

34 **Keywords:** organisational pride, psychological safety, workplace, deep learning,
35 automatic psychometric assessment, companies

36 1 Introduction

37 “*Psychological safety isn’t about being nice. It’s about giving candid feedback, openly*
38 *admitting mistakes, and learning from each other*” [1] — Amy Edmondson

39 Understanding employees’ work experiences offers valuable insights into an orga-
40 nization’s culture, contributing to employee engagement, job satisfaction, and overall
41 productivity [2]. A positive workplace environment can enhance innovation and cre-
42 ativity by encouraging employees to share ideas and take initiative [3]. To maintain
43 a supportive culture, many organizations proactively address workplace challenges
44 through open communication channels, which helps improve employee well-being and
45 reduce stress [4]. However, companies today face significant pressures from global dis-
46 ruptions such as the COVID-19 pandemic [5], geopolitical conflicts [6], and shifts in
47 the talent market [7], which impact workforce dynamics and raise the need for strong
48 organizational cultures. Consequently, a deep understanding of employees’ percep-
49 tions—from their attitudes towards company strategy to team-level culture—allows
50 organizations to strengthen their public reputation, align strategic goals, and foster
51 resilience in a rapidly evolving business landscape.

52 Companies commonly rely on psychological constructs to measure various aspects
53 of organisational culture. Organisational culture encompasses the shared values,
54 beliefs, and practices of a company’s employees [8], and it emerges from the interplay
55 of top-down expectations and bottom-up norms [9]. Given its multidimensional nature
56 [10], various aspects of organisational culture are captured, ranging from leadership,
57 resolving conflicts and negotiating, independence, and freedom to make decisions, to
58 training and teaching others [8]. Two widely discussed aspects of organisational cul-
59 ture are *organisational pride* (OP) and *psychological safety* (PS). Both OP and PS are
60 significant aspects of organisational culture that shape employee values, beliefs, and
61 practices [11–14]. Employees’ pride in their workplace (OP) often indicates a positive
62 and healthy organisational culture [11, 12, 15, 16]. OP reflects employees’ emotional
63 and attitudinal attachment to their organisation, influencing employee behaviour, cre-
64 ativity, turnover, and commitment. It serves as both an antecedent and a consequence
65 of corporate success [11, 15, 17]. On the other hand, PS pertains to employees’ shared
66 perception that it is safe to take interpersonal risks, express ideas, and voice opinions
67 without fear of negative consequences or judgement. It enables a supportive work envi-
68 ronment where employees feel comfortable being themselves and contributing their
69 perspectives [1, 13]. OP is acknowledged as a key component in shaping organisational
70 culture, yet its impact has not been extensively quantified on a large scale. Our focus
71 on OP stems from substantial qualitative evidence suggesting its significant role in
72 company culture. However, OP alone does not fully capture the grassroots aspects of
73 organisational culture. Indeed, organisational culture is not a mere aggregation of ele-
74 ments; it evolves from the intricate interaction of various factors [8], including both

75 OP and PS. PS plays a crucial role in building organisational resilience and fostering a
76 culture where employees can navigate uncertainty and challenges effectively [1, 14, 18–
77 20]. Hence, we integrate PS with OP to comprehensively understand organisational
78 culture’s emergence from this dynamic and multifaceted interaction [12, 15, 21, 22].
79 Traditionally, these constructs have been assessed through surveys, which can be
80 costly [23] and limited to self-selected participants, posing challenges to generalisabil-
81 ity [24–26]. In this work, we developed automated methods for assessing aspects of
82 organisational culture, and, in so doing, we made three contributions:

- 83 1. We implemented a deep-learning Natural Language Processing (NLP) framework
84 that accurately captures OP and PS at company-level (§4).
- 85 2. We applied this framework to over 430,000 publicly available Glassdoor employee
86 reviews, and validated it internally in terms of keywords expressed in high OP/PS
87 companies, and externally in terms of associations of company OP/PS scores with
88 the ratings. We also surfaced variations in OP/PS levels across different indus-
89 try sectors from 318 of S&P 500 companies. Lastly, we uncovered that despite
90 being correlated aspects of organizational culture, OP and PS each independently
91 contribute to predicting employee satisfaction (§5).

92 Our automatic operationalisation of these two pivotal organisational constructs
93 suggests avenues for the development of *automated psychometric assessment*, allow-
94 ing for analysing organisational culture across sectors and countries, in addition to
95 exploring the interactions between various organisational constructs. Ultimately, the
96 results generated from this process could contribute to the creation of more satisfying
97 and productive work environments (§6).

98 2 Background and Related Work

99 First, we provide the reader with a background on psychological constructs commonly
100 used to assess organisational culture, as discussed in the Organisational Behaviour
101 Literature, with a focus on our two constructs of interest: *organisational pride* and
102 *psychological safety*. Following that, we place our work within the broader context of
103 existing literature on Natural Language Processing (NLP) research in Computational
104 Social Science, which attempted at capturing aspects of organisational culture on a
105 large scale.

106 2.1 Organisational Behaviour Literature

107 Organisational Culture (OC) refers to the shared values, beliefs, assumptions, norms,
108 and behaviours that collectively shape the identity and character of an organi-
109 sation [27]. It significantly impacts how employees align their actions with the
110 organisation’s goals and expectations, ultimately influencing work attitudes and
111 overall performance [28].

112 2.1.1 Organisational Pride (OP)

113 It refers to the emotional and attitudinal attachment that employees develop towards
114 their organisation [15]. It encompasses both the emotional attachment triggered by

115 perceiving a successful organisational event and the lasting cognitive attitude resulting
116 from a general perception of the organisation [11, 29, 30]. As a psychological construct,
117 OP serves as an indicator of both internal and external organisational reputation and
118 can act as both a driver and an outcome of corporate success [11, 15, 17, 31].

119 Previous research has identified various factors that contribute to the develop-
120 ment of employees' OP. Companies that provide job autonomy, job variety, rewards,
121 recognition, and demonstrate success, social responsibility, and internal sustainabil-
122 ity are more likely to evoke a sense of pride among their employees [17, 29]. In turn,
123 OP influences employee behaviour in several positive ways. It leads to favourable
124 job perceptions, customer-oriented behaviour, enhanced creativity, reduced turnover,
125 increased commitment, and pro-environmental behaviours [11, 12, 31–36]. Further-
126 more, OP is positively correlated with organisational attractiveness and job pursuit
127 intentions among prospective employees [12, 32–34]. It has also been found to medi-
128 ate the relationship between corporate social responsibility (CSR) and organisational
129 citizenship behaviour (OCB) [37, 38].

130 The measurement and monitoring of OP are crucial for organisations due to its
131 impact on company reputation, employee outcomes, and turnover [33, 39, 40]. How-
132 ever, there is still a need for a comprehensive examination of the conceptual scope
133 and structural properties of OP. Previous empirical studies have often relied on
134 small survey samples from a limited number of organisation's, raising concerns about
135 unreliability [33, 34, 39–42]. While these studies have explored various antecedents
136 and consequences of OP, such as length of affiliation, work-related emotions, and
137 corporate social responsibility, the limited availability of data has hindered the exam-
138 ination of these consequences at higher-order levels of analysis [12, 33, 34, 40, 41, 43].
139 Consequently, potential macro-level influences and outcomes across industries and
140 geographical regions have remained unexplored [33, 34, 44, 45].

141 2.1.2 Psychological Safety (PS)

142 It encompasses the shared belief among employees that they can take interper-
143 sonal risks within the workplace [1, 13, 19, 20, 46]. This includes freely expressing
144 ideas, voicing opinions, and being authentic, without the fear of facing negative con-
145 sequences or judgement, encouraging employee engagement and the integration of
146 diverse perspectives [13].

147 Traditionally, psychological safety has primarily been assessed at the team level
148 using measurement items developed by Edmondson [1]. While a few studies have exam-
149 ined the measurement of psychological safety at the organisational level [14, 47, 48],
150 they typically modified the team-level items by substituting “team” with “organiza-
151 tion” [49]. However, reaching a consensus on psychological safety can be challenging
152 in larger organisations, where employees may have diverse experiences within their
153 respective teams [50, 51]—leadership style and team dynamics shape the perception
154 of psychological safety [49]. For this reason, previous research indicates that psy-
155 chological safety has a more substantial impact at the team level compared to the
156 organisational level. Additionally, it has been suggested that smaller organisations
157 with frequent collaboration and strong corporate cultures are more likely to foster
158 a climate of psychological safety, although further empirical evidence is needed [49].

159 Furthermore, studies have explored potential adverse outcomes associated with exces-
160 sively high levels of psychological safety within teams, such as unethical behaviours
161 and reduced organisational performance [14, 46].

162 Researchers have traditionally relied on qualitative interviews to gain insights into
163 psychological safety at both the team and individual levels. However, in order to evalu-
164 ate levels of psychological safety within teams or organisations, the use of observational
165 methods has been proposed as an alternative approach [49]. Despite its potential to
166 provide valuable insights into the dynamics of psychological safety, including macro-
167 level influences and outcomes across various industries and geographical regions, this
168 complementary approach, similar to OP, has been scarce.

169 2.2 NLP Research in Computational Social Science

170 Das Swain et al. [8] introduced the use of natural language processing techniques on
171 Glassdoor review data to automatically evaluate OC using psychological constructs
172 similar to those in our study, such as the Organisation Cultural Inventory [52], the
173 Organisation Culture Profile [53], Hofstede’s Organisation Culture Questionnaire [54],
174 and the Organisation Culture Survey [55]. They compared word-level vector repre-
175 sentations of descriptions of organisational constructs with embeddings of Glassdoor
176 reviews and successfully demonstrated the face and construct validity of their language
177 model-based assessment of organisational culture. Additionally, they provided empiri-
178 cal evidence, based on Fortune 500 companies, of the connection between OC and
179 job performance. Similarly, Sen et al. [56] and Ceccarini et al. [57] utilised Glassdoor
180 reviews to examine Internal Sustainability Efforts (ISEs) at the organisational level
181 within S&P 500 companies. They confirmed the positive associations between ISEs
182 and stock growth. Furthermore, Lee [58] quantified internal corporate social responsi-
183 bility (CSR) and discovered a positive impact on Organisational Citizenship Behaviour
184 (OCB), which encompasses voluntary actions by employees that go beyond their formal
185 job requirements. In line with these findings, Šćepanović et al. [59] analysed thousands
186 of Glassdoor reviews and uncovered that different types of stress contribute differ-
187 ently to companies’ success. Specifically, companies whose employees tend to perceive
188 stress positively (i.e., describing it in terms of a high-growth and collaborative work
189 environment) enjoy higher stock growth compared to those whose employees perceive
190 stress negatively.

191 At the time of writing, there is a noticeable gap in comprehensive empirical research
192 that explores OP and PS, two constructs that hold significant importance in under-
193 standing employee experiences in the workplace. What sets them apart from other
194 constructs like OC, CSR, and ISEs is their assessment using established psychologi-
195 cal measures [1, 41]. Unlike these constructs, which have been measured on a large
196 scale but with adapted definitions and descriptions, OP and PS metrics can rely on
197 validated measures. Hence, they presents valuable opportunities to develop empirical
198 methodologies that quantify organisational constructs based on validated psychologi-
199 cal measures, while encompassing comprehensive aspects of the workplace, from the
200 global organisation level to the team level, using online data.

201 3 Datasets

202 3.1 Description and Statistics

203 We collected data from Glassdoor, a well-known platform for company reviews,
204 where individuals, including current and former employees, share their reviews and
205 experiences regarding various aspects of their corporate life. The platform ensures
206 high-quality reviews through three key measures. First, it employs both proprietary
207 automated tools and manual moderation to oversee content, while also allowing users
208 to flag inappropriate posts. This dual approach helps reduce the risk of fake reviews,
209 such as those submitted under coercion from employers to post positive feedback.
210 Second, users are required to contribute at least one review before gaining access to
211 others' content. This policy fosters a more balanced range of opinions and helps miti-
212 gate *non-response bias*, where the views of respondents might significantly differ from
213 those who choose not to participate. Lastly, the platform limits each employee to a
214 single review per company per year. This cap prevents individuals from dominating
215 the review landscape, addressing *sampling bias*, which occurs when certain perspec-
216 tives are overrepresented in the data. As discussed in the related work section, this
217 data collection has been previously employed in studies examining other organisa-
218 tional constructs on a large scale [8, 56, 59]. However, we acknowledge that Glassdoor
219 reviews, like any source of self-reported data, are not free from bias and should be
220 seen as a complementary source rather than a replacement for survey data.

221 In addition to being efficient and low-cost, a potentially unique advantage of using
222 Glassdoor reviews is their relative independence from organizational-level psychologi-
223 cal safety. In traditional surveys conducted within organizations, employees may feel
224 pressured to inflate their assessments of organizational culture or job satisfaction due
225 to fear of repercussions, especially in environments where psychological safety is low.
226 This bias may be less prevalent in anonymous Glassdoor reviews, offering a distinct
227 perspective on employee experiences.

228 The dataset used in this study consisted of Glassdoor reviews spanning a twelve-
229 year period from 2008 to 2020. Each review in the dataset contained various
230 components, including a title, a 'pro' section highlighting positive aspects of the com-
231 pany, a 'con' section discussing negative aspects, four ratings (ranging from 0 to 5)
232 evaluating the company's *career opportunities*, *work-life balance*, *culture*, and *man-*
233 *agement*, as well as an *overall rating* of the company. Some reviewers also provided
234 their location, enabling us to identify the states for a subset of the reviews. To ensure
235 the reliability and robustness of our text processing method, we applied specific crite-
236 ria for dataset selection. We retained companies that had a minimum of 100 reviews,
237 resulting in a final dataset of 430,788 reviews from 318 US-based companies, which
238 represents 99.1% of the original dataset. Notably, 80% of these companies were listed
239 in the S&P 500 index. To assess the representativeness of these 318 companies, we
240 examined the distribution of industry sectors, which was found to be comparable to
241 that of the S&P 500 companies. Additionally, we computed the correlation between
242 the number of reviews per state and the total number of employment positions per
243 state, and found that they are highly correlated ($r = 0.97$, $p < 0.001$) as shown in
244 Figure 7 in the Appendix.

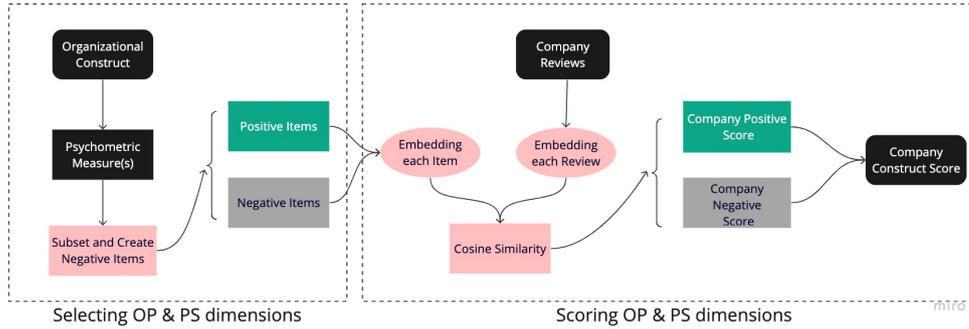


Fig. 1: The framework for measuring an organisational construct score (e.g., OP or PS score) from company reviews. Starting from the validated psychometric measures for the construct of interest, we subset negative from positive items, and, if no negative items were present, we first create them as opposites to the positive items. We then embed each item and measure its similarity to the embedding of the company reviews. For each company, the proportion of reviews found similar to the positive statements, gives company positive score, and the proportion of reviews found similar to the negative statements, gives company negative score. The final company score for the given construct is calculated as the difference between positive and negative scores.

245 4 Framework for Automatically Assessing OP and 246 PS

247 We employed a mixed-method approach to develop and validate an automated,
248 language-based assessment of OP and PS, similar to Sen et al. [56]’s framework for
249 measuring ISEs (Figure 1). We identified key language features related to OP and PS
250 by reviewing the organisational behaviour and psychometrics literature for existing
251 assessments. We separated the positive and negative items and created opposing nega-
252 tive items when they were not initially available. We then employed the sentence-level
253 BERT (SBERT) model [60] to generate embeddings for each item and subsequently
254 calculated their similarity with the embeddings of company reviews. Specifically, we
255 used `sentence-t5-xl` for its effectiveness in similarity comparisons while still being
256 computationally efficient [61]. It has been trained on the Massive Text Embedding
257 Benchmark, an evaluation benchmark across 58 datasets and 112 languages for embed-
258 ding for tasks such as classification, clustering, and semantic textual similarity [62].
259 Adhering to the recommendations by Reimers and Gurevych [60], we used SBERT
260 with its default settings and specifically set the maximum sequence length to 256 and
261 employed token-based pooling.

262 SBERT is adept at capturing and processing the semantics of noisy reviews. This
263 deep learning-based model is designed to interpret and process text that contains
264 irregularities such as typos and grammatically inconsistent language by using subword
265 embeddings [63, 64]. For example, the word “learning” could be divided into sub-
266 words like “learn” and “ing”. The model then learns vector representations for each

267 subword. This allows the model to infer meanings of words it has not seen before by
268 understanding their component subwords. Hence, unlike traditional natural language
269 processing techniques, SBERT allows for bypassing conventional preprocessing steps
270 like stemming or lemmatization.

271 To assess the company’s positive score, we determined the proportion of reviews
272 aligning with positive statements. Similarly, the negative score was determined by
273 examining the proportion of reviews aligning with negative statements. Finally, we
274 obtained the overall company score for the specific construct by taking the difference
275 between the positive and negative scores.

276 4.1 Step 1: Selecting the OP and PS dimensions

277 4.1.1 Curating Positive OP and PS Items

278 To begin with, we resorted to validated psychological scales to construct items for
279 OP and PS. Next, we explain the curation process.

280 **Organisational Pride.** In Psychology, three widely used measures for organisational
281 pride are developed by Jones [41], Turban et al. [12], and Cable and Turban [32].
282 These three measures have demonstrated high reliability (i.e., with reliability coeffi-
283 cients of at least .80). Considering that Glassdoor reviews were mostly expressed in
284 the past tense, we deliberately omitted items that included future tense language. As
285 a result, we chose four specific items from the measure developed by Jones [41] (OP1
286 to OP4 in Table 1).

287
288 **Psychological Safety.** To measure PS at both the company and team levels, we
289 utilised the items from Edmondson [1], which is the most commonly used measure in
290 the literature on organisational behaviour and psychometrics [49]. Note that we needed
291 to paraphrase one item from “*No one on this team would deliberately act in a way that*
292 *undermines my efforts.*” to “*Everyone on this team would act in a supportive manner*
293 *towards my efforts.*” to avoid a double negative, which would otherwise introduce
294 more noise when extracting embeddings from the deep learning model (PS1 to PS4 in
295 Table 1).

297 4.1.2 Creating Negative OP and PS Items

298 Glassdoor requires users to separate their reviews into pros and cons sections. This
299 helps to ensure a fair evaluation that covers both the positive and negative aspects
300 of a company, allowing users to make informed decisions about job opportunities and
301 company cultures. Therefore, this separation of pros and cons allowed us to align the
302 items with their corresponding sentiments in the reviews, leading to a more precise
303 evaluation of similarity. The original OP measure includes only positive items to assess
304 OP. We manually developed additional items reflecting opposite valence to those in
305 the original OP measure. These items assess ‘lack of organisational pride’ (LOP),
306 showing absence of positive attachment and identification with the organisation (LOP1
307 to LOP4 in Table 1). Since the original measure includes both positive and negative
308 items, for PS, we separated existing negative items from positive ones to gauge ‘lack of

Table 1: Final selected or adapted items from psychometric measures for measuring OP and PS.

Abbr.		Abbr.	
Organisational Pride (OP) [41]		Psychological Safety (PS) [1]	
OP1	I am proud to work for this firm.	PS1	Members of this team are able to bring up problems and tough issues.
OP2	People respect what this firm does.	PS2	It is safe to take a risk on this team.
OP3	I am proud to be associated with this firm.	PS3	Everyone on this team would act in a supportive manner towards my efforts.
OP4	I am proud of what this firm accomplishes.	PS4	Working with members of this team, my unique skills and talents are valued and utilized.
Lack of OP (LOP) adapted from [41]		Lack of PS (LPS) [1]	
LOP1	I am displeased to work for this firm.	LPS1	If you make a mistake on this team, it is often held against you.
LOP2	People look down upon what this firm does.	LPS2	People on this team sometimes reject others for being different.
LOP3	I am displeased to be associated with this firm.	LPS3	It is difficult to ask other members of this team for help.
LOP4	I am dissatisfied with how this firm accomplishes what it does.		

309 psychological safety’ (LPS), which involves shared perception of fear about potential
 310 negative consequences for speaking up (LPS1 to LPS3 in Table 1). Note that because
 311 the original PS scale has seven items total, we ended up with four positive and three
 312 negative items.

313 4.1.3 Final OP and PS Items

314 Combining the previous two steps, we were left with a total of 15 items: 8 for measuring
 315 OP construct, and 7 for measuring PS construct (Table 1).

316 4.2 Step 2: Scoring the OP and PS dimensions

317 4.2.1 Calculating Separate Item Scores

318 We examined the 15 items at the company-level (denoted as u) to quantify organi-
 319 sational behaviour across multiple companies. To achieve this, we defined the score
 320 $s(u, i)$ of the i^{th} item for company u by calculating the fraction of reviews mentioning
 321 i within company u .

$$s(u, i) = \frac{\sum_{p \in R(u)} sim_t(v_p, v_i)}{|R(u)|} \quad (1)$$

322 Here, $R(u)$ represents the collection of reviews for company u , v_i denotes the SBERT
 323 (Sentence-BERT) vector corresponding to item i , and $sim_t(v_p, v_i)$ refers to the *thresh-*
 324 *olded* SBERT similarity score [60] between the SBERT vector of review p and the
 325 SBERT vector associated with item i . We applied this similarity thresholding tech-
 326 nique to account for noise and potential variations in review expressions. For instances
 327 of missing data, specifically reviews without accompanying text or which are too short,
 328 this similarity thresholding effectively filtered out such instances since they yielded
 329 low similarity scores and were thus excluded from our analysis. Items with similarity
 330 scores below a certain threshold are considered dissimilar and assigned a score of 0,
 331 while those above the threshold retain their similarity scores. To be more specific, the

332 definition of $sim_t(v_p, v_i)$ is as follows:

$$sim_t(v_p, v_i) = \begin{cases} sim(v_p, v_i), & \text{if } sim(v_p, v_i) > \max(\theta, p_{50}(i)) \\ 0, & \text{otherwise} \end{cases} \quad (2)$$

333 We chose the threshold as the maximum value between θ and $p_{50}(i)$, where θ is the
 334 average value of the median similarities (one for pros and one for cons), and $p_{50}(i)$
 335 is the median similarities for item i .

336 Additionally, it is important to highlight that we empirically identified a connec-
 337 tion between OP/PS and the pros, as well as between LOP/LPS and the cons, using
 338 similarity scores. Pairing the cons with OP and PS (or pros with LOP and LPS)
 339 was found to negatively affect the performance of our framework during the valida-
 340 tion phase of our mixed-method approach. Specifically, when cons were paired with
 341 OP/PS, negative statements about the company or teams were sometimes misinter-
 342 preted as similar to our items, thereby contributing to the construct scores despite
 343 their opposite meanings. This issue arises because the similarity measure struggles
 344 to handle negations effectively. To mitigate such inconsistencies, we naturally aligned
 345 positive construct items with pros and negative ones with cons.

346 4.2.2 Combining Item Scores

347 To derive composite scores for the constructs, namely OP and PS, we aggregated the
 348 scores of individual items. This aggregation enabled us to create more comprehen-
 349 sive representations of these underlying constructs. For OP, we calculated \bar{OP}_u
 350 for company u by averaging the scores of four items, as shown in Equation 3:

$$\bar{OP}_u = \frac{s(u, OP1) + s(u, OP2) + s(u, OP3) + s(u, OP4)}{4} \quad (3)$$

351 The same approach was then applied to calculate composite scores for PS, LOP, and
 352 LPS. Subsequently, we standardised the values for each category by subtracting the
 353 mean and dividing it by the standard deviation, yielding the z-score. This process is
 354 illustrated in the following equation.

$$z_{OP,u} = \frac{\bar{OP}_u - \mu(\bar{OP})}{\sigma(\bar{OP})} \quad (4)$$

$$z_{PS,u} = \frac{\bar{PS}_u - \mu(\bar{PS})}{\sigma(\bar{PS})} \quad (5)$$

$$z_{LOP,u} = \frac{L\bar{OP}_u - \mu(L\bar{OP})}{\sigma(L\bar{OP})} \quad (6)$$

$$z_{LPS,u} = \frac{L\bar{PS}_u - \mu(L\bar{PS})}{\sigma(L\bar{PS})} \quad (7)$$

355 Finally, the OP and PS scores for company u were obtained by subtracting the
356 z-score of the pro construct category from the con construct category.

$$OP_{\text{score},u} = z_{OP,u} - z_{LOP,u} \quad (8)$$

$$PS_{\text{score},u} = z_{PS,u} - z_{LPS,u} \quad (9)$$

357 4.3 Step 3: Validating the scoring of OP and PS

358 We focused on OP and PS since neither was previously studied and captured on a
359 large scale. That means that there are no ground truth labels nor previous work to
360 compare to. Hence, we assessed the effectiveness of our proposed method in two steps
361 as follows. First, we implemented a baseline approach using extracted keywords from
362 the OP and PS statements for the two construct measures. We automatically identi-
363 fied keywords and phrases from each construct item using the Rapid Automatic
364 Keyword Extraction (RAKE) algorithm. This domain-independent method identifies
365 key phrases—such as single keywords, bigrams and n-grams—by analyzing word fre-
366 quency and co-occurrence patterns within the text as outlined in [65]. We then labelled
367 a review as belonging to a certain construct, if we could find any of the phrases asso-
368 ciated with that construct in the review, and scored the strength of the association
369 based on the number of such phrases identified.

370 We randomly sampled 400 reviews: 200 from pros and 200 from cons. For pros, for
371 each of the two positive constructs (OP and PS), we took 50 reviews with composite
372 scores (Eq. 3) above the 75th percentile (i.e., corresponding to a high construct value),
373 and 100 with a composite score of 0 (i.e., not pertaining to any of the two constructs).
374 Similarly, we sampled 200 reviews from cons. This stratified sampling ensured that we
375 covered all the cases, i.e., reviews that were high or low in either of the constructs, and
376 their combinations, or that did not express either of them. Three independent anno-
377 tators then manually assessed the relevance of these reviews. Finally, we compared
378 our proposed framework against the keyword-based method, as a lower bound base-
379 line, on the set of manually labelled reviews. Additionally, we also benchmarked our
380 method against predictions from GPT-4o-mini [66], a state-of-the art Large Language
381 Model [67], as a more competitive baseline.

382 4.4 Step 4: Analysing OP and PS

383 To understand what characterises OP and PS, we adopted a linguistic approach, pre-
384 viously explored by Das Swain et al. [8], through which we identified the most frequent
385 keywords, including 1, 2, 3, and 4-grams, from the reviews that our method deemed
386 relevant for each construct. Since a review may pertain to multiple constructs, we
387 divided the reviews into sentences. Next, we calculated the TF-IDF scores for the identi-
388 fied n-grams, with each document consisting of all shortlisted sentences for a specific
389 construct. Finally, we ranked the keywords based on their TF-IDF scores, providing
390 insights into the importance of certain terms for each construct as determined by our
391 embedding-based method. This process was performed separately on the pro reviews
392 paired with OP/PS and on the con reviews paired with LOP/LPS (as detailed in
393 Section 4.1).

394 As previous literature suggests a positive correlation between both OP and PS and
395 employee satisfaction [11, 13, 15, 17, 31, 49], we investigated the relationship between
396 OP/PS scores and the company ratings through a linear regression analysis, while
397 controlling for the number of reviews.

398 As we found that the distribution of OP and PS scores varies across different indus-
399 try sectors (Figure 4), we examined whether these two constructs relate differently to
400 industry sectors. To determine whether the differences in the two scores across sec-
401 tors are statistically significant, we conducted ANOVA followed by post-hoc analyses
402 using Tukey’s (honestly significant difference) HSD tests [68]. Note that we only anal-
403 ysed sectors with more than 10 companies to ensure the analysis possesses sufficient
404 statistical power. A larger sample size within each sector allows for more reliable and
405 generalisable findings, reduces the impact of outliers, and enables more nuanced sub-
406 group analyses. We also conducted a qualitative analysis using the publicly available
407 reviews. To identify reviews that encapsulate the predominant themes within each sec-
408 tor, we followed two steps that are commonly adopted in thematic analysis [69]. We
409 first applied open coding to identify key concepts that emerged across multiple reviews;
410 specifically, two of the authors read all the reviews that scored in the top 5% from
411 OP/PS, and marked them with keywords that reflected the key concepts expressed
412 in the text. They then used axial coding to identify relationships between the most
413 frequent keywords, and select representative reviews. We summarise our qualitative
414 analysis with respect to OP and PS. Any excerpts utilised to support the analysis
415 were paraphrased to ensure anonymity and avoid the identification of specific reviews.

416 5 Results

417 We applied the framework (§4) on 430K employee reviews from 2008 to 2020 in 318
418 major U.S. companies, and showed that these two constructs are associated differently
419 with organisational culture. Table 4 (Appendix) shows the summary statistics of all
420 variables used in our analysis, and Table 7 (Appendix) shows the outcomes of our
421 manual evaluation and the comparative analysis of our method versus the keyword-
422 and LLM-based baselines. This table highlights the high accuracy of our approach and
423 its clear advantage over the baseline method. Moreover, our method also outperformed
424 or matched the LLM baseline, highlighting its state-of-the art performance for this
425 task. In this section, we present our findings about the two constructs examined across
426 companies and industries.

427 **OP applies to a company as a whole, and it is high in utility and energy sec-**
428 **tors and low in consumers and communications.** When analysing the positive
429 and negative construct items, which represent the pros and cons of review sections,
430 we observed a consistent language pattern. Initially, we discovered that all items asso-
431 ciated with both OP and PS shared common keywords such as *work*, *people*, and
432 *employee* (Figure 2). This outcome was expected since both aspects revolve around
433 the employee experience in the workplace. However, OP-related items tended to focus
434 more on the overall “company” aspect, indicated by words like *company*, *job*, and *bene-*
435 *fit*. This finding aligns with the definitions of these concepts in organisational research,
436 where it primarily encompasses employee experiences related to the organisation as a

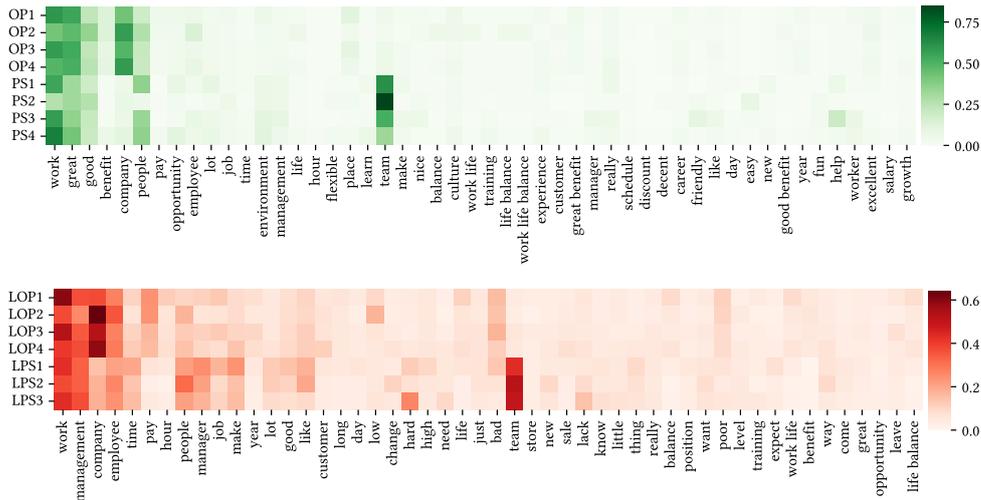


Fig. 2: Top 50 n-grams in sentences expressing OP & PS. Darker colours (higher normalised TF-IDF score) indicate greater relative relevance to a particular construct. The distinctive keywords for OP and PS are “company” and “team”, respectively. The n-grams are sorted by their frequency in all the documents.

437 whole [15]. By exploring more specific keywords without the six most frequent key-
 438 words (*company, good, great, people, team,* and *work*) as shown in Figure 3, we found
 439 that OP was now associated with (*good/great*) *benefits, culture, employees,* and (*good/-*
 440 *great*) *place*). These findings are consistent with previous research indicating a positive
 441 relationship between OP and work-life balance [17], as well as the benefits of internal
 442 CSR practices for employees [70]. This suggests that our OP measure applies to
 443 a company as a whole. Examples of reviews with high and low OP/PS scores can be
 444 found in Table 2.

Table 2: Example reviews for high and low OP/PS scores.

#	High OP	#	Low OP
1	Company that you are proud to work for.	1	One of the most unethical and corrupt companies I have ever worked for.
2	This company has some of the best and brightest talent and I am happy to work alongside of them.	2	Felt disjointed from the firm.
3	The company has an excellent reputation and I am proud to work here.	3	Constant lawsuits and bad press make one not want to admit to working there.
4	I love working for a company with such brand recognition.	4	One of the worst companies I have ever experienced.
#	High PS	#	Low PS
1	Excellent benefits, team members that are willing to assist with issues and questions.	1	If you make an error, they hold it against you.
2	Bunch of smart people working on difficult problems.	2	There's much confusion and miscommunication among teams.
3	Great team facing challenging environment. Strong commitment by all coworkers.	3	Lot of politics. If you end up being with a wrong team, your experience will be really bad.
4	Terrific people, strong team emphasis.	4	It is a team environment. Some team members are not as good as others.

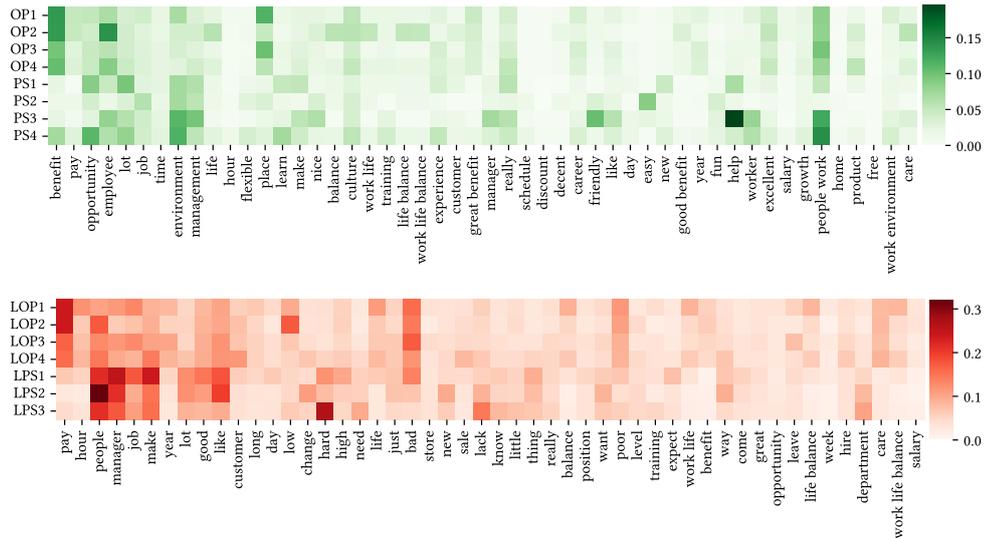


Fig. 3: Top 50 n-grams in sentences expressing OP & PS. Darker colours (higher normalised TF-IDF score) indicate greater relative relevance to a particular construct. We removed top shared keywords (*company, good, great, people, team, work* for pro reviews and *work, management, company, employee, time, team* for con reviews) and to let more construct-specific keywords emerge. The n-grams are sorted by their frequency in all the documents.

445 *OP across industry sectors.* ANOVA indicated statistically significant differences in
 446 OP scores across sectors ($F = 4.11, p < 0.001$). Post-hoc analysis with Tukey’s HSD
 447 identified significant mean differences ($p < 0.05$) between consumer discretionary and
 448 four other sectors: energy, health care, information technology, and utilities.

449 Our analysis showed above-average scores in the *utilities* and *energy* sectors (Figure
 450 4). The high OP in these sectors may be attributed to employees engaged in renew-
 451 able energy sources, as they take pride in contributing to clean energy solutions and
 452 addressing pressing environmental issues. Example reviews: ‘*Leading the way to a
 453 greener future. Leading the renewable and energy storage world transformation. Keep-
 454 ing the lights on means 24x7 dedication to customers*’ and ‘*I can sense an exciting
 455 future that will reshape the company completely. The people here are great.*’ and ‘*There
 456 is no slack anywhere in company X, which is a good thing for shareholders but means
 457 it can at times be quite exhausting working at company X.*’

458 The *consumer staples* sector, which includes companies producing and selling
 459 essential consumer goods, exhibited substantial variability in OP scores. This can be
 460 attributed to the diverse nature of companies within this category. For example, com-
 461 panies offering food, beverages, or personal care products (e.g., Costco, Coca-Cola
 462 Company, and Procter & Gamble Co.) may experience different levels of pride (e.g.,
 463 ‘*The CEO and CFO and in company X support their workers, and they believe in com-
 464 pensating and growing the people from within. They really care about their employees.*’)

465 compared to those involved in the production and sale of alcohol or tobacco products
466 (e.g., Philip Morris International Inc.). The latter group, particularly tobacco com-
467 panies, faces growing external critique and societal pressures due to the prioritisation
468 of healthier lifestyles, which can influence employee sentiments towards the organi-
469 sations. Example reviews: *‘New purpose to deliver smoke-free products jumpstarted a
470 need to address the culture and explore new ways of working. [...] Cost pressure due
471 to lower sales volumes results in budget cuts.’* and *‘[...] they did not mention that
472 the role involved standing in a retail shop the entire day, approaching smokers and
473 enrolling them to try smoke-free products. The management did not mention that the
474 job involved “educating retailers” which in reality means bribing them to sell some-free
475 in exchange for money vouchers, ipads and airpads.’*

476 Lastly, both the *consumer discretionary* and *communication services* sectors
477 demonstrated significantly lower levels of OP. The *consumer discretionary* sector
478 comprises companies that produce non-essential or discretionary goods and services,
479 including industries with high turnover rates such as retail and e-commerce platforms.
480 Example reviews: *‘Very repetitive after a while!’* and *‘Like most retail businesses the
481 pay is substandard. Many employees have second jobs to make ends meet even if they
482 are working full-time here.’* and *‘Very stressful, lots of rude and antsy customers espe-
483 cially contractors.’* and *‘Staff keeps getting reduced, therefore not enough time to serve
484 everyone well. Then “they” wonder why sales are down.’* and *‘Some poor policies that
485 leave you dealing with some angry customers.’* Moreover, ethical concerns surround-
486 ing sustainability, animal welfare, and labour practices, particularly within industries
487 such as apparel and luxury goods, contribute to diminished feelings of pride among
488 employees. Example reviews *‘Minimal pay, rude customers, long holiday hours.’*, and
489 *‘Management seems clueless outside of their department.’* and *‘Retail is boring and
490 monotonous’* and *‘long hours over Christmas - have to work weekends’* and *‘People
491 no longer matter. Values and principles the company once was built on are no longer
492 there.’* and *‘Hours are not as organised as they could be, targets sometimes unrealistic
493 - management need to understand many customers will shop online as websites offer
494 better discounts.’*

495 In the *communication services* sector, especially social media and telecommu-
496 nication companies, negative public sentiment related to issues like data privacy,
497 misinformation, or service problems can impact employees’ pride in their organisation.
498 This sector often grapples with ethical issues such as data privacy, content regulation,
499 and the digital divide. If employees feel that their company is not effectively addressing
500 these issues, it can reduce their sense of pride in the organisation. Example reviews:
501 *‘Low pay and questionable privacy practices.’* and *‘You face angry customers, [...] The
502 trainers were not helpful and you feel like you’ve been left to handle things on your
503 own.’* Additionally, in large communication service companies, employees may per-
504 ceive themselves as small contributors in a large system, leading to a diminished sense
505 of pride in their work’s impact on the company and its customers. Example reviews:
506 *‘Whilst trying to keep the best person in each position, the company creates too much
507 turnover.’*, *‘We faced frequent layoffs when projects would complete [...]’*.

508 **PS applies to the team as a unit, and it is high in IT and low in communi-**
509 **cations and healthcare.** PS-related items were centred around the “team” aspect,

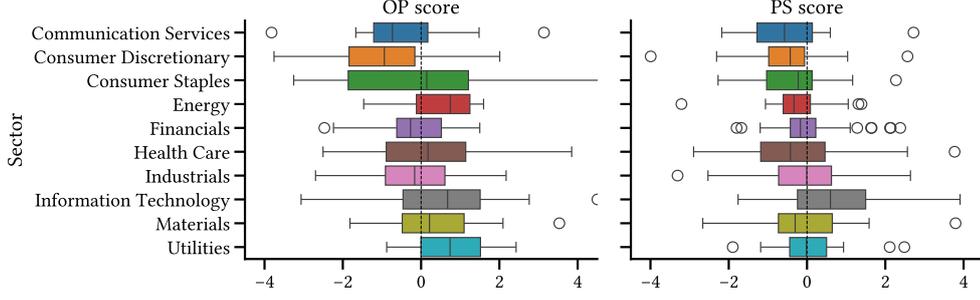


Fig. 4: OP and PS scores across industry sector. Box plots showing the distribution of OP and PS scores across different industry sectors.

Table 3: Predicting company ratings using OP and PS scores. Standard errors are shown below the coefficients.

	Career	Balance	Culture	Management	Overall
Intercept	2.650*** (0.058)	3.127*** (0.090)	2.731*** (0.065)	2.543*** (0.056)	2.914*** (0.047)
OP_{score}	0.201*** (0.010)	0.156*** (0.016)	0.228*** (0.011)	0.204*** (0.010)	0.227*** (0.008)
PS_{score}	0.025** (0.011)	0.052*** (0.018)	0.058*** (0.013)	0.056*** (0.011)	0.022** (0.009)
$OP_{score} \times PS_{score}$	0.010*** (0.004)	-0.010* (0.006)	-0.002 (0.004)	0.013*** (0.004)	0.001 (0.003)
$\log(n_{reviews})$	0.085*** (0.009)	0.028** (0.014)	0.089*** (0.010)	0.059*** (0.009)	0.069*** (0.007)
R^2	0.749	0.470	0.773	0.797	0.842
Adjusted R^2	0.745	0.463	0.770	0.794	0.840
AIC	-167.621	112.588	-95.732	-185.739	-301.657
BIC	-148.811	131.398	-76.922	-166.929	-282.847
Residual Std. Error	0.184	0.287	0.207	0.179	0.149
F Statistic	233.107***	69.263***	266.298***	306.434***	416.059***
Note:					*p<0.1; **p<0.05; ***p<0.01

510 evident from words like *team*, *people*, *help*, and *friendly*, as depicted in Figure 2. This
511 finding aligns with previous literature, suggesting that PS mainly pertains to employee
512 experiences within their teams [49]. By exploring more specific words associated with
513 PS (Figure 3), we found that it exhibited stronger associations with keywords such as
514 *environment*, *friendly*, *help*, *training*, and *management*. These findings align with pre-
515 vious studies that highlight the interaction between PS and leadership qualities and
516 team characteristics [49, 50]. Overall, this shows that our PS measure applies to the
517 team as a unit.

518 *PS across industry sectors.* ANOVA also confirmed statistically significant differences
519 in PS scores across sectors ($F = 2.75, p < 0.01$). Post-hoc analysis with Tukey’s HSD
520 identified significant mean differences ($p < 0.05$) between information technology and
521 three other sectors: consumer discretionary, consumer staples, and health care.

522 We found high levels of PS in the *information technology* sector (Figure 4). This
523 can be attributed to the career growth opportunities provided and the supportive work

524 environment within the *information technology* sector companies. Example reviews:
525 ‘[...] it was all about what you could bring to the table and they have an environment
526 that made you want to bring all that you had.’ and ‘Great career growth opportunities,
527 excellent horizontal industry mobility, and chance to truly own your career trajectory.’
528 and ‘Amazing benefits, lots of time spent on your training and development.’ and
529 ‘Diversity among teams. [...]’

530 On the other hand, the *utilities* sector exhibits lower levels of PS. As discussed
531 above, the *utilities* sector is often undergoing transformations due to advances in
532 renewable and green technologies, regulatory changes, and shifting energy policies.
533 This environment of change and uncertainty may impact job security, thereby affecting
534 psychological safety among employees. Example reviews: ‘Difficult to change things
535 for the better. [...] They rather conform everyone to the same standard rather than
536 embracing new and better ideas.’ and ‘Constant pressure. Micro managed in ways that
537 have never been seen before.’ and ‘Current efforts to downsize are disturbing.’ and
538 ‘Lack of communication between coworkers.’

539 In the *communication services* sector, which includes telecommunications, broad-
540 casting, media and entertainment, and internet service companies, lower levels of both
541 PS and OP may result from the fast-paced and dynamic focus on innovation, adapt-
542 ability, and customer demands. This constant change can create uncertainty and stress,
543 which in turn impacts psychological safety. Mergers, acquisitions, and layoffs are com-
544 mon in this sector due to its dynamic nature, further contributing to a sense of job
545 insecurity that negatively impacts psychological safety. Example reviews: ‘So many
546 redundancies, too often.’ and ‘You need to work hard.’ and ‘Terrifying with anxiety
547 24/7. [...] Your job is always on the line, you can be let go at any point.’

548 **OP and PS are related organizational culture aspects, but each indepen-**
549 **dently predicts employee satisfaction.** Although OP and PS scores are correlated
550 (Appendix, Table 5), the interaction term in the joint regression is not significant
551 for most satisfaction ratings, including overall satisfaction (Table 3), indicating that
552 neither aspect amplifies nor diminishes the other’s effect. The exceptions are career
553 and management ratings, where small but significant positive interaction terms show
554 that higher OP or PS amplify each other’s effects. The joint regression results also
555 show that OP effects are significantly stronger than PS effects in predicting ratings,
556 sometimes by an order of magnitude (e.g., 0.227 for OP *vs.* 0.022 for PS when predict-
557 ing overall satisfaction). Accordingly, we do not pursue a combined OP \times PS analysis;
558 instead, for an interested reader, we present an exploratory, construct-by-construct
559 comparison of company standings across the two constructs in Appendix §9.

560 6 Discussion

561 Organisational pride and psychological safety provide a holistic understanding of
562 employee experiences and attitudes throughout the organisational hierarchy. To assess
563 these constructs automatically, we developed and validated an automated language-
564 based framework, which we applied to hundreds of U.S. S&P 500 companies. By
565 employing this framework, we identified significant variations in these constructs
566 across industry sectors and categorised companies based on their respective levels

567 of organisational pride and psychological safety. Next, we discuss the theoretical
568 and practical implications of our work, while also acknowledging its limitations and
569 outlining potential future directions.

570 **6.1 Theoretical Implications**

571 From a theoretical perspective, our work contributes to developing automated ways of
572 modelling workplace-related psychological constructs such as organisational culture.
573 Our work builds upon previous research [8, 56] that aimed to automatically quantify
574 organisational culture on a large scale, particularly by utilising Glassdoor data. We
575 have made technical improvements to enhance the construct validity of this approach.
576 In our study, we proposed a framework that captures organisational constructs from
577 free-form employee reviews about the company by leveraging validated psychomet-
578 ric measures. While we demonstrated the framework’s effectiveness in capturing OP
579 and PS, it is important to note that the framework can be applied to any validated
580 organisational construct as long as the corresponding psychometric measure exists.
581 For constructs that have both positive and negative items in their psychometric mea-
582 sures (as was the case for PS in this study), our framework can directly incorporate
583 them. However, for constructs with only affirmative items (such as OP in this study),
584 the creation of negative items, as detailed in Section 4.1.2, can be beneficial.

585 Our study contributes to the theoretical understanding of organisational culture
586 by examining two psychological constructs of OP and PS. To the best of our knowl-
587 edge, this is the first study to report the varying degrees of OP and PS across different
588 industry sectors, addressing the call for organisational behavioural research to disen-
589 tangle context-specific theories rather than relying solely on context-general ones [71].
590 The consistent relationships between OP, PS, and ratings across the five Glassdoor
591 categories suggest the general importance of these constructs in fostering employee
592 satisfaction. As we demonstrated in Table 3, OP is a strong and robust predictor of
593 all employee satisfaction ratings, with coefficients consistently larger than those for
594 PS. Interestingly, for overall satisfaction, the effect of OP is an order of magnitude
595 greater than that of PS (0.227 vs. 0.022). This likely reflects the fact that OP is mea-
596 sured at the company level and thus aligns closely with overall company satisfaction,
597 while PS primarily captures team-level dynamics. This pattern holds across specific
598 domains such as career, work-life balance, culture, and management. Although PS also
599 significantly predicts satisfaction outcomes, its independent contribution is smaller.
600 The interaction term between OP and PS is generally not significant, with the excep-
601 tion of career and management satisfaction, where the small but significant positive
602 coefficients suggest a modest amplifying effect when both pride and safety are high.
603 However, these interaction effects are limited in magnitude, and the primary drivers
604 of employee satisfaction appear to be the main effects of OP and, to a lesser extent,
605 PS. These findings indicate that initiatives aiming to increase organizational pride
606 may have a particularly strong impact on employee satisfaction, though fostering psy-
607 chological safety can provide additional benefits, especially in areas related to career
608 advancement and management quality.

6.2 Practical Implications

From a practical standpoint, our work offers immediate benefits to organisations by integrating automated psychological assessment into their workflows, conducting sector-specific analyses, and cultivating a positive and productive work environment.

Automated Psychological Assessment. Our automated framework for assessing organisational constructs, including organisational pride, overcomes several limitations associated with conventional self-report assessments. Traditional internal corporate surveys often suffer from method bias and inconsistent implementation across companies. Our framework mitigates biases related to self-report data by using anonymous employee-provided information. This approach ensures more reliable and consistent measurements of organisational pride across diverse industry sectors, allowing organisations to obtain accurate insights into their employees' experiences and attitudes. A practical way to use our framework would be the integration with interactive dashboards where HR and managers can observe OP and PS scores over time, at both team and company levels, identifying trends and pinpointing areas for improvement. These dashboards, coupled with actionable insights and recommendations, can guide strategies to cultivate a positive company culture. For example, if a company scores low on psychological safety, HR could implement specific programs aimed at building trust, such as anonymous feedback tools or conflict resolution workshops. On the contrary, a company with low organisational pride might need to enhance internal communication regarding successes and milestones to boost pride, ensuring that the company's purpose is clearly communicated and that employees understand how their daily tasks fit into the bigger picture. The creation of educational tools and training materials focused on practices that foster organisational pride and psychological safety can educate both employees and leaders alike, and keep everyone engaged and informed about the best practices to enhance the company's cultural health.

Sector-Specific Analyses. By using our framework, any organisation, whether it be a small company or a large corporation, can identify the level of its workforce's organisational pride and psychological safety within its industry sector, and compare it across other sectors. This is important because organisations within specific sectors may face unique challenges and can leverage their strengths to address them effectively. By understanding the sector-specific dynamics related to organisational pride and psychological safety, organisations can tailor their strategies and interventions to create an environment that aligns with their industry's requirements and capitalise on their strengths [72].

Positive and Productive Work Environment. Organisations that prioritise creating a positive work environment can reap numerous benefits. Research has consistently shown that a positive work environment contributes to higher levels of employee well-being and performance [2, 3]. Strategies that foster emotional attachment, commitment, and psychological safety among employees not only can significantly enhance well-being and performance but also productivity. When employees feel valued, supported, and psychologically safe to take interpersonal risks and express their ideas, they are more likely to engage in proactive behaviours and contribute to the organisation's success [1].

653 Our framework’s practical implications extend beyond organisational stakeholders
654 to prospective employees and internal stakeholders. Job seekers can utilise our frame-
655 work to compare organisational culture across companies and make more informed
656 employment decisions (e.g., through visualisation tools; similar tools have been devel-
657 oped to allow prospective employees to compare their values with their prospective
658 employers [57]). This can benefit both job seekers and current employees in reflecting
659 on their assumptions and expectations regarding the work environment [73, 74].

660 **6.3 Limitations and Future Work**

661 Our work is subject to five main limitations, primarily related to the dataset used and
662 how we measured the two psychological constructs studied.

663 The first three limitations pertain to the dataset itself. Firstly, our analysis focused
664 on a subset of S&P 500 companies for which data were available, comprising 318 com-
665 panies. This focus may limit the applicability of our findings to smaller firms with fewer
666 online reviews. For instance, the observed distinctions between OP and PS in relation
667 to employee satisfaction might be less evident in smaller companies, which typically
668 have fewer and less diverse teams compared to larger corporations. To broaden the
669 applicability of our findings, future research could use stratified sampling to gather
670 reviews from a more varied range of businesses, considering both economic status
671 and cultural context. Secondly, it is important to acknowledge the ethno-cultural
672 bias inherent in our dataset, as we exclusively considered U.S.-based companies and
673 English-language reviews. This restricts the cross-cultural generalisability of our find-
674 ings, and it is not clear if and which of our findings would be impacted. To enhance the
675 cultural and linguistic breadth of future studies, researchers could use large language
676 models to process and analyse organisational reviews in multiple languages from differ-
677 ent platforms, thus offering a more nuanced understanding of organisational behaviour
678 across diverse cultural backdrops. Lastly, although our dataset includes timestamps,
679 we did not extensively analyse the temporal dynamics of OP or PS, partly because
680 of the insufficient number of reviews over time for smaller companies, which would
681 not allow for a comprehensive assessment. Future research, leveraging our methodol-
682 ogy and larger temporal datasets, could incorporate longitudinal analysis to uncover
683 time-sensitive effects, providing a more dynamic understanding of organisational cul-
684 ture over time. Examining these temporal aspects could reveal how perceptions and
685 attitudes toward organisations change in response to macroeconomic fluctuations or
686 significant internal changes.

687 The second two limitations pertain to the psychological constructs themselves.
688 Firstly, our assessment of OP and PS focused on the organisation level, based on
689 employee reviews about their overall company experience. While this approach pro-
690 vides a broad overview of psychological safety within an organisation, our findings
691 suggest that PS may not be fully captured at this level because it often operates more
692 meaningfully within smaller teams. PS is inherently relational and context-dependent,
693 deeply influenced by immediate interpersonal dynamics, the support structures of team
694 members, and leadership styles within teams [1]. This suggests that PS may be most
695 accurately assessed within the context of these smaller, functionally cohesive units
696 rather than solely at the organizational level. Our results reinforce this perspective by

697 revealing that organizational-level PS may overlook significant variations across teams,
698 which play a critical role in shaping employees' daily experiences. Treating PS solely
699 as an organisation-wide construct may mask these intra-organisational nuances, as
700 employees within certain departments or project teams could experience significantly
701 different levels of psychological safety. For example, employees working in high-stress
702 or high-risk areas, such as customer-facing roles, may feel varying degrees of PS com-
703 pared to those in internal roles, depending on the unique culture and support systems
704 within their teams. To address this limitation, future studies could adapt the frame-
705 work we experimented with to assess PS at the team level by analysing reviews from
706 employees grouped by team, department, or project if such data is available. Secondly,
707 our measure of employee satisfaction was limited to the online company ratings data
708 currently accessible to our research team. Future research could incorporate additional
709 indices of success, such as those measured by industry standards organisations, annual
710 revenue, or social media sentiment, to provide a more comprehensive understanding
711 of employee satisfaction.

712 **6.4 Ethical Considerations**

713 In this study, we used public and anonymised data obtained from a reputable com-
714 pany review website. The qualitative analysis was conducted using data from publicly
715 accessible sources, and any excerpts utilised to support the analysis were paraphrased
716 to ensure anonymity and avoid the identification of specific reviews. The statistics
717 presented at the company level were derived from publicly available information.

718 **7 Conclusion**

719 This work highlights the importance of understanding employee experiences and atti-
720 tudes through the lens of organisational pride and psychological safety to cultivate
721 a positive work environment and drive organisational success. By utilising a deep-
722 learning framework and analysing a vast dataset of employee reviews from 318 major
723 U.S. companies, our findings emphasise the crucial role of organisational pride and
724 psychological safety in shaping employee engagement, satisfaction, productivity, and
725 innovation. By prioritising organisational pride and psychological safety, organisations
726 can create a work environment that fosters employee well-being and drives long-term
727 success.

728 **List of Abbreviations**

- 729 1. OP: Organisational Pride
- 730 2. PS: Psychological Safety
- 731 3. OC: Organisational Culture
- 732 4. NLP: Natural Language Processing
- 733 5. CSR: Corporate Social Responsibility
- 734 6. ISE: Internal Sustainability Efforts
- 735 7. OCB: Organisational Citizenship Behavior
- 736 8. S&P 500: Standard and Poor's 500

- 737 9. SBERT: Sentence-level BERT
- 738 10. SNLI: Stanford Natural Language Inference
- 739 11. LOP: Lack of Organisational Pride
- 740 12. LPS: Lack of Psychological Safety
- 741 13. TF-IDF: Frequency-inverse Document Frequency
- 742 14. CEO: Chief Executive Officer
- 743 15. CFO: Chief Financial Officer
- 744 16. IT: Information Technology
- 745 17. GICS: Global Industry Classification Standard
- 746 18. U.S.: United States
- 747 19. HR: Human Resources
- 748 20. LLM: Large Language Model

749 **Declarations**

750 **Ethics approval and consent to participate**

751 Not applicable.

752 **Consent for publication**

753 Not applicable.

754 **Availability of data and material**

755 Data and code are available at: <https://github.com/aliakbars/org-pride>.

756 **Competing interests**

757 The authors declare that they have no competing interests.

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764 contributions’ section.

765 **Author’s contributions**

766 AS collected the data and conducted the analysis. AS, SS, MC, and DQ conceived the
767 experiments and wrote the manuscript.

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1018 **Appendix**

1019 **Organisational Pride (OP) & Psychological Safety (PS)**
 1020 **Psychometrics Measures Considered**

1021 OP items by Jones [41]:

- 1022 1. “I am proud to work for [this company].”
 1023 2. “People respect what [this company] does.”
 1024 3. “I am proud to be associated with [this company].”
 1025 4. “I am proud of what [this company] accomplishes.”

1026 OP items by Turban et al. [12]:

- 1027 1. “I would be proud to be an employee at this firm”
 1028 2. “I would receive respect from others if I worked for this firm.”
 1029 3. “My family would feel proud of me if I worked for this firm.”

1030 OP items by Cable and Turban [32]; adapted from [12]:

- 1031 1. “I would feel proud to be an employee of this firm.”
 1032 2. “I would be proud to tell others that I work for this firm.”
 1033 3. “I would be proud to identify myself personally with this firm.”

1034 PS items by Edmondson [1]:

- 1035 1. “If you make a mistake on this team, it is often held against you.”
 1036 2. “Members of this team are able to bring up problems and tough issues.”
 1037 3. “People on this team sometimes reject others for being different.”
 1038 4. “It is safe to take a risk on this team.”
 1039 5. “It is difficult to ask other members of this team for help.”
 1040 6. “No one on this team would deliberately act in a way that undermines my efforts.”
 1041 7. “Working with members of this team, my unique skills and talents are valued
 1042 and utilised.”

1043 **Summary statistics of the variables used in this study**

Table 4: Summary statistics of analyzed variables.

	Mean	Std. dev	Min	Median	Max	Histogram
OP_{score}	0.00	1.46	-4.78	-0.05	4.97	
PS_{score}	0.00	1.29	-3.99	-0.14	4.82	
# reviews	1354.68	2013.33	102.00	452.00	12707.00	
Balance rating	3.30	0.39	2.16	3.33	4.51	
Career rating	3.21	0.37	2.16	3.19	4.47	
Culture rating	3.30	0.43	2.17	3.31	4.64	
Management rating	2.94	0.39	2.00	2.93	4.35	
Overall rating	3.36	0.37	2.25	3.36	4.60	

1044 **Correlations between OP, PS, and overall ratings**

Table 5: Correlation matrix among OP and PS scores, and the overall company rating.

	OP score	PS score	overall rating
OP score	1.00		
PS score	0.69	1.00	
overall rating	0.89	0.67	1.00

1045

1046 **Multicollinearity between OP, PS, and overall ratings**

Table 6: Variance Inflation Factor (VIF) Results.

Variable	VIF
OP score	1.904
PS score	1.896
overall rating	1.009

1047 **Results of Manual Assessment and Baseline Comparison**

1048 We validated our methods on a set of manually annotated reviews against two base-
1049 lines: a phrase-based Rapid Automatic Keyword Extraction (RAKE), and a large
1050 language model (LLM)-based GPT-4o-mini.

1051 First, we selected 400 reviews using stratified random sampling: 200 from pros,
1052 such that 50 were labeled with our method as OP, 50 as PS, and 100 that were labeled
1053 as not pertaining to either of the two constructs; and equivalently, 200 from cons
1054 such that 50 were labeled with our method as LOP, 50 as LPS, and 100 that were
1055 labeled as not pertaining to either of the two constructs. Then, three authors indepen-
1056 dently annotated the 400 reviews for the presence of each construct, as defined in the
1057 literature, keeping the construct items in mind. Specifically, they did so in the follow-
1058 ing way. For the reviews sampled from *cons*, they assigned one of four labels—LOP,
1059 LPS, LOP&LPS, or NA (inapplicable)—to match the automatic framework; likewise,
1060 reviews sampled from *pros* were labeled OP, PS, OP&PS, or NA. For example a cons
1061 review “*HR department is protective of the company, not the employees. Hard to move*
1062 *up the food chain for some positions. Company is so big that individuals are some-*
1063 *times unvalued.*” would be labeled as LOP&LPS since it indicates both a lack of OP
1064 (individuals are unvalued) and PS (HR is protective of the company, not employees),

1065 whereas a pros review “*Conservative and rarely have layoffs.*” would be labeled with
 1066 NA because it mentions job security but does not clearly express high OP or PS.

1067 All three annotators initially agreed on 44.5% of the pros and 49.0% of the cons,
 1068 while two out of the three annotators agreed on 80.3% of the pros and 80.8% of
 1069 the cons. To ensure label quality, the annotators subsequently met to discuss and
 1070 resolve disagreements. These primarily involved clarifying interpretations of the two
 1071 constructs and assessing the level of implied information within brief reviews. For
 1072 instance, some reviews mentioning low job safety were initially labeled as LOP by one
 1073 annotator, but the group collectively agreed this did not equate to LPS. Similarly,
 1074 reviews highlighting high wages or a good work-life balance were occasionally labeled
 1075 as OP by an individual annotator, but upon discussion, it was concluded that these
 1076 alone were insufficient to imply pride in the company. The final consensus labels were
 1077 used for evaluating our method and the baselines.

1078 Table 7 shows that our method outperforms the RAKE baseline implemented using
 1079 `rake-nltk` Python library on the set of 400 manually annotated reviews. Moreover,
 1080 our method also outperforms GPT-4o-mini in predicting three out of four constructs
 1081 from the reviews (OP, LOP, LPS), and is at par for the last construct (PS).

Table 7: Comparison of our method and a keyword-based and an LLM (GPT-4o-mini) baselines.

Method	F1 OP	F1 PS	F1 LOP	F1 LPS
GPT-4o-mini	39.75%	61.64%	28.32%	49.61%
Keyword-based	28.57%	40.91%	8.57%	43.75%
Our method	42.50%	57.80%	34.78%	52.46%

1082 IT Sector Analysis

1083 Upon excluding more generic keywords like ‘benefits’, ‘company’, ‘good’, ‘great’,
 1084 ‘people’, and ‘work’, we found that IT companies with low public sentiment scores
 1085 frequently had positive reviews mentioning “work-life balance” (Figure 5). This con-
 1086 trasted with companies with Company and Teams Leadership (high OP, high PS),
 1087 which did not emphasise work-life balance to the same degree, according to reviews.
 1088 Moreover, “opportunities” and “environment” were frequently cited as advantages
 1089 across reviews.

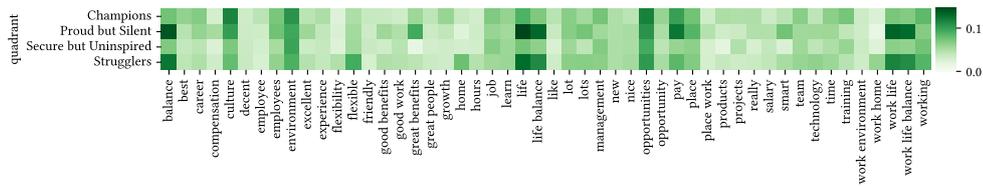


Fig. 5: Keywords associated with each quadrant for IT companies.

1090 Although “work-life balance” was brought up in reviews for all analysed companies
 1091 across different sectors, this theme did not emerge as strongly outside the IT indus-
 1092 try (Figure 6). Interestingly, when PS was low, employees tended to only mention
 1093 “pay” as a positive factor rather than citing deeper advantages like work-life balance
 1094 or opportunities for growth. This suggests that a lack of psychological safety may
 1095 limit employees’ ability to find meaning and fulfilment in their work beyond mon-
 1096 etary compensation. In contrast, the prevalence of “work-life balance” comments for
 1097 IT companies indicates that employees feel safe voicing appreciation for experiences
 1098 beyond just pay when psychological safety is higher.

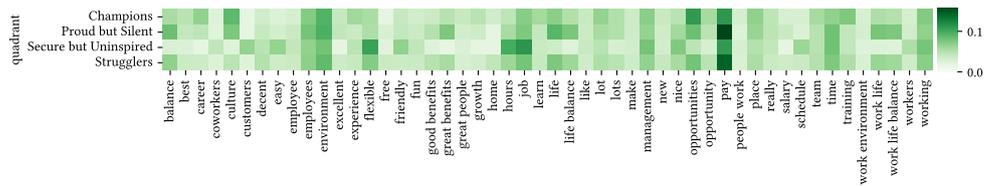


Fig. 6: Keywords associated with each quadrant for all analyzed companies across different sectors.

1099 **Sensitivity Analysis**

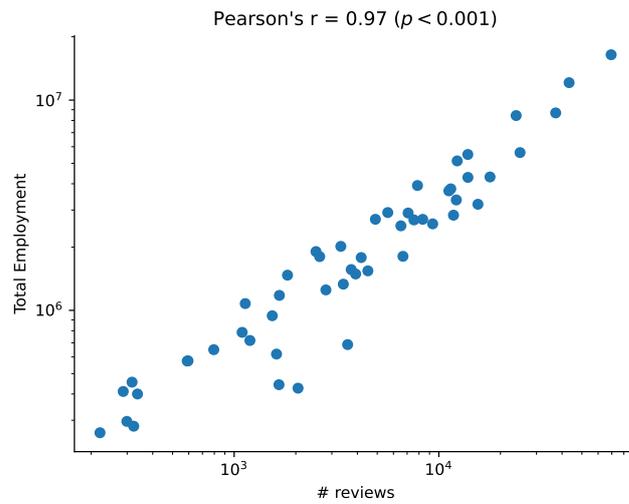


Fig. 7: Correlation between the number of reviews per state and the total number of employment positions per state using Occupational Employment and Wage Statistics from U.S. Bureau of Labor Statistics in 2022.

1100 8 Analysing OP and PS in Combination

1101 To visualise the interplay between OP and PS, we projected the companies onto a
1102 Cartesian plane with two dimensions (Figure 8). This method allowed us to succinctly
1103 and clearly represent each company’s positioning in relation to these two constructs.

1104 To understand what characterises these four quadrants, we identified the most
1105 frequent keywords, including 1, 2, 3, and 4-grams, from the pro and con reviews. Next,
1106 we removed stopwords before calculating the TF-IDF scores for the identified n-grams.
1107 We only kept the top 100 keywords to identify the discriminative ones. Finally, we fit
1108 a multinomial logistic regression model using the top 100 keywords as the input to
1109 predict the quadrant of the companies from which the reviews are coming. Finally, we
1110 identified the most prominent keywords by looking at the model’s coefficients for each
1111 quadrant.

1112 9 Studying OP and PS in Combination

1113 As both OP and PS show significant correlations with overall company ratings
1114 (Appendix Table 5), with Pearson coefficients of $r = 0.89$ and $r = 0.67$, respectively,
1115 it is unsurprising that these constructs both emerged as significant *positive* predictors
1116 of the ratings individually (Table 3). However, given that OP and PS are also moder-
1117 ately correlated ($r = 0.69$), we sought to explore how the interplay between these
1118 two constructs relates to the company scores. Since the variance inflation factor (VIF)
1119 analysis found that the multicollinearity among our variables was low-to-moderate
1120 and, it did not require corrective measures (see Appendix Table 6), we began by exam-
1121 ining a model that predicts employee ratings using OP and PS scores together, while
1122 controlling for the number of reviews. The results in Table 3 show that the model
1123 coefficients for OP are positive when predicting various ratings (ranging from 0.156
1124 for balance to 0.228 for company culture), while the coefficients for PS are also posi-
1125 tive but with a weaker effect (ranging from 0.022 for overall to 0.058 for culture). We
1126 also estimated a regression model incorporating an interaction term between OP and
1127 PS (see Appendix Table 3). The results remained consistent: OP and PS continued to
1128 show a positive association with the rating after controlling for the number of reviews,
1129 while the coefficient for their interaction term was not statistically significant.

1130 We named the four quadrants shown in Figure 8 as *No Designated Leadership (low*
1131 *OP & low PS)*, *Teams Leadership (low OP & high PS)*, *Company Leadership (high*
1132 *OP & low PS)*, and *Company and Teams Leadership (high OP & high PS)* for ease of
1133 reference.

1134 We conducted an ANOVA on the mean company ratings across the four groups
1135 and found statistically significant differences ($F = 105.36, p < 0.001$). The post-hoc
1136 analysis presented in Table 8 further demonstrates that the differences are significant
1137 for most pairwise comparisons between the quadrants.

1138 We next provide a definition for each quadrant, its characteristics derived from the
1139 most prominent keywords (Figure 9), and the sectors that are related to it (Figure 10).

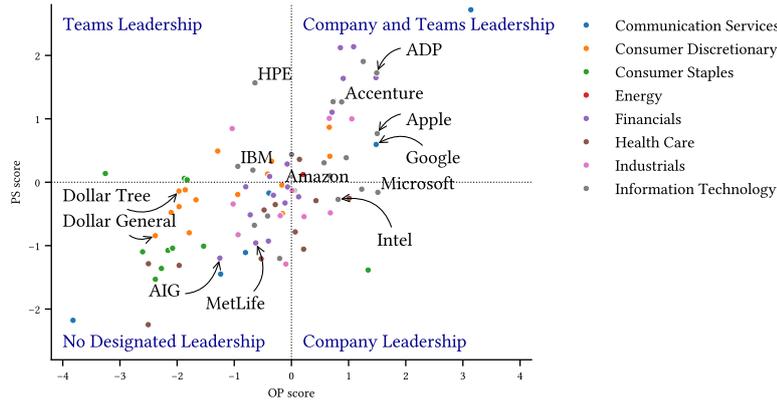


Fig. 8: Scatter plot of the Organisational Pride (OP) vs. Psychological Safety (PS) scores for the top 100 companies based on the number of reviews. Consumer discretionary companies like Dollar Tree and Dollar General displayed lower scores in OP and PS. On the other hand, tech companies like Google and Microsoft scored higher in OP but not as high in PS. It is important to note that Google is classified as Communication Services, according to the Global Industry Classification Standard (GICS) Sector definition.

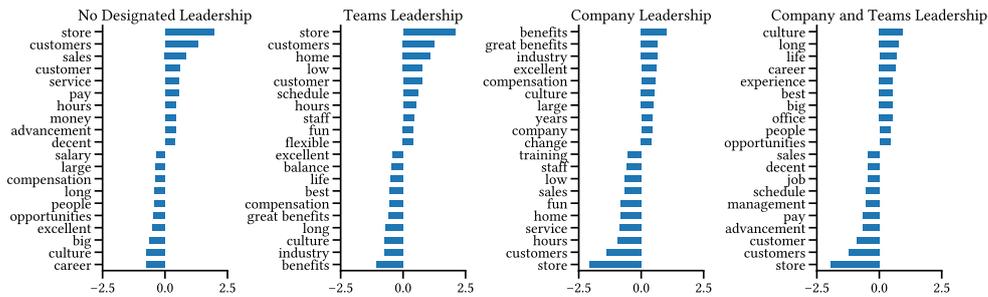


Fig. 9: Top 20 n-grams in sentences expressing OP & PS across the companies in the four quadrants. The bar length signifies the mean coefficients from a 10-fold cross validation of the multinomial logistic regression based on the normalised TF-IDF score. Positive scores imply greater relative relevance of keywords for companies in that quadrant. Negative scores suggest that the mentioned keyword is less prevalent in the companies in that quadrant, compared to others. The notable keywords for *No Designated Leadership* companies include “store”, “customers”, “sales,” and “pay”. For *Team Leadership* quadrant companies, distinctive keywords are “store”, “schedule”, “fun”, and “flexible”. Conversely, in the *Company Leadership* quadrant, distinctive keywords are “large”, “benefits”, “culture,” and “compensation”. Lastly, for companies in the *Company and Teams Leadership* quadrant, distinctive keywords are “opportunities”, “culture”, “career”, and “experience”.

Table 8: Tukey HSD Test Results for Multiple Comparison of Means (FWER = 0.05)

Group 1	Group 2	Mean Difference	p-adj	Reject
Company Leadership	Company & Teams Lead	0.1449	0.008	T
Company Leadership	No Designated Leadership	-0.4393	0.000	T
Company Leadership	Teams Leadership	-0.3889	0.000	T
Company & Teams Leadership	No Designated Leadership	-0.5843	0.000	T
Company & Teams Leadership	Teams Leadership	-0.5338	0.000	T
No Designated Leadership	Teams Leadership	0.0504	0.781	F

1140 **No Designated Leadership.** These companies exhibit a persistent challenge in cul-
1141 tivating a workplace culture that fosters pride and belonging (low OP) while also
1142 struggling to establish a psychologically supportive and safe environment (low PS).
1143 This is evident as employees predominantly focus on retirement (*401K* in Figure 9)
1144 and significantly mention the following words less frequently compared to employees in
1145 the three other company types: *culture* (indicating low pride), and *friendly*, *work-life*
1146 *balance*, and *care* (indicating low safety). Notable companies in this category include
1147 MetLife, Nielsen, AIG, Dollar Tree, and Comcast, primarily representing sectors such
1148 as *financials*, *consumer discretionary*, and *communication services*.

1149 **Teams Leadership.** These companies might face challenges in organisational leader-
1150 ship and fostering a strong sense of belonging (low OP), yet they excel in strong team
1151 leadership by creating a supportive and inclusive environment that prioritises employee
1152 well-being and open collaboration (high PS). This is apparent as employees frequently
1153 discuss *managers*, *coworkers*, and *flexible* work environments (signifying high safety).
1154 Conversely, there is less emphasis on discussions about *great benefits*, *career*, *projects*,
1155 and *culture* (indicating low pride). Companies such as Kmart and Staples in consumer
1156 staples ensure employee safety through a less demanding environment or more tightly
1157 knit teams, but this approach might inadvertently impact the pride associated with
1158 working in these organisations.

1159 **Company Leadership.** These companies are distinguished by their robust organi-
1160 sational leadership, fostering strong pride and loyalty among employees towards the
1161 organisation (high OP). However, they encounter difficulties in establishing a psy-
1162 chologically safe workplace that promotes openness, vulnerability, and collaborative
1163 risk-taking (low PS). This is reflected in employee discussions about *culture*, *great ben-*
1164 *efits*, and *compensation*, (signifying high pride), while conversations about *coworkers*,
1165 *people work*, and *flexibility* are less frequent (indicating low safety). Companies such
1166 as BlackRock and JPMorgan in financials, Intel and Microsoft in information technol-
1167 ogy, and Pfizer and Merck in health care are examples, known for their challenging
1168 and dynamic environments, potentially affecting their employees' psychological safety.
1169 As we have seen from the results on predicting ratings from both OP and PS, in com-
1170 panies with high OP, it is lower PS that contributes to a higher rating. Hence, many
1171 of the companies in this quadrant are highly rated by their employees.

1172 **Company and Teams Leadership.** These companies are lauded for their outstand-
1173 ing leadership, both organizationally (high OP) and in team dynamics, fostering a
1174 supportive, inclusive, and empowering work culture (high PS). Employees in such
1175 environments are generally more motivated and experience greater job satisfaction,

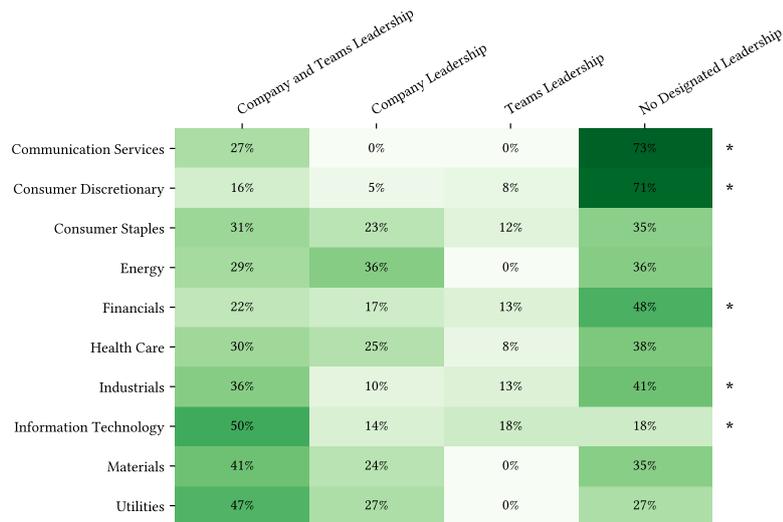


Fig. 10: Distribution of companies per industry sector across the four quadrants. After dividing the companies into quadrants based on OP & PS scores, we can see that Communication Service and Consumer Discretionary companies are more likely to score low in OP & PS (No Designated Leadership) whereas Information Technology companies have high scores in OP & PS (Company and Teams Leadership). The * symbol indicates a statistically significant result using a Chi-squared test at $p < 0.05$.

1176 contributing to higher productivity. This is reflected in employee discussions about
 1177 *projects, career, learning, and work-life balance*, contrasting with lesser focus on *hours*
 1178 and *balance*, and seldom mentioning attributes like *easy* and *free*. As Šćepanović et al.
 1179 [59] suggest, fulfilling workplaces often involve challenges and positive stress, rather
 1180 than ease. These companies span diverse sectors, including notable ones like Apple and
 1181 Nvidia in information technology, Halliburton in energy, Eastman Chemical Company
 1182 in materials, and General Electric Company and FedEx in industrials.

1183 Figure Legends

- 1184 1. Fig 1: **Our framework for measuring an organisational construct score**
1185 **(e.g., OP or PS score) from company reviews.** Starting from the validated
1186 psychometric measures for the construct of interest, we subset negative from
1187 positive items, or, if no negative items are present, we create them as opposites
1188 to the positive items. We then embed each item and measure its similarity to the
1189 embedding of the company reviews. For each company, the proportion of reviews
1190 found similar to the positive statements, gives company positive score, and the
1191 proportion of reviews found similar to the negative statements, gives company
1192 negative score. The final company score for the given construct is calculated as
1193 the difference between positive and negative scores.
- 1194 2. Fig 2: **Top 50 n-grams in sentences expressing OP & PS.** Darker colours
1195 (higher normalised TF-IDF score) indicate greater relative relevance to a partic-
1196 ular construct. The distinctive keywords for OP and PS are “company” and
1197 “team”, respectively.
- 1198 3. Fig 3: **Top 50 n-grams in sentences expressing OP & PS.** Darker colours
1199 (higher normalised TF-IDF score) indicate greater relative relevance to a partic-
1200 ular construct. We removed top shared keywords (*company, good, great, people,*
1201 *team,* and *work*) to let more construct-specific keywords emerge.
- 1202 4. Fig 4: **OP and PS scores across industry sector.** Box plots showing the
1203 distribution of OP and PS scores across different industry sectors.
- 1204 5. Fig 5: **Scatter plot of the Organisational Pride (OP) vs. Psychological Safety**
1205 **(PS) scores for the top 100 companies based on the number of reviews.** Consumer
1206 staples companies like Walgreens, KMart, and Staples displayed lower scores in
1207 OP but higher scores in PS. On the other hand, tech companies like Amazon,
1208 Google, and Microsoft scored higher in OP but not as high in PS. It is important
1209 to note that Amazon is classified as Consumer Discretionary, and Google is clas-
1210 sified as Communication Services according to the Global Industry Classification
1211 Standard (GICS) Sector definition.
- 1212 6. Fig 6: **Top 20 n-grams in sentences expressing OP & PS across the**
1213 **companies in the four quadrants.** The bar length signifies the coefficients of
1214 the multinomial logistic regression based on the normalised TF-IDF score. Pos-
1215 itive scores imply greater relative relevance of keywords for companies in that
1216 quadrant. Negative scores suggest that the mentioned keyword is less prevalent
1217 in the companies in that quadrant, compared to others. The notable keywords
1218 for *No Designated Leadership* companies include “home”, “technology”, “sales,”
1219 and “401K”. For *Team Leadership* quadrant companies, distinctive keywords are
1220 “store”, “customers”, “easy”, and “coworkers”. Conversely, in the *Company Lead-*
1221 *ership* quadrant, distinctive keywords are “smart”, “benefits”, “work-life”, and
1222 “culture”. Lastly, for companies in the *Company and Teams Leadership* quad-
1223 rant, distinctive keywords are “projects”, “culture”, “career”, “learning”, and
1224 “work-life balance”.
- 1225 7. Fig 7: **Distribution of companies per industry sector across the four**
1226 **quadrants.** After dividing the companies into quadrants based on OP & PS
1227 scores, we can see that Communication Service companies are more likely to

1228 score low in OP & PS (No Designated Leadership) whereas Material companies
1229 have high scores in OP & PS (Company and Teams Leadership). The * symbol
1230 indicates a statistically significant result using a Chi-squared test at $p < 0.05$.
1231 8. Fig 8: Keywords associated with each quadrant for IT companies.
1232 9. Fig 9: Keywords associated with each quadrant for all analyzed companies across
1233 different sectors.

1234 Table Legends

- 1235 1. Table 1: Final selected or adapted items from psychometric measures for
1236 measuring OP and PS.
- 1237 2. Table 2: **Predicting company ratings from the standardised OP score.**
1238 We added the number of company reviews as a control variable. The numbers in
1239 parentheses show the standard errors of the coefficients.
- 1240 3. Table 3: **Predicting company ratings from the standardised PS score.**
1241 We added the number of company reviews as a control variable. The numbers in
1242 parentheses show the standard errors of the coefficients.
- 1243 4. Table 4: **Predicting company ratings using both OP and PS scores.** Both
1244 scores are statistically significant to predict overall ratings even after adding the
1245 number of reviews as a control variable. The numbers in parentheses show the
1246 standard errors of the coefficients.
- 1247 5. Table 5: Comparison of our method and keyword-based baseline.