

Conformity and Soft Skills as Determinants of Alternatively Credentialed Non-College Graduate hirability

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Abstract

Despite targeting technical skills, vocational school graduates are paid less than college graduates. This paper hypothesizes that conformity selection and a perceived deficit in soft skills substantially explain reduced alternatively credentialed non-college graduate (ACNG) hirability. Microdata from the United States confirm a perceived soft skill deficit for ACNG labor. Results also indicate that conformity is a key factor of hirability, but the direction of effect is heterogeneous by employer type. Conformity and perceived skill gaps explain about one third of hirability variance. Perceived soft skill gaps explain about as much as the sum of the effects from state of residence and industry of occupation. Opposite a conventional explanation, the results of this study suggest that hirability is negatively correlated to conformity on average. ACNG job candidates tend to be perceived as creative types and an even mix of high and low performers. Evidence of risk aversion from employers with respect to labor productivity is found, which contributes to lower ACNG demand. The conclusion incorporates discussion of public misperception on the cost of vocational school and suggests that nontraditional postsecondary education is undervalued in the United States.

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1. Introduction

A substantial gap exists between the skills expected by employers and those possessed by college graduates[1, 2, 3, 4]. Experts view college alternatives including vocational school as useful for technical training, but the traditional college degree retains a wage premium over vocational education. Unemployment, underemployment, and other negative labor outcomes follow a similar pattern[5]. This paper seeks to resolve the apparent discrepancy between these outcomes while preserving the mainline view that employers pay for perceived job candidate skill. To explain the apparent discrepancy, this paper tests the hypothesis that employers expect an offsetting non-technical skill deficit when considering an alternatively credentialed non-college graduate (ACNG). I find evidence that the general population of the United States, including employers, does apply a stigma to the ACNG in which soft skills are assumed to be deficient.

Alternative credentials refer to credentials other than the undergraduate degree[6]. The category includes, for example, industry certifications, portfolios of work, digital badges, and other records of unaccredited learning and achievement. Individuals pursuing alternative credentials typically intend to leverage the credential toward better employment. That is, they typically have the same goals as a college student. Many individuals obtain alternative credentials as a supplement to the college degree. Such a situation is pareto-superior to degree attainment alone and is therefore intentionally excluded from analysis. This paper focuses on factors of ACNG hirability in order to validate whether ACNG labor outcomes are a general problem, or a problem limited in scope to specific skills. If the gap is limited to a particular set of skills, such as soft skills, then alternative credential suppliers could modify their credentialing requirements to overcome the outcome deficit.

The signalling model has become one of the two standard explanations of the value of the college degree. Signalling theory provides three advantages over human capital theory for the purposes of the present study. First, signalling

theory is able to explain labor outcome variance when human capital is held constant.

Second, the signalling model empowers a survey research design. In an idealized human capital model, the measures of human capital would correspond
35 to production process inputs. To establish a wide array of skill measures would be complicated and prone to measurement sensitivities, assumptions, and errors of various and potentially subtle kinds. Survey measures could be used as a second-best proxy, but they would never be an ideal measure of human capital. Signaling theory takes the reverse approach. According to the signalling model,
40 labor demand is formed on the basis of job candidate value as perceived by an employer. Whether this corresponds to any concrete ability is secondary. Employer perception can be assessed through a simple survey. An additional benefit of using a questionnaire is the ability to ask hypothetical questions. In pondering hypotheticals, employer evaluation of a credential or signal can be
45 isolated from job candidate human capital variance.

Third, signalling theorists have laid out a testable hypothesis for weak labor outcomes among non-college graduates. Following this model, scholars claim that the college degree signals intelligence, conscientiousness, and conformity[7]. Non-traditional education, in contrast, is hypothesized to signal nonconformity.
50 Non-traditional courses can also be completed in a shorter span of time and with reduced entry qualifications relative to the traditional degree. For this reason, alternative credentials are thought not to signal conscientiousness.

Research indicates a goldilocks level or bliss point for both conscientiousness and conformity is likely to exist. Excess individual conscientiousness can disturb team performance[8]. Conformity can lead to a lack of innovation and suboptimal organizational practices[9]. Psychologists also state that conformity selection may occur through heuristic decisioning rather than conscious choice.

Risk aversion is another explanation for conformity selection. An employer may not be able to value an alternative credential. From the point of view of
60 such an employer, an ACNG may range in value from a positive outlier to a negative outlier. The employer may not prefer to hire such a candidate on the

basis of risk aversion, even if their point estimate for ACNG labor value is higher than their point estimate for a recent college graduate. If employers with many employees are positively associated with ACNG hirability, this will add weight
65 to an explanation based on risk aversion.

2. Data and Methodology

The hypothesis in this paper is based on a simple model of demand for labor which is clarified in Equations 1a and 1b:

$$S_j = f(H_j) \tag{1a}$$

$$w_{ij} = E_i(MRP_j) = f_i(S_j) \tag{1b}$$

70 Job candidate j , generates a signal of productivity, S_j from unobserved human capital, H_j . Employer i , forms an expectation of the marginal revenue product of j on the basis of an employer-specific evaluation of S_j . The specific employer is willing to pay the specific job candidate wages of w_{ij} .

This study uses ordinary least squares (OLS) regression analysis to estimate
75 the effect of perceived skill gaps on hirability. hirability is a proxy for w_{ij} . That is, an employer would be willing to pay more for a candidate with higher hirability. In order to explain reduced willingness to pay for ACNG labor relative to college graduate labor, this paper hypothesizes that employers preferentially value soft skills. If employers preferentially value soft skills then soft skill gaps
80 should be identified as important and negatively relate to ACNG hirability.

Regression analysis was conducted using original cross-sectional data from an online self-completed questionnaire ($n = 212$). The data is available for replication or any other use¹. Respondents were obtained through the Amazon Mechanical Turk crowdsourcing service. Respondents were United States

¹See https://osf.io/8qtxf/?view_only=95b0c0b0c65e4b7983198cc87c2ab733 for data used in this study.

85 citizens at or over the age of eighteen, paid for participation, and selected on
an opt-in, first-come, first-serve basis. The survey administration took place in
July of 2020.

The survey includes 65 questions in two sections². The first section captures
respondent characteristics and the second section captures perceived skill rel-
90 ative to hypothetical job applicants. Employer responses did not significantly
differ from the general population, so respondent characteristics are also inter-
preted as employer characteristics.

Regression variables in this study are categorical or Likert-type responses
based on a scale from 1 to 10. Higher Likert-type values indicate greater agree-
95 ment with a statement that varies by variable. Categorical variables include
state of residence, industry of occupation, employer status, firm size, and a
measure called duration.

Duration measures the length of time the respondent believes it takes to
obtain an alternative credential. Employer status describes whether an individ-
100 ual makes hiring and firing decisions in the course of their employment. The
variable takes one of three values: yes, no, or unemployed. Employer effects
refer to the case where an individual states that they do make hiring and firing
decisions. State of residence refers to a state within the United States. Respon-
dents were allowed to select the District of Columbia as a state of residence, but
105 no such responses were obtained.

Three other factor groups are investigated in the regression analysis. These
variables are measured using Likert-type units and they include hirability, rule-
breaker effects, and perceived skill gaps. hirability is the dependent factor and
it indicates the degree of agreement that, "For many professions, alternative
110 credentials can qualify a person for an entry-level position."

Rulebreaker effects refer to a collection of three factors that measure re-
spondent agreement with statements about rulebreakers, or "People who are
willing to break formal or informal rules and norms." The first statement indi-

²See Appendix A for a full copy of the survey.

cates that rulebreakers present a risk to the reputation, productivity, or value
115 of a company. This statement received the least agreement and greatest response variance among three qualitatively different descriptions of people that are willing to break rules ($\mu = 6.40, \sigma = 2.55$).

The second statement indicates that rulebreakers are held back by rules and "could just as easily be high performers as low performers." This statement
120 received the most agreement and least variance among rulebreaker statements ($\mu = 7.42, \sigma = 1.91$). The agreement with this statement provides evidence against the thesis that employers value conformity for its own sake. In turn, this adds weight to the theory that employers value conformity as a risk aversion tactic while knowing that nonconformity signals positive outlier potential. The
125 third description of rulebreakers states that they tend to be gifted in the areas of innovation or creativity, and that such people may benefit the culture of a company ($\mu = 7.25, \sigma = 2.03$).

Perceived skill gaps are computed two ways from perceived skill questions in the second section of the survey. Perceived skill gaps are measured separately
130 with and without overqualification effects. Overqualification effects have been identified as important[10, 11], but these effects are sometimes ignored during skill gap analysis[12].

Perceived skill is a Likert-type response reporting agreement with the statement that a particular candidate has a particular skill. For each of 13 skills,
135 the respondent is asked to imagine and report skill levels for the ideal candidate, the average actual employee, the average recent college graduate, and the average ACNG. Raw perceived ACNG skill gaps are calculated by differencing the perceived skill of an ideal candidate with the perceived skill of an ACNG. The perceived skill gap with overqualification effects equals the raw perceived
140 skill gap. The perceived skill gap without overqualification effects is calculated as the raw skill gap or zero if the raw skill gap value is negative.

Rulebreaker effects and perceived skill gaps are structurally linked. The willingness to break formal and informal norms is a soft skill and respondents did evaluate this skill. Structurally, the gap in the rulebreaking skill interacts

145 with employer perception on rulebreaking. For this reason, discussions on the
importance of skill gaps will consider the importance of rulebreaker effects as
well.

These methods allow for identification of a preferred model that explains
hirability using perceived ACNG skill gaps. The identified model will support
150 the hypothesis if soft skills are more important than technical skill gaps. The
model will support the risk aversion explanation of ACNG hirability over an ex-
planation from conformity selection if large employer size is positively associated
with hirability.

Comparative analysis provides additional confidence in the data by repli-
155 cating a hirability gap between ACNG labor and recent college graduates. A
comparative skill gap variable is constructed for each perceived skill gap that
is important in the preferred hirability model. Comparative skill gap variables
are constructed by subtracting perceived recent college graduate skill from per-
ceived ACNG skill. Multiple regression of these comparison factors on hirability
160 demonstrates which, if any, perceived skill gaps are important distinguishers of
the ACNG from the college graduate. Identification of significant differences
with a negative total effect on hirability will replicate external data on the
lower job market value of ACNG labor and provide a diagnostic on which skill
or skills must be better addressed through alternative learning programs.

165 **3. Results**

3.1. Identification of the Preferred Model

Results confirm that employers, and the population in general, associate a
soft skill deficit with ACNG candidates. At the same time, ACNG hirability
was generally positive. The mean response was 7.5 on a scale from one to ten
170 ($\sigma = 1.80$). hirability critically depends on rulebreaker effects. Rulebreaker
effects have more explanatory power than perceived skill gaps. Evidence favors
an explanation of ACNG hirability from risk aversion over conformity selection.
Employer status was associated with an insignificant positive coefficient.

Table 1 reports selected factor statistics across five OLS multiple regressions.

175 The selected factors which are reported include any perceived skill gap which is important in any specification. Factor importance is determined by the ability of a factor to improve adjusted explanatory power. Model 1 is a multiple regression using skill gaps that allow for overqualification. Model 2 is a multiple regression without overqualification.

180 Models 3 and 4 are equivalent to models 1 and 2, respectively, after normalizing for industry, state, and firm size effects. These effects are normalized for robustness by retaining those factors which appear in both models 1 and 2. For example, Alabama had a significant effect when skills are measured with overqualification in Model 1. Alabama was not significant in Model 2, so it was
185 dropped in models 3 and 4.

Model 5 is specified as Model 4 plus two adjustments. First, the factor for salary is dropped. The salary factor improved adjusted explanatory power in Model 2, but it provided no such benefit in any other model. Moreover, the p-value of this factor was unacceptably low in Model 4 ($p > 0.9$).

190 The second adjustment is to insert a variable for duration³. The belief that it takes more than a year to obtain an alternative credential is importantly and positively associated with willingness to hire ($\beta = 0.875, p < 0.01$). Employer effects are positively signed in all five models, but the significance is lost after normalizing effects. This suggests that ACNG hirability is sensitive to industry,
195 state of residence, and firm size, which are the normalized effects.

The preferred model is able to explain roughly one third of the variance in hirability. Rulebreaker effects are identified as significant regardless of specification. Seven perceived skill gaps are included in the preferred model. The perceived technical skill gap is included in the preferred model. The coefficient

³Duration is a categorical variable which was important in both Models 1 and 2. As a categorical variable, it was decomposed into a boolean series for factor analysis. Models 1 and 2 retained one or more duration boolean factors, but none overlapped. As a result, duration was dropped from Models 3 and 4.

Table 1: Table of Coefficients for Multiple Regressions on hirability, Selected Variables

	Model 1	Model 2	Model 3	Model 4	Model 5
Gap, Body Language	-2.240e-01* (8.314e-02)	-3.831e-01** (1.124e-01)	-1.507e-01+ (8.980e-02)	-3.155e-01* (1.173e-01)	-3.060e-01* (1.145e-01)
Gap, Body Language-IT	2.199e-01+ (1.269e-01)	2.298e-01 (1.656e-01)	1.837e-01 (1.334e-01)	2.791e-01 (1.707e-01)	2.771e-01+ (1.665e-01)
Gap, Commute		-2.320e-01++ (9.720e-02)	-4.953e-02 (6.862e-02)	-1.197e-01 (1.023e-01)	-1.582e-01 (1.010e-01)
Gap, Conscientiousness	2.416e-01* (8.000e-02)	3.223e-01* (1.045e-01)	1.387e-01 (8.483e-02)	2.174e-01+ (1.129e-01)	2.175e-01++ (1.093e-01)
Gap, Customer Service	-1.259e-01+ (6.389e-02)	-1.512e-01 (9.599e-02)	-1.253e-01+ (7.162e-02)	-1.276e-01 (1.037e-01)	-1.323e-01 (1.009e-01)
Gap, Rule Breaker		-6.336e-02 (1.028e-01)	-3.896e-02 (6.054e-02)	-8.535e-02 (1.082e-01)	-1.034e-01 (1.036e-01)
Gap, Salary		-1.135e-01 (8.284e-02)	3.873e-02 (6.597e-02)	-6.250e-03 (8.575e-02)	
Gap, Teamwork		1.227e-01 (9.179e-02)	6.812e-02 (6.963e-02)	1.287e-01 (9.697e-02)	1.131e-01 (9.505e-02)
Gap, Technical	-1.274e-01+ (7.443e-02)		-9.408e-02 (7.702e-02)	-1.010e-01 (1.023e-01)	-9.806e-02 (1.001e-01)
Rulebreaker, Culture Add	2.612e-01** (7.057e-02)	2.829e-01** (7.015e-02)	2.114e-01* (7.187e-02)	2.279e-01* (7.190e-02)	2.235e-01* (7.036e-02)
Rulebreaker, Risky	1.688e-01** (4.993e-02)	1.758e-01** (4.813e-02)	1.517e-01* (5.160e-02)	1.472e-01* (5.063e-02)	1.686e-01** (5.006e-02)
Rulebreaker, Rockstars	1.406e-01+ (7.646e-02)	1.748e-01++ (7.245e-02)	1.669e-01++ (7.851e-02)	1.546e-01++ (7.754e-02)	1.655e-01++ (7.599e-02)
Adj R-sqr	0.3100	0.3491	0.2317	0.2554	0.2866
R-sqr	0.4408	0.4663	0.3409	0.3613	0.3880

Standard errors in parentheses. + $p < 0.10$, ++ $p < 0.05$, * $p < .01$, ** $p < .001$

For each model, the probability of a greater F-statistic is less than 0.0001 and the sample size is 212.

200 is statistically insignificant, but it is robust in sign across models and it does possess the expected negative sign. The other six perceived skill gap factors in the preferred model are soft skills.

In the classic signalling explanation of ACNG aversion, the ACNG is expected to have a deficit of conscientiousness and conformity. The explanation of ACNG aversion due to conformity selection is constrained in explanatory power
205 because nonconformity is generally valued. The perceived conscientiousness gap is not significantly different between an ACNG and the average recent college graduate. The data does validate that soft skills altogether are importantly different for ACNG and recent college graduate labor. Conscientiousness and body
210 language are the two most important skill gaps in the model. Other soft skills are less significant, but the coefficients of all soft skills in the preferred model are strictly larger than the coefficient for technical skill.

An important and complicated finding involves conscientiousness. The effect is robustly positive in multiple specifications. Simple intuition would indicate
215 that a large conscientiousness gap is associated with reduced hirability. A simple regression of conscientiousness on hirability does produce the expected negative coefficient. There are two reasons for the sign change on conscientiousness in the multiple regression. The first reason is bliss point seeking and the second is an explanation from attenuation.

220 As previously discussed, excess individual conscientiousness can disturb team performance[8]. It follows that there is some bliss point level of conscientiousness. As a result, the sign of the coefficient on conscientiousness is sensitive to the relationship between the model constant and the bliss point. If the model constant is above the bliss point, conscientiousness is expected to be positively
225 signed. The current data is consistent with this explanation from bliss point seeking. In a simple regression of conscientiousness in hirability, the model constant is about 7.5, and the coefficient on conscientiousness is negative. In the multiple regression, the model constant is near 0.5, and the coefficient on conscientiousness is positive.

230 The second explanation is that the direct measure of conscientiousness is

attenuating an overstatement of the effect in the multiple regression. Conscientiousness is importantly cross-correlated with several factors including willingness to commute and customer service skill. Conscientiousness also structurally interacts with rulebreaker effects. In theory, a person that is high in
235 conscientiousness will tend not to break rules. These factors are entered independently in the multiple regression, so a partial measure of conscientiousness is entered in redundantly. The direct factor for conscientiousness corrects, or attenuates, the overstated effect which is represented in those the correlated independent factors. Removal of other skill gap factors and rulebreaker factors
240 from Model 5 demonstrates this by yielding a negative conscientiousness coefficient ($\beta = -0.084, p < 0.31$).

The importance of conscientiousness does not add weight to the classic signalling explanation. The ACNG conscientiousness gap and the recent college graduate conscientiousness gap are not significantly different.

245 Employer size is an important factor in the preferred model. The largest category of employer is positively associated with willingness to hire an ACNG. This matches the risk aversion explanation of ACNG hirability.

Some state and industrial effects are identified. No particular relation among state effects was found, but further comparative policy research is encouraged.
250 With respect to industry, an interesting interaction between body language skill and employment in the information technology industry yields a positive coefficient. Body language skill gaps on their own are associated with reduced willingness to hire. This specifically indicates a reduced penalty for lacking body language communication skills in the information technology industry. With less
255 confidence and more generality, a positive coefficient to this interaction variable indicates a reduced penalty for generalized soft skill deficiency in the information technology industry.

A reduced penalty for soft skill deficit helps explain the particular flourishing of alternative credentials in the information technology industry. The reduced
260 penalty in this particular industry might be related to a relative lack of deregulation in the industry. Another explanation is that the reduced penalty may be

related to cultural norms in the industry. There is less technical need for social skill in programming, so introverts may obtain a comparative advantage in this field.

265 3.2. *Explanatory Share of Perceived Gaps*

The preferred model explains about one third of hirability variance, but how much of the explanatory power is attributable to perceived skill gaps? Table 2 provides evidence on the importance of perceived skill gaps and rulebreaker effects. This table compares the explanatory power of selected factor groups.
270 Industry and state effects are widely regarded as important in explaining labor outcomes. The table shows that perceived skill gaps and rulebreaker effects are even more important in explaining hirability. Rulebreaker effects collectively explain more than three times as much response variance as do industrial or state effects.

275 Allowing for overqualification seems to weaken explanatory power. Overqualification effects seem to be heterogeneously signed per skill, so generalizing weakens overall explanatory power relative to ignoring these effects. With overqualification, perceived skill gaps explain about fifty percent more than industrial effects or robust state effects. Without overqualification, the adjusted explanatory
280 power of perceived skill gaps is about three times the adjusted explanatory power of industry or state effects. Semi-robust state effects are state effects which are significant in any multiple regression described in Table 1. Robust state effects are the significant factors in a simple regression of semi-robust state effects on hirability. An example of a state which appears in a multiple regression
285 from Table 1 but is not significant in a simple regression of semi-robust state effects on hirability would be [TODO].

Table 2 also describes the explanatory power of so-called rulebreaker effects. Whether the candidate is perceived as a rulebreaker is a perceived skill gap, but employers evaluate this gap in a heterogenous and multispecific way. In the
290 first place, this heterogenous evaluation has sign and magnitude implications for the dependent variable of interest. Secondly, heterogenous evaluation implies

Table 2: Factor Group Explanatory Power in a Simple Regression

Effect Group Name	Adj R-Sqr	R-Sqr	Max p-value
Industry	0.0185	0.0510	0.288
Rulebreaker	0.1432	0.1554	0.053
Skill Gaps with Overqualification	0.0558	0.0737	0.106
Skill Gaps without Overqualification	0.0758	0.0933	0.115
State, Robust	0.0177	0.0503	0.227
State, Semi-Robust	0.0034	0.0648	0.831

a qualitatively different evaluation. These differences are captured using three questions in the first section of the survey.

Employer status is significantly and positively related to each of the rule-breaker effects. Rulebreaker effects are about twice as important as perceived skill gaps. These results provide evidence that ACNG hirability depends importantly on the way a particular employer views rulebreakers.

3.3. Perceived Gaps Compared to the Recent College Graduate

Table 3 provides two models of hirability based only on comparative skill gaps. Model 6 is specified with differences in each perceived skill gap that was identified as important in Model 5. Model 7 consolidates factors in order to optimize adjusted explanatory power. The four factors that were eliminated in the consolidation had a p-value greater than 0.55. This indicates all four eliminated factors were more likely not to have a relation than to have a relation. The maximum p-value in Model 7 is less than 0.16.

The preferred model demonstrates that a perceived soft skill deficit explains

Table 3: Multiple Regression of Comparative Skill Gap on Hirability

	Model 6	Model 7
Body Language	-3.295e-01*	-3.395e-01*
	(1.138e-01)	(1.049e-01)
Commute	1.498e-01	1.574e-01
	(1.101e-01)	(1.069e-01)
Concientiousness	1.416e-01	1.508e-01
	(1.119e-01)	(1.061e-01)
Customer Service	-1.493e-02	
	(1.143e-01)	
Technical	4.955e-02	
	(1.164e-01)	
Teamwork	1.552e-02	
	(1.008e-01)	
Rulebreaker	-5.822e-02	
	(1.108e-01)	
Adj R-sqr	0.0311	0.0474
R-sqr	0.0633	0.0609
p(F)	0.0610	0.0044

14

Standard errors in parentheses

* $p < .01$, ** $p < .001$

ACNG hireability, but it does show that the soft skill deficit is a comparative difference between ACNG labor and the recent college graduate. Models 6 and 7 demonstrate that a few soft skills do constitute a significant difference between groups. Model 7 in combination with summary statistics on the average value of each factor also demonstrates that the total hireability effect for the collection of these three factors is negative on average⁴. This replicates external research which shows lower demand for ACNG labor.

Model 7 also provides a skill-level diagnostic in service of closing the demand gap. This model indicates that the soft skill of communication using body language is the only skill difference associated with reduced hireability. It is also the most significant and largest in magnitude of such differences. This result should not be interpreted as a hireability difference due to generalized communication skill. Perceived skill gaps for verbal and written communication are insignificant.

4. Conclusion

This study provides evidence that skill signals are an important factor of hirability and are unique for the ACNG. Perceived skill gaps do a better job of explaining willingness to hire than do other widely recognized effects including industry and state effects. Employer factors better explain candidate hirability than do the perceived skill gaps themselves. Technical skill gaps were identified with less relevance to the hiring decision when compared with soft skill gaps for the ACNG job candidate.

This paper provides evidence that some employers engage in conformity selection as a means of avoiding risk to the reputation, productivity, or value of a company. An explanation from risk aversion fully this kind of conformity

⁴The respective mean values for the comparative gaps in perceived body language skill, willingness to commute, and conscientiousness are -0.3395, 0.1574, and 0.1508. The total average hireability effect attributable to these effects is described by the equation: $Y = -0.3395 * 0.1415 - 0.1575 * 0.0943 + 0.1508 * 0.0330 = -0.0579$.

selection and also explains other behavior. Respondents were most favorable to the description of rulebreakers as individuals that could just as easily be high performers as low performers. Aversion to this kind of labor is better explained
335 as risk aversion rather than positive selection for conformity.

Risk aversion and conformity selection are both partially unconscious biases which lead to suboptimal organizational operation. A practical recommendation is for organizations to implement bias controls with respect to ACNG evaluation. An example control would be to provide human resource procedures for formal
340 evaluation of particular credentials which are relevant to specified job families. These procedures can be immediately executed among known credentials and job families. These procedures should be retained for ongoing application as new credentials are developed and encountered over time.

Some evidence on the role of misinformation is demonstrated in a survey
345 on trade schooling taken in 2019[13]. Only 27 percent of respondents correctly responded that lower debt is an advantage of enrolling in trade school relative to college. Additionally, over 75 percent of respondents failed to notice that trade school graduates receive industry employment sooner and receive specialized training when compared to a four-year college. Obtaining a college degree after
350 obtaining some work experience will allow students to leverage employer tuition benefits. Because ACNG hirability varies importantly by particular employer, ACNG job candidates are advised to apply to a substantial number of employers at the outset of the job search. Insight into ACNG hirability at a particular firm can be obtained prior to application through social networking with employees
355 of the firm, online research into the policies of the target employer, or consulting a recruiter that specializes in the target employment industry.

The preferred model explains about one third of willingness to hire. Perceived skill gaps and rulebreaker effects account for most of the explanatory power in the model. The explanatory power of this study can be meaningfully
360 improved in a few ways. Longitudinal study would allow for causal analysis and improve forecasting of ACNG hirability in the future. Other research has conducted some dynamic analysis of the same dependent variable with differ-

ent regressors[14]. Integrated analysis would be useful for replication and the generation of new models of better explanatory power. Analysis that includes
365 overqualification effects and heterogeneous nonlinear relations between skill gaps and hirability would improve estimates of hirability for a candidate of a particular skill profile.

This paper noted that large employers and the information technology industry has a peculiar susceptibility to alternative credentials, so recent changes
370 implemented by Google may be indicative of future trends. Google has not required a college degree since prior to 2013[15]. Laszlo Bock, then Senior Vice President of People Operations at Google, stated the following in 2013: "After two or three years, your ability to perform at Google is completely unrelated to how you performed when you were in school, because the skills you required in
375 college are very different." In 2020, Google added three new certificate programs to an existing set and declared that all of its certificates would be treated as the equivalent of an undergraduate degree for their hiring purposes[16].

If perceived skill is representative of actual skill, then the current study concludes that employers should be more willing to hire an ACNG. At the
380 same time, this paper provides evidence that perceived and actual skill levels sometimes do not align. For example, the average recent college graduate in the sample is perceived to have better technical skill compared to the average ACNG. This is surprising because last mile training, a kind of alternative education, has been specifically recommended in popular literature as a remedy for
385 the technical skill gaps that exist among recent college graduates.

Employers seem to be favorable to individuals with alternative credentials. In many cases, employer-perceived skill gaps are not statistically different when comparing recent college graduates with ACNG candidates. Social preference for the college degree may be better explained by public ignorance about appropriate
390 alternative programs, a lack of appropriate programs for certain occupations, and government financial and other policy which gives preference to accredited education.

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