

# Educational Investment and Labor Market Outcomes: A Microeconomic Perspective on Skill Transformation

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## ABSTRACT

This research examines how educational investment and skill development affect labor market outcomes, including income, employment status, and job relevance, among educated individuals in Central Java. Additionally, the study examines how demographic factors such as gender, age, and residential location moderate these relationships. This is a quantitative research study employing a descriptive-inferential approach. It adopts a microeconomic perspective to understand individual-level variation in education-to-employment pathways. The study was conducted in Central Java Province, Indonesia, from July to October 2025. The region was selected due to its diverse socio-economic conditions and high rate of educated unemployment. Using stratified random sampling, data were collected from 400 respondents aged 18-45 who had completed at least secondary education. Data collection instruments included structured questionnaires and semi-structured interviews. The study applied multiple linear regression analysis using SmartPLS to test both direct and moderating effects among variables.

The findings indicate that educational investment has a significant impact on monthly earnings, formal job placement, and job alignment, although it does not notably affect job satisfaction. In contrast, skill transformation, particularly in digital and soft skills, shows a more substantial and consistent influence across all labor market outcomes. Demographic factors, particularly gender and urban-rural location, significantly moderate these relationships, indicating structural inequalities in access and impact.

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*Keywords: Educational Investment, Skill Transformation, Labor Market Outcomes, Job Relevance, Demographic Factors*

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## 1. INTRODUCTION

18 Human capital development is a key pillar  
19 for sustainable economic growth,  
20 particularly in developing regions. Among  
21 its critical components, educational  
22 investment plays a transformative role in  
23 shaping workforce quality and  
24 productivity. Grounded in (Prasetya and

25 Prakarsa 2021) (Becker 1964) human  
26 capital theory, education is framed not  
27 just as a social good but as a long-term  
28 economic investment. (Psacharopoulos  
29 and Patrinos 2018) emphasized that  
30 returns to education, both social and  
31 private, remain significant across global

32 economies, particularly in labor-driven  
33 markets. More recently, (Aslam and  
34 Rawal 2020) reaffirm that well-targeted  
35 educational investments contribute  
36 significantly to individual income growth  
37 and national productivity, primarily when  
38 supported by equitable access.

39 Several studies show that each additional  
40 year of education is associated with  
41 income growth. For instance, (UNESCO  
42 2023) highlights that in lower-middle-  
43 income countries, an extra year of  
44 schooling may increase individual  
45 earnings by approximately 10% on  
46 average. (Montenegro and Patrinos 2022)  
47 confirmed that the rate of return on  
48 tertiary education continues to grow  
49 globally, despite rising levels of degree  
50 saturation. However, these returns are not  
51 uniform across regions, as local  
52 educational quality, labor market  
53 structures, and economic sectors  
54 influence the conversion of education into  
55 income. The heterogeneity of these  
56 effects has not been adequately captured  
57 in Indonesia's subnational labor data.

58 In Central Java Province (Jawa Tengah),  
59 one of the most populous regions in  
60 Indonesia, with over 36 million residents  
61 (Statistik 2023), education and  
62 employment outcomes display a complex  
63 relationship. Although the gross  
64 enrollment ratio in secondary and tertiary  
65 education has increased, the open  
66 unemployment rate among educated  
67 youth remains relatively high, at 5.55% in  
68 2023. Interestingly, BPS also reports that  
69 graduates of vocational high schools  
70 (SMK) face higher unemployment rates  
71 (8.18%) than general high school  
72 graduates, indicating a mismatch between  
73 educational output and market needs.  
74 This issue mirrors similar trends found in  
75 Vietnam and the Philippines, where  
76 technical and vocational education fails to  
77 align with labor demands (Nguyen and  
78 Taylor 2021).

79 These statistics illustrate a critical  
80 problem in the region: skill mismatch.  
81 (Reiskarts and Romele 2025) define this  
82 as the growing gap between the skills  
83 offered by educational institutions and  
84 those demanded by employers. In Central

85 Java, the persistence of educated  
86 unemployment suggests that formal  
87 qualifications alone are insufficient to  
88 secure employment, particularly in a labor  
89 market being restructured by technology,  
90 automation, and platform-based work.  
91 (McGuinness, Pouliakas, and Redmond  
92 2019) Highlight that the mismatch is  
93 highest in lower-income regions, where  
94 labor markets lack mechanisms for  
95 validating skills beyond formal degrees.

96 This brings attention to the concept of skill  
97 transformation, the process through which  
98 individuals acquire adaptive, soft, and  
99 digital competencies to remain  
100 competitive. (Dong, Li, and Chang 2025)  
101 argue that academic-to-vocational  
102 transitions are increasingly crucial in  
103 digital economies. Similarly, a report by  
104 the (Forum 2020) ranked problem-solving,  
105 digital literacy, and teamwork as essential  
106 competencies in the post-pandemic labor  
107 market. Yet, in Central Java, adult  
108 participation in non-formal skill training  
109 remains limited, and the infrastructure for  
110 lifelong learning, such as community-  
111 based technical education or micro-  
112 credentialing systems, is underdeveloped  
113 (Bank 2022).

114 The gap between formal education and  
115 actual employability underscores the  
116 importance of integrating skill  
117 transformation into the broader framework  
118 of educational investment. (Peña-Lang  
119 and Astigarraga 2025) Advocate for  
120 responsive education systems that adapt  
121 to rapid industrial and technological shifts.  
122 In Central Java, where a significant  
123 portion of the labor force is absorbed in  
124 agriculture and informal sectors, upskilling  
125 for digital readiness remains an uphill  
126 challenge. (Prasetya and Prakarsa 2021)  
127 and point out that despite national policies  
128 promoting Industry 4.0 readiness, regional  
129 disparities in digital training access persist  
130 across Indonesia.

131 Prior research in Indonesia has  
132 predominantly focused on national-level  
133 statistics or urban labor centers, often  
134 neglecting microeconomic variation at the  
135 provincial level. For instance, (Suryahadi  
136 2020) examine national employment  
137 shocks but do not address how education

138 and skills interact at the household or  
139 individual level. Studies in Java that adopt  
140 a micro-lens, such as (Trisnawati 2023),  
141 remain limited in scope and fail to  
142 incorporate demographic moderators,  
143 such as gender, age, and urban-rural  
144 disparities. This study, therefore,  
145 addresses a significant empirical gap in  
146 Indonesian labor economics.

147 Hence, this research aims to explore how  
148 educational investment and skill  
149 transformation affect labor-market  
150 outcomes, including employment status,  
151 income, and job relevance, among  
152 educated individuals in Central Java. It  
153 also examines how demographic  
154 variables such as gender, age, and place  
155 of residence moderate these  
156 relationships. The study applies a  
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### 177 **2.1 Research Location and Time**

178 This study was conducted in Central Java  
179 Province, Indonesia, a region  
180 characterized by diverse socio-economic  
181 conditions and varying educational and  
182 employment profiles. The research period  
183 spanned from July to October 2025.  
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### 185 **2.2 Research Type and Approach**

186 This research adopted a quantitative  
187 method with a descriptive-inferential  
188 approach. The goal was to analyze the  
189 statistical relationship between  
190 educational investment, skill  
191 transformation, and labor market  
192 outcomes, while also examining how  
193 socio-demographic factors may moderate  
194 these relationships.  
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### 196 **2.3 Population and Sample**

197 The population targeted in this study  
198 comprised working-age individuals (18-45  
199 years old) residing in Central Java who  
200 had completed at least secondary or  
201 tertiary education. The sampling  
202 technique used was stratified random  
203 sampling, ensuring representation across  
204 education levels and employment  
205 categories. A sample size of 400  
206 respondents was determined to ensure  
207 statistical validity and generalizability of  
208 the results.  
209

157 microeconomic perspective to understand  
158 how individual decision-making on  
159 education and skills translates into  
160 economic returns within a region-specific  
161 context.

162 The findings are expected to provide  
163 contextualized, evidence-based  
164 recommendations for policymakers,  
165 educational institutions, and local  
166 governments by elucidating how  
167 educational pathways and skill acquisition  
168 shape labor market entry and  
169 progression, especially in an increasingly  
170 digitized economy like Central Java. This  
171 study contributes to the development of  
172 adaptive, inclusive, and regionally  
173 responsive education-to-employment  
174 strategies.

## 176 **2. MATERIAL AND METHODS**

### 210 **2.4 Research Variables and 211 Operational Definitions**

212 The variables and their indicators are  
213 summarized in Table 1.  
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### 215 **2.5 Data Collection Techniques**

216 Data were gathered through two main  
217 instruments:

218 Structured questionnaires: Distributed  
219 through both online (Google Forms) and  
220 offline (paper-based) formats, covering all  
221 variable indicators. **Semi-structured**  
222 interviews: Conducted with a selected  
223 sub-sample to capture contextual factors  
224 affecting skill transformation and  
225 employment quality.

### 226 **2.6 Instrument Testing**

227 To ensure the reliability and validity of the  
228 research instruments:  
229 Validity testing was conducted using  
230 Pearson's Product-Moment Correlation,  
231 where items with a correlation coefficient  
232 ( $r$ ) > 0.30 were considered valid.  
233 Reliability testing used Cronbach's Alpha,  
234 with  $\alpha \geq 0.70$  indicating an acceptable  
235 level of internal consistency.

### 236 **2.7 Data Analysis**

237 Data were analyzed using multiple linear  
238 regression with the aid of Smart PLS  
239 software to test the partial and

240 simultaneous effects of the independent 241 variables on teacher performance.

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**Table 1. Research variables and indicators**

Variable Type	Variable Name	Indicators	Scale
Independent Variable	Educational Investment ( $X_1$ )	- Years of formal education - Total education cost - Highest qualification	Ratio / Ordinal
Independent Variable	Skill Transformation ( $X_2$ )	- Number of training programs - Mastery of digital skills - Soft skills (communication, teamwork, adaptability) - Skill-job match	Ordinal
Dependent Variable	Labor Market Outcomes ( $Y$ )	- Employment status (formal/informal) - Monthly income - Job type (aligned with education) - Job satisfaction	Ratio / Ordinal / Nominal
Moderating Variable	Demographic Factors ( $Z$ )	- Gender - Age - Residential location (Urban/Rural)	Nominal / Ratio

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246 **3. RESULTS AND DISCUSSION**

247 **3.1 Descriptive Statistics**

248 From a total of 400 respondents across  
 249 Central Java Province, the majority were  
 250 high school graduates (45%) and  
 251 university graduates (35%), while the  
 252 remaining respondents consisted of  
 253 vocational school (SMK) graduates (12%)  
 254 and diploma holders (8%). The gender  
 255 composition was relatively balanced, with  
 256 51% male and 49% female participants,  
 257 and approximately 60% residing in urban  
 258 areas. The respondents had an average  
 259 of 13.2 years of formal education, with an  
 260 average educational expenditure of IDR  
 261 23.5 million. Furthermore, 67% of  
 262 participants had attended non-formal  
 263 training programs; however, only 38%  
 264 reported that such training was directly  
 265 relevant to their current occupations.

266 **3.2 Interpretation of Hypothesis**

267 **Test Result**

268 **3.2.1 Educational Investment ( $X_1$ ) →**  
 269 **Labor Market Outcomes ( $Y$ )**

270 Result: Significant ( $p < 0.05$ );  $\beta = 0.32$   
 271 Interpretation: Educational investment  
 272 exerts a positive and statistically  
 273 significant influence on labor market  
 274 outcomes. Each additional year of  
 275 schooling or higher educational  
 276 attainment is associated with improved  
 277 monthly income, greater likelihood of  
 278 formal job placement, and better job-

279 education alignment. These findings align  
 280 with (Psacharopoulos and Patrinos 2018)  
 281 and reinforce (Becker 1964) human  
 282 capital theory, which posits that education  
 283 functions as a productive investment that  
 284 enhances individual economic returns.

285 **3.2.2 Skill Transformation ( $X_2$ ) → Labor**  
 286 **Market Outcomes ( $Y$ )**

287 Result: Highly Significant ( $p < 0.01$ );  $\beta =$   
 288 0.45  
 289 Interpretation: Skill transformation  
 290 demonstrates the strongest effect among  
 291 all predictors. Respondents who actively  
 292 engaged in non-formal training—  
 293 particularly in *digital literacy* and *soft*  
 294 *skills*—reported higher earnings,  
 295 improved job relevance, and greater job  
 296 satisfaction. This supports the (World  
 297 Economic Forum 2023) report  
 298 emphasizing adaptive skills as key  
 299 determinants of employability in the digital  
 300 economy. The result suggests that in  
 301 dynamic labor markets, continuous skill  
 302 enhancement has become a more  
 303 immediate driver of employment success  
 304 than traditional educational credentials.

305 **3.2.3 Moderating Effect of**  
 306 **Demographic Factors ( $Z$ ) on the**  
 307 **Relationship between Educational**  
 308 **Investment ( $X_1$ ) and Labor Market**  
 309 **Outcomes ( $Y$ )**

310 Result: Partially Significant ( $p < 0.05$  for  
 311 gender and location)

312 Interpretation: Demographic variables  
 313 such as gender and residential location  
 314 partially moderate the relationship  
 315 between education and labor outcomes.  
 316 The positive impact of higher education is  
 317 more pronounced among males in urban  
 318 areas than among females in rural  
 319 regions. This disparity reflects structural  
 320 inequalities in access, quality, and labor  
 321 market valuation of education. It also  
 322 resonates with global findings on  
 323 gendered returns to education in  
 324 developing economies (ILO, 2022).

### 325 **3.2.4 Moderating Effect of** 326 **Demographic Factors (Z) on the** 327 **Relationship between Skill** 328 **Transformation (X<sub>2</sub>) and Labor Market** 329 **Outcomes (Y)**

330 Result: Significant ( $p < 0.01$  for all  
 331 moderators)  
 332 Interpretation: The moderating role of  
 333 demographics is more consistent in the  
 334 skill transformation model. The effects are  
 335 strongest among younger, productive-age  
 336 respondents (25–35 years) and those  
 337 living in urban areas. This indicates that  
 338 digital and soft-skill training programs  
 339 have not yet achieved equitable reach or  
 340 effectiveness across all demographic  
 341 segments. Rural participants and older  
 342 respondents report lower perceived  
 343 benefits, suggesting gaps in training  
 344 accessibility and adaptability.

### 345 **3.3 Discussion**

346 The results provide substantial empirical  
 347 support for the human capital framework.  
 348 Investment continues to yield measurable  
 349 labor market advantages, and the findings  
 350 reveal that skill transformation now serves  
 351 as a more dynamic and responsive  
 352 predictor of labor success in the digital  
 353 age. This shift indicates that static  
 354 academic credentials are increasingly  
 355 being supplanted by *transferable*  
 356 *competencies*, particularly digital  
 357 proficiency, communication, and problem-  
 358 solving skills, as core drivers of  
 359 employability.

360 Furthermore, the significant moderating  
 361 effects of demographic factors underscore

362 the persistence of inequality in  
 363 educational and skill-based opportunities.  
 364 The gender gap in returns to education  
 365 and the rural–urban divide in skill access  
 366 reflect both socio-economic and  
 367 infrastructural disparities. As noted by  
 368 (UNESCO 2023), digital learning  
 369 ecosystems remain concentrated in urban  
 370 centers, leaving peripheral regions  
 371 underserved. This imbalance may hinder  
 372 inclusive economic participation and  
 373 widen income inequality if not addressed  
 374 through policy reform.

375 The study's evidence also suggests that  
 376 non-formal education and skill training  
 377 programs, when aligned with local labor  
 378 market demands, can bridge structural  
 379 gaps that formal education alone cannot.  
 380 Policymakers should thus prioritize  
 381 flexible, regionally adaptive education  
 382 frameworks integrating both formal and  
 383 informal learning pathways. Such  
 384 integration will enhance workforce agility,  
 385 reduce structural unemployment, and  
 386 promote inclusive labor participation,  
 387 particularly among marginalized groups.

388 In summary, while education remains  
 389 foundational, the synergy between  
 390 educational investment and continuous  
 391 skill transformation determines labor  
 392 market competitiveness in the digital era.  
 393 Equitable access to skill development  
 394 opportunities across gender, geography,  
 395 and socio-economic strata should  
 396 therefore become a central agenda in  
 397 Indonesia's human capital policy.

## 398 **4. CONCLUSION**

399 This study examined the influence of  
 400 educational investment and skill  
 401 transformation on labor market outcomes  
 402 in Central Java Province, Indonesia, while  
 403 considering demographic factors as  
 404 moderating variables. Based on data  
 405 analysis from 400 respondents of  
 406 productive age, several key conclusions  
 407 emerged:  
 408

409 Educational investment exerts a positive  
 410 and significant effect on labor market  
 411 outcomes, particularly in enhancing

412 monthly earnings, formal employment  
 413 status, and the alignment between  
 414 educational background and job type.  
 415 This finding reinforces the principles of  
 416 Human Capital Theory, confirming that  
 417 education remains an essential and  
 418 valuable economic asset.

419 Skill transformation demonstrates a  
 420 stronger impact than formal education  
 421 across all indicators of labor market  
 422 success. Mastery of digital competencies,  
 423 soft skills, and participation in non-formal  
 424 training programs substantially contribute  
 425 to higher productivity levels and greater  
 426 job satisfaction.

427 Demographic factors, including gender,  
 428 age, and residential location, significantly  
 429 moderate the relationships between  
 430 education, skills, and labor outcomes.  
 431 Disparities in training access and  
 432 educational quality in rural areas, as well  
 433 as gender gaps in employment results,  
 434 highlight the need for more inclusive and  
 435 equitable policy interventions.

436 The mismatch between educational  
 437 output and labor market demand remains  
 438 a major challenge. The relatively high  
 439 unemployment rate among vocational  
 440 school (SMK) graduates indicates a  
 441 persistent skills mismatch, underscoring  
 442 the importance of curriculum reform and  
 443 stronger industry–education partnerships  
 444 to align competencies with market needs.

445 **DISCLAIMER (ARTIFICIAL**  
 446 **INTELLIGENCE)**

447 Author(s) hereby declare that NO  
 448 generative AI technologies such as Large  
 449 Language Models (ChatGPT, COPILOT,  
 450 etc.) and text-to-image generators have  
 451 been used during the writing or editing of  
 452 this manuscript.

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