi, F., & Terzani, S. (2018). Role of country- and firm-level determinants in environmental, social, and governance disclosure. *Journal of Business Ethics*, 150, 79–98.
- Barnea, A., & Rubin, A. (2010). Corporate social responsibility as a conflict between shareholders. *Journal of Business Ethics*, 97, 71–86.
- Baron, R. M., & Kenny, D. A. (1986). The moderator–mediator variable distinction in social psychological research: Conceptual, strategic, and statistical considerations. *Journal of Personality and Social Psychology*, 51(6), 1173.
- Bartkus, B. R., Morris, S. A., & Seifert, B. (2002). Governance and corporate philanthropy: Restraining Robin Hood? *Business & Society*, 41(3), 319–344.
- Bayoud, N. S., Kavanagh, M., & Slaughter, G. (2012). Factors influencing levels of corporate social responsibility disclosure by Libyan firms: A mixed study. *International Journal of Economics and Finance*, 4(4), 13–29.
- Becchetti, L., Ciciretti, R., & Giovannelli, A. (2013). Corporate social responsibility and earnings forecasting unbiasedness. *Journal of Banking & Finance*, 37(9), 3654–3668.
- Bénabou, R., & Tirole, J. (2010). Individual and corporate social responsibility. *Economica*, 77(305), 1–19.
- Berg, F., Koelbel, J. F., & Rigobon, R. (2022). Aggregate confusion: The divergence of ESG ratings. *Review of Finance*, 26(6), 1315–1344.
- Berkan, A., Leonardo, B., & Stefano, M. (2021). Media coverage, corporate social irresponsibility conduct, and financial analysts' performance. *Corporate Social Responsibility and Environmental Management*, 28(5), 1456–1470.
- Beyhaghi, M., Khashabi, P., & Mohammadi, A. (2023). Pre-grant patent disclosure and analyst forecast accuracy. *Management Science*, 69(5), 3140–3155.
- Brammer, S., Brooks, C., & Pavelin, S. (2006). Corporate social performance and stock returns: UK evidence from disaggregate measures. *Financial Management*, 35(3), 97–116.
- Buertey, S., Sun, E. J., Lee, J. S., & Hwang, J. (2020). Corporate social responsibility and earnings management: The moderating effect of corporate governance mechanisms. *Corporate Social Responsibility and Environmental Management*, 27(1), 256–271.
- Cabaleiro-Cervino, G., & Mendi, P. (2024). ESG-driven innovation strategy and firm performance. *Eurasian Business Review*, 14(1), 137–185.
- Cai, L., Cui, J., & Jo, H. (2016). Corporate environmental responsibility and firm risk. *Journal of Business Ethics*, 139(3), 563–594.
- Campbell, J. L. (2007). Why would corporations behave in socially responsible ways? An institutional theory of corporate social responsibility. *Academy of Management Review*, 32(3), 946–967.
- Chai, S., Cao, M., Li, Q., Ji, Q., & Liu, Z. (2023). Exploring the nexus between ESG disclosure and corporate sustainable growth: Moderating role of media attention. *Finance Research Letters*, 58, Article 104519.
- Chaney, P. K., Faccio, M., & Parsley, D. (2011). The quality of accounting information in politically connected firms. *Journal of Accounting and Economics*, 51(1–2), 58–76.
- Cheng, X., Wang, H. H., & Wang, X. (2022). Common institutional ownership and corporate social responsibility. *Journal of Banking & Finance*, 136, Article 106218.
- Cho, S. Y., Lee, C., & Pfeiffer, R. J., Jr. (2013). Corporate social responsibility performance and information asymmetry. *Journal of Accounting and Public Policy*, 32(1), 71–83.
- Christensen, D. M., Serafeim, G., & Sikochi, A. (2022). Why is corporate virtue in the eye of the beholder? The case of ESG ratings. *The Accounting Review*, 97(1), 147–175.
- Chubchuwong, M. (2019). The impact of CSR satisfaction on destination loyalty: A study of MICE travelers in Thailand. *Asia Pacific Journal of Tourism Research*, 24(2), 168–179.
- Cohen, D. A., Dey, A., & Lys, T. Z. (2008). Real and accrual-based earnings management in the pre- and post-Sarbanes-Oxley periods. *The Accounting Review*, 83(3), 757–787.
- Cui, J., Jo, H., & Na, H. (2018). Does corporate social responsibility affect information asymmetry? *Journal of Business Ethics*, 148(3), 549–572.

- De Franco, G., Kothari, S. P., & Verdi, R. S. (2011). The benefits of financial statement comparability. *Journal of Accounting Research*, 49(4), 895–931.
- DellaVigna, S., & Pollet, J. M. (2009). Investor inattention and Friday earnings announcements. *Journal of Finance*, 64(2), 709–749.
- Dhaliwal, D. S., Radhakrishnan, S., Tsang, A., & Yang, Y. G. (2012). Nonfinancial disclosure and analyst forecast accuracy: International evidence on corporate social responsibility disclosure. *The Accounting Review*, 87(3), 723–759.
- Driskill, M., Kirk, M. P., & Tucker, J. W. (2020). Concurrent earnings announcements and analysts' information production. *The Accounting Review*, 95(1), 165–189.
- Duru, A., & Reeb, D. M. (2002). International diversification and analysts' forecast accuracy and bias. *The Accounting Review*, 77(2), 415–433.
- Eccles, R. G., Ioannou, I., & Serafeim, G. (2014). The impact of corporate sustainability on organizational processes and performance. *Management Science*, 60(11), 2835–2857.
- Edmans, A. (2023). The end of ESG. *Financial Management*, 52(1), 3–17.
- Flammer, C. (2013). Corporate social responsibility and shareholder reaction: The environmental awareness of investors. *Academy of Management Journal*, 56(3), 758–781.
- Gamerschlag, R., Möller, K., & Verbeeten, F. (2011). Determinants of voluntary CSR disclosure: Empirical evidence from Germany. *Review of Managerial Science*, 5, 233–262.
- Garavaglia, S., Van Landuyt, B. W., White, B. J., & Irwin, J. (2024). The ESG stopping effect: Do investor reactions differ across the lifespan of ESG initiatives? *Accounting, Organizations and Society*, 113, Article 101441.
- Garriga, E., & Mele, D. (2004). Corporate social responsibility theories: Mapping and territory. *Journal of Business Ethics*, 53(1/2), 51–74.
- Ghoul, S., Guedhami, O., Kwok, C. C., & Mishra, D. R. (2011). Does corporate social responsibility affect the cost of capital? *Journal of Banking & Finance*, 35(9), 2388–2406.
- Godfrey, P. C. (2005). The relationship between corporate philanthropy and shareholder wealth: A risk management perspective. *Academy of Management Review*, 30(4), 777–798.
- Goss, A., & Roberts, G. S. (2011). The impact of corporate social responsibility on the cost of bank loans. *Journal of Banking and Finance*, 35(7), 1794–1810.
- Gu, F., & Wang, W. (2005). Intangible assets, information complexity, and analysts' earnings forecasts. *Journal of Business Finance & Accounting*, 32(9–10), 1673–1702.
- Guo, Z., Chan, K. C., & Huang, J. (2018). Can media coverage restrain executive empire building and pursuit of a quiet life? Evidence from China. *International Review of Economics & Finance*, 56, 547–563.
- Hahn, R., & Kühnen, M. (2013). Determinants of sustainability reporting: A review of results, trends, theory, and opportunities in an expanding field of research. *Journal of Cleaner Production*, 59, 5–21.
- Haw, I. M., Hu, B., & Lee, J. J. (2015). Product market competition and analyst forecasting activity: International evidence. *Journal of Banking & Finance*, 56, 48–60.
- Hirschleifer, D., Lim, S. S., & Teoh, S. H. (2009). Driven to distraction: Extraneous events and underreaction to earnings news. *Journal of Finance*, 64(5), 2289–2325.
- Hong, H., & Kacperczyk, M. (2010). Competition and bias. *The Quarterly Journal of Economics*, 125(4), 1683–1725.
- Huang, A. H., Lehavy, R., Zang, A. Y., & Zheng, R. (2018). Analyst information discovery and interpretation roles: A topic modeling approach. *Management Science*, 64(6), 2833–2855.
- Huang, C., Huang, H. Y., & Ho, K. C. (2024). Media coverage and stock liquidity: Evidence from China. *International Review of Economics & Finance*, 89, 665–682.
- Huang, D. Z. (2022). Environmental, social and governance factors and assessing firm value: Valuation, signaling and stakeholder perspectives. *Accounting and Finance*, 62(2), 1983–2010.
- Huang, K. J., Bui, D. G., Hsu, Y. T., & Lin, C. Y. (2024). The ESG washing in banks: Evidence from the syndicated loan market. *Journal of International Money and Finance*, 142, Article 103043.
- Jackson, A. B. (2022). Financial statement analysis: A review and current issues. *China Finance Review International*, 12(1), 1–19.
- Jennings, J., Lee, J., & Matsumoto, D. A. (2017). The effect of industry co-location on analysts' information acquisition costs. *The Accounting Review*, 92(6), 103–127.
- Jensen, M. C., & Meckling, W. H. (1976). Theory of the firm: Managerial behavior, agency costs and ownership structure. *Journal of Financial Economics*, 3(4), 305–360.
- Jun, W., Shiyong, Z., & Yi, T. (2022). Does ESG disclosure help improve intangible capital? Evidence from A-share listed companies. *Frontiers in Environmental Science*, 10, Article 858548.
- Khanchel, I., & Lassoued, N. (2022). ESG disclosure and the cost of capital: Is there a ratcheting effect over time? *Sustainability*, 14(15), 9237.
- Kim, Y., Park, M. S., & Wier, B. (2012). Is earnings quality associated with corporate social responsibility? *The Accounting Review*, 87(3), 761–796.
- Koenker, R., & Bassett, G., Jr. (1978). Regression quantiles. *Econometrica: Journal of the Econometric Society*, 33–50.
- Lin, Y., Fu, X., & Fu, X. (2021). Varieties in state capitalism and corporate innovation: Evidence from an emerging economy. *Journal of Corporate Finance*, 67, Article 101919.
- Lind, J. T., & Mehlum, H. (2010). With or without U? The appropriate test for a U-shaped relationship. *Oxford Bulletin of Economics and Statistics*, 72(1), 109–118.
- Lisin, A., Kushnir, A., Koryakov, A. G., Fomenko, N., & Shchukina, T. (2022). Financial stability in companies with high ESG scores: Evidence from North America using the Ohlson O-Score. *Sustainability*, 14(1), 479.
- Liu, X., Dai, J., Dong, X., & Liu, J. (2024). ESG rating disagreement and analyst forecast quality. *International Review of Financial Analyst*, 95, Article 103446.
- Loh, R. K., & Stulz, R. M. (2018). Is sell-side research more valuable in bad times? *Journal of Finance*, 73(3), 959–1013.
- Lopatta, K., Buchholz, F., & Kaspereit, T. (2016). Asymmetric information and corporate social responsibility. *Business & Society*, 55(3), 458–488.
- Luo, K., & Wu, S. (2022). Corporate sustainability and analysts' earnings forecast accuracy: Evidence from environmental, social and governance ratings. *Corporate Social Responsibility & Environmental Management*, 29(5), 1465–1481.
- Luo, W., Tian, Z., Fang, X., & Deng, M. (2024). Can good ESG performance reduce stock price crash risk? Evidence from Chinese listed companies. *Corporate Social Responsibility and Environmental Management*, 31(3), 1469–1492.
- Naqvi, S. K., Shahzad, F., Rehman, I. U., Qureshi, F., & Laique, U. (2021). Corporate social responsibility performance and information asymmetry: The moderating role of analyst coverage. *Corporate Social Responsibility and Environmental Management*, 28(6), 1549–1563.
- Ng, J., Wu, H., Zhai, W., & Zhao, J. (2021). The effect of shareholder activism on earnings management: Evidence from shareholder proposals. *Journal of Corporate Finance*, 69, Article 102014.
- Park, G. (2024). The impact of performance reporting on investment behavior: Evidence from disclosure reform in the UK. *The Accounting Review*, 99(4), 427–453.
- Patten, D. M. (2002). The relation between environmental performance and environmental disclosure: A research note. *Accounting, Organizations and Society*, 27(8), 763–773.
- Putra, A. A., Mela, N. F., & Putra, F. (2021). Managerial ability and real earnings management in family firms. *Corporate Governance: The International Journal of Business in Society*, 21(7), 1475–1494.
- Rau, P. R., & Yu, T. (2024). A survey on ESG: Investors, institutions and firms. *China Finance Review International*, 14(1), 3–33.
- Reverte, C. (2009). Determinants of corporate social responsibility disclosure ratings by Spanish listed firms. *Journal of Business Ethics*, 88, 351–366.
- Riedl, A., & Smeets, P. (2017). Why do investors hold socially responsible mutual funds? *Journal of Finance*, 72(6), 2505–2550.
- Roger, T. (2024). Do financial analysts care about ESG? *Finance Research Letters*, 63, Article 105289.
- Roychowdhury, S. (2006). Earnings management through real activities manipulation. *Journal of Accounting and Economics*, 42(3), 335–370.
- Schipper, K. (1991). Analysts' forecasts. *Accounting Horizons*, 5(4), 105–121.
- Shaw, K. W. (2003). Corporate disclosure quality, earnings smoothing, and earnings' timeliness. *Journal of Business Research*, 56(12), 1043–1050.
- Sun, W., Chen, S., Jiao, Y., & Feng, X. (2024). How does ESG constrain corporate earnings management? Evidence from China. *Finance Research Letters*, 61, Article 104983.
- Sun, X., & Gunia, B. C. (2018). Economic resources and corporate social responsibility. *Journal of Corporate Finance*, 51, 332–351.
- Tang, N., Xu, X., Hsu, Y. T., & Lin, C. Y. (2024). The impact of ESG distance on mergers and acquisitions. *International Review of Financial Analysis*, 96, Article 103677.
- Vanstraelen, A., Zarzeski, M. T., & Robb, S. W. (2003). Corporate nonfinancial disclosure practices and financial analyst forecast ability across three European countries. *Journal of International Financial Management & Accounting*, 14(3), 249–278.
- Waddock, S. A., & Graves, S. B. (1997). The corporate social performance–financial performance link. *Strategic Management Journal*, 18(4), 303–319.
- Wang, L., Chen, C., & Zhu, B. (2023). Earnings pressure, external supervision, and corporate environmental protection investment: Comparison between heavy-polluting and non-heavy-polluting industries. *Journal of Cleaner Production*, 385, Article 135648.
- Wang, Z., & Xing, T. (2023). Does environmental, social and governance performance lessen analyst optimistic bias: Evidence from China. *Asia-Pacific Journal of Financial Studies*, 52(5), 793–818.
- Weber, O. (2014). Environmental, social and governance reporting in China. *Business Strategy and the Environment*, 23(5), 303–317.
- Wong, W. C., Batten, J. A., Mohamed-Arshad, S. B., Nordin, S., & Adzis, A. A. (2021). Does ESG certification add firm value? *Finance Research Letters*, 39, Article 101593.
- Wu, K. S., & Chang, B. G. (2022). The concave–convex effects of environmental, social and governance on high-tech firm value: Quantile regression approach. *Corporate Social Responsibility and Environmental Management*, 29(5), 1527–1545.
- Wu, W., Zhang, S., Fan, Y., & Shi, Y. (2024). Financial flexibility, firm performance, and financial distress: A comparative study of China and the US during pandemics. *International Review of Financial Analysis*, 96, 103706.
- Xu, J., Wu, W., & Feng, X. (2025). The impact of ESG performances on analyst report readability: Evidence from China. *International Review of Financial Analysis*, 102, Article 104056.
- Yu, E. P. Y., Guo, C. Q., & Luu, B. V. (2018). Environmental, social and governance transparency and firm value. *Business Strategy & the Environment*, 27(7), 987–1004.
- Yuan, X., Li, Z., Xu, J., & Shang, L. (2022). ESG disclosure and corporate financial irregularities—Evidence from Chinese listed firms. *Journal of Cleaner Production*, 332, Article 129992.
- Zhang, J., Li, Y., Xu, H., & Ding, Y. (2023). Can ESG ratings mitigate managerial myopia? Evidence from Chinese listed companies. *International Review of Financial Analysis*, 90, Article 102878.

- Zhang, Z., & You, J. (2024). Does firms' ESG information disclosure have contagion effect? Evidence from China. *Corporate Social Responsibility and Environmental Management*, 31(4), 3274–3296.
- Zhao, T., & Chan, K. C. (2023). Corporate social network and corporate social responsibility: A perspective of interlocking directorates. *International Review of Financial Analysis*, 88, Article 102711.

- Zhou, F., Zhu, J., Qi, Y., Yang, J., & An, Y. (2021). Multi-dimensional corporate social responsibilities and stock price crash risk: Evidence from China. *International Review of Financial Analysis*, 78, Article 101928.
- Zumente, I., & Läce, N. (2021). ESG rating-necessity for the investor or the company? *Sustainability*, 13(16), 8940.