

Rethinking the Skills Gap, *Roland Rathelot and Thijs van Rens*

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Teaser:

Better understanding of skills mismatch is essential to finding effective policy options.

Elevator pitch:

Evidence suggests that productivity would be much higher and unemployment would be lower if the supply of and demand for skills were better matched. As a result, skills mismatch between workers (supply) and jobs (demand) commands the ongoing attention of policymakers in many countries. Policies intended to address the persistence of mismatch focus the supply side of the issue by emphasizing worker education and training. The role of the demand side, employers' wage-setting practices, garners comparatively little policy attention.

Pros:

- Analysis shows that 4% of workers are under-skilled, and 10% are over-skilled for their jobs.
- Mismatch is an important determinant of productivity and wages.
- Joblessness in an economic downturn would also affect one-third fewer people if mismatch were not a problem.
- Policies intended to address mismatch focus on education and training to boost worker skills, even though the phenomenon may also persist because of firms' actions – among them, a failure to adjust wages to reflect relative skill shortages.

Cons:

- Nearly nine in 10 workers have skills that correspond to the requirements of their jobs.
- Though reliable estimates show that skills mismatch negatively affects *individual* workers' productivity, estimates about the effect of mismatch on *aggregate* productivity remain largely speculative.
- Estimates of the effect of skills mismatch on unemployment suffer from serious measurement issues. Among other problems, no study fully accounts for the interconnections among occupations.
- Evidence from the United States shows that mismatch across geographic areas has a negligible effect on unemployment and productivity – that is, workers and jobs are, overall, well distributed over the territory.

Authors' main message:

Skills mismatch has large effects on productivity and unemployment, and therefore is – and should be – an important concern to economic policymakers. Almost all proposed policy interventions put forward reforms of education and training as solutions to address a perceived shortage of skills, while little addresses wage setting. This is problematic because such reforms, which are often expensive, will be ineffective if wages do not reflect relative shortages of skills. If mismatch instead reflects an unresponsiveness of wages, then workers will “sell” their skills where they fetch the best price rather than where they are most needed.

Motivation

The idea that the labor market suffers from severe imbalances in terms of skills offered by workers and those required by employers is a pervasive one. Skills mismatch is viewed as a structural issue – that is, an issue that is present whether the economy is in good shape or in crisis. However, its salience re-emerges during recessions. During the recent Great Recession, for example, questions arose about whether increased mismatch was the reason that unemployment remained at high levels long after the initial, precipitating events. In government circles, the issue is perceived as independent of business cycles, some sectors complaining about the trouble they experience to find workers, while unemployment rates remain stubbornly high.

Three possible reasons would explain why a skills gap persists: workers do not adjust to changes in skills demand by acquiring the skills they need to find a job; firms do not adjust to changes in skills supply by creating jobs that utilize the skills available in the labor market; or, wages do not reflect the skills shortages by creating incentives for workers to acquire scarce skills, or to abandon other occupations.

An important component of the European Union’s strategic framework for education policy aims “to better identify and manage the availability of required skills, competences, and qualifications, and to help prevent skills gaps and mismatches.”¹ European countries and other advanced economies worry about the “growing gulf between the skills workers possess today and the skills businesses say they need,” as stressed in the 2014 report of the European Intelligence Unit. Concerns often focus on shortages of workers with skills in the science, technology, engineering and maths (STEM) subjects, but increasingly concerns also extend to “soft skills,” such as communications, team work, and problem solving.

At the same time, many academic economists remain unconvinced of the existence of a skills gap. Accustomed to the idea of the “invisible hand” equating supply and demand, they are naturally skeptical about the idea that large segments of the labor market would persistently be in disequilibrium; that is, they find it hard to believe that employers would not be able to convince workers who have, or could acquire, needed skills to work for them, perhaps in return for a higher wage than they might earn elsewhere.

¹ http://ec.europa.eu/education/policy/strategic-framework/skills-development_en

Discussion of pros and cons

Recently, researchers have begun to examine these issues in greater detail and in new ways. The availability of large datasets containing information about workers and firms has made it possible to gauge the effect of skills mismatch on workers' productivity and on aggregate unemployment. The literature has also started to explore the causes of mismatch, suggesting what policies may and may not be effective in addressing the issue.

Workers and jobs: Skills mismatch and productivity

The most immediate problem associated with mismatch concerns its effect on productivity. The literature studying this effect looks at existing matches of workers and jobs and tries to determine the extent to which workers have adequate skills to perform their jobs. This issue has two sides. Over- or under-qualification (also called vertical mismatch) occurs when workers have the right types of skills, but too much or not enough. Think of, in the over-skilled category, a linguist teaching Spanish class, or, in the under-skilled one, a mechanic working as an engineer. Horizontal mismatch (also called field-of-study mismatch) occurs when workers do not have the type of skills required by the job, but they have other skills at a similar level – such as a biology teacher taking care of physics classes in a school that does not manage to recruit a physics teacher.

The early literature on mismatch used self-reported data generated from workers' responses to questions about whether they felt under- or over-qualified for the job. An obvious limitation of the approach is that it captures workers' under- or over-confidence at the same time as mismatch. Beginning in 2011, the OECD's Programme for the International Assessment of Adult Competencies (PIAAC) began its Survey of Adult Skills, an assessment designed to provide representative data on workers' skills. The survey provided an innovation by providing data involving skill proficiencies that are assessed, rather than self-declared. Thus far, the data include measurements of skills of 250,000 individuals in 33 countries.

Fichen and Pellizzari (forthcoming) introduce the measure that has been officially adopted by OECD since 2013. Their approach combines declarative measures of qualification in the job with objective measures of skill proficiency. For each occupation and country, the authors define a range of skill proficiencies based on the minimum and maximum proficiencies of workers who have defined themselves as being well-matched to their work positions. A worker is defined as over-skilled if her skill proficiency is higher than this maximum. Under-skilled status is similarly defined. They find that 86% of workers are well-matched, 4% are under-skilled, and 10% are over-skilled. Still using PIAAC data, Adalet-McGowan and Andrews (2015) show that there is a negative correlation between this measure of mismatch and labor productivity at the industry level: industries where there are more under-qualified or over-skilled workers exhibit lower levels of labor productivity.

Two papers provide evidence beyond developed countries. Leuven and Oosterbeek (2011) provide a review of worldwide mismatch research. They find that over-education affects 30% of workers, and under-education affects 26% of workers, with some variation

surfacing across continents. Handel et al. (2016) apply the OECD method to measure mismatch in developing countries, using the World Bank's STEP Household Skills Survey. They find that over-education is the most prevalent form of mismatch.

A framework by Guvenen et al. (2015) analyzes worker-occupation matches. If a worker does not possess the abilities that are necessary to learn the skills required by an occupation, then she is "mismatched." Estimating a structural model on U.S. data, they find that being mismatched early in one's career harms a worker's wages in a large and persistent manner.

Quantifying the effect of the *overall level* of labor market mismatch is much more difficult than measuring the effect of being mismatched on the productivity of *individual workers*. For example, given the skills workers have and the skills jobs require at some point in time, how much would production increase if it were possible to reallocate workers to jobs so that production could be maximized? Answering this question convincingly is difficult because it requires assumptions about the specific forms of the production technologies. Considering both labor and physical capital, Hsieh and Klenow (2009) show very large effects from their misallocation across firms, responsible for a productivity gap of 40-60% between India and the United States, and 30-50% between China and the United States.

Job seekers and vacancies: Skills mismatch and unemployment

If the skills that firms require, and the skills that workers possess are sufficiently far apart, then at least some workers will not be hired. Therefore, skills mismatch generates not only a productivity loss, but unemployment as well. Unemployment carries with it huge economic and personal costs. Therefore, understanding the effect of skills mismatch on unemployment is important for crafting effective policy.

Labor market mismatch generates unemployment if the unemployed workers looking for jobs, and firms with vacant positions looking for workers cannot form a match because the worker and vacancies are "not right" for each other. This idea can be formalized by modeling the labor market as divided in segments, with workers (and vacancies) unable to move from one labor market segment to another. If there are deviations between the distributions of workers and jobs among the various segments of the labor market, then some workers will remain unemployed, while, at the same time, some firms will not be able to fill all positions.

If unemployment is caused by mismatch, then there is a tight link between the dispersion in labor market conditions across labor market segments and the aggregate unemployment rate. The idea is that, if there are jobs in occupations with certain skill sets while unemployed workers are available with different skill sets, then we should see large differences in the ratio of vacancies over unemployment across occupations with different skill requirements. This prediction allows empirical researchers to quantify the aggregate effect of mismatch on unemployment. The challenge is to measure how much lower the unemployment rate would be if – hypothetically, of course – we were able to reallocate unemployed workers to those occupations where they are most likely to find jobs.

Despite severe measurement issues (see section on limitations and gaps below), there is a remarkable amount of consensus in the literature about some basic facts about unemployment due to labor market mismatch. We summarize the consensus in the literature around three main findings:

Geographic mismatch is negligibly small. Şahin et al. (2014) find that mismatch across U.S. counties and Metropolitan Statistical Areas contributed less than half a percentage point to unemployment - and that this contribution did not go up notably in the Great Recession. They thus conclude that “geographic mismatch plays no apparent role.” This finding is confirmed by Marinescu and Rathelot (2014), who account for the fact that workers are not stuck in their counties or states; they look for jobs not only where they live but also in surrounding areas. The finding is also consistent with evidence showing that geographic (inter-state) mobility did not decrease during the Great Recession. Therefore, mismatch unlikely to have contributed much to the very large increase in unemployment during this recession.

Skills mismatch, as measured by mismatch across occupations or industries, is an important contributor to unemployment. Using Standard Occupational Codes to categorize the nature of certain kinds of work, Şahin et al. (2014) find that increased mismatch across three-digit occupations accounted for around 1.5%-points (or about one-third) of the increase in unemployment in United States in the Great Recession. Research by Lazear and Spletzer (2012) shows similar results for the United States; Patterson et al. (2016) also reach similar conclusions in research analyzing the situation for the United Kingdom.

Skills mismatch is countercyclical; that is, it is not a new phenomenon, but a recurrent problem that surfaces in recessions. Herz and Van Rens (2014) show that U.S. unemployment due to mismatch across industries over the 1979-2010 period evolves over time in a very similar fashion to the overall unemployment rate. The results in Şahin et al. (2014) show a very similar pattern over the 2001-2012 sample. These results were surprising to many, because early commentators attributed the rise in mismatch unemployment to a structural change in the labor market (Kocherlakota 2010).

The causes of skills mismatch and how to address them

Unexpected events or phenomena may affect occupations in different ways. For instance, Autor, Levy and Murnane (2003) illustrate that the emergence of computers and information technologies reduced the demand for routine jobs, which were to some degree made obsolete by computers, and increased the demand for non-routine jobs, which proved to be relatively complementary to the computer. For example, bank tellers and bank managers likely have been affected in opposite ways by the IT revolution in general, and by ATMs, in particular. Recessions change the relative demand of different goods, too. Sectoral shocks may translate into asymmetric occupational shocks: for example, bakers may be less sensitive to recessions, and restaurant waiters may be more sensitive. As a result, the fact that mismatch exists is not in and of itself surprising. The relevant question is: why does it seem to be so persistent?

Workers who work (or look for a job) in an occupation where the number of workers exceeds the number of positions have ways to adjust. They can apply to other higher-demand occupations that require similar skills, or they can acquire new skills through training. Alternatively, employers could also adjust to workforce shortages, by changing the skill content of occupations, or by training up workers from similar occupations to fit new skill requirements. Some recent empirical literature documents that employers adjust the task and skill content of jobs (for the same occupation) with the business cycle, “upskilling” when workers are more abundant in a recession (e.g. Hershbein and Kahn, 2016).

Adjustment, whether by workers or employers, may be difficult and costly to achieve in the short run, especially when confronting large skill differences between origin and target occupations. Most policy interventions are based on the implicit assumption that this is the reason for the skills gap. The European Commission, for instance, believes that “Europe needs a radical rethink on how education and training systems can deliver the skills needed by the labor market.”² As a result, it set up the Rethinking Education initiative “to reform education systems across the EU so as to meet growing demand for higher skills levels and reduce unemployment.”³

Herz and van Rens (2014) show that it is possible, using data on wages and profits across industries in addition to the data on job-finding rates, to quantify how much of mismatch unemployment stems from lack of adjustment by workers – or from a lack of adjustment by firms. On the workers’ side, they find this scenario: There are industries where workers have a hard time finding jobs, but where they earn high wages if they do; and there are other industries where jobs are plentiful, but wages are low. This is what we would expect to see if workers operate along a “no arbitrage” condition. That is, they can move between industries, but will only do so if they are given the right incentives. If, on the other hand, there are many industries where both job-finding rates and wages are high, and others where both are low, the logical conclusion would be that mismatch persists because workers lack the skills required to move into better jobs. Using data for the United States over the 1979-2010 period, Herz and van Rens find that mismatch cannot be fully explained by barriers to workers and firms in adjusting to changing skill demand and supply, respectively.

² http://europa.eu/rapid/press-release_IP-12-1233_en.htm

³ http://ec.europa.eu/education/policy/multilingualism/rethinking-education_en

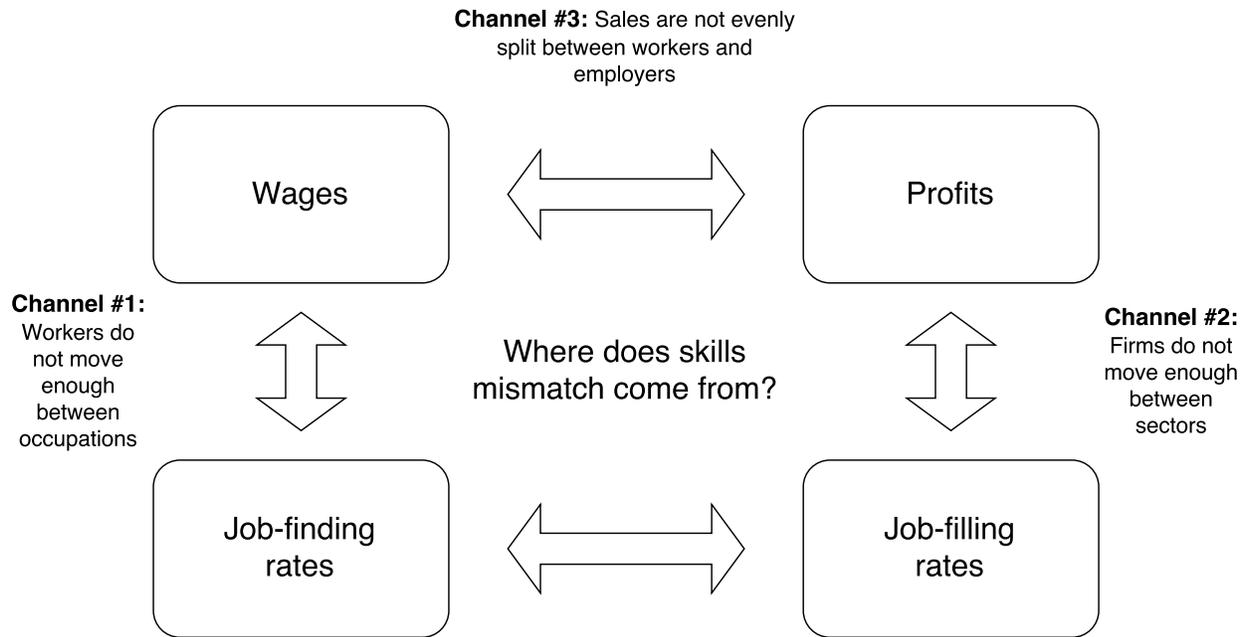


Figure 1: A decomposition of the potential sources of skills mismatch

Source: Adapted from Herz and van Rens (2014)

This then raises the question: If workers adjust to changes in skill demand, and employers adjust to changes in the supply, how can mismatch persist? The answer lies in the way wages are set. If wages reflect the relative abundance or the relative shortage of skilled workers, then workers' and employers' capacity to adjust would lead to the elimination of mismatch. However, jobs in industries that generate high profits (retail trade, educational services, mining and forestry) tend to pay low wages, and are therefore unattractive to workers, while jobs in industries that pay higher wages (finance, computer and electronics manufacturing, paper and printing) are not profitable to firms.

Other researchers, based on very different approaches, have emphasized the role of wage setting as well. Among the forces that they suggest may be at work are: automated screening systems that rule out potential candidates who might have surfaced in subjective, human resources screening; and a preference for hiring experienced candidates over investing in training for inexperienced but promising candidates (Cappelli 2012). If workers do not move into low-unemployment occupations, the problem may not be that they cannot train or adjust, but that wages are too low to attract them. In the UK, for instance, less than half of STEM graduates work in scientific occupations, and there is no wage premium in other occupations for having a STEM degree (Chevalier 2015). Firms, on the other hand, are more interested in hiring workers with STEM skills because these workers are not only very productive but also relatively cheap – despite widespread public

perception that STEM graduates earn high salaries. Thus, companies open lots of vacancies for STEM positions, but then find it very difficult to fill them.⁴

Limitations and gaps

While the literature has progressed in the last decade to provide measures of the extent of mismatch and how it affects unemployment and productivity, many issues regarding measurement still raise concerns. Potentially, these issues are large enough to affect the qualitative conclusions we draw here.

The concept of skills is multi-dimensional, including the amount and quality of education, field of study, experience in current and previous jobs. There are many sorts of skills: technical skills, cognitive skills, soft skills such as communications, problem solving, the ability to work well in teams, and perhaps even having certain personality traits. The extent to which skills are transferable varies. Some skills are general; others are entirely job specific. An ideal dataset would detail the precise set of skill requirements of the job, as well as the precise skill set of the worker. However, most data sources, particularly the ones that cover a representative sample of the labor force, rely on only responses to a few questions that generate a description of a worker's skill set. A related issue stems from the difficulty in gathering symmetric information about workers' skill sets and jobs' skill requirements.⁵

The studies we describe differ in the ways they address this measurement issue. As a result, it is difficult to pinpoint a "consensus estimate" of the effect of skills mismatch on productivity. The measurement problem is more difficult if we try to estimate not the effect of being mismatched on an individual worker's productivity, but the effect of overall mismatch on aggregate productivity. Moreover, a substantial amount of uncertainty shrouds the few results we discuss here. In the literature on mismatch and unemployment, the measurement issue takes a different form.

In this literature, researchers think of the labor market as being segmented into submarkets, and the primary difficulty with measuring the effects of mismatch lies in finding the correct partitioning of the labor market. Ideally, the partition needs to satisfy two properties. First, submarkets must be closed. No job seeker should end up finding a vacancy in a different submarket than her own. This means that the empirical definition of a submarket should be coarse enough to accommodate some degree of labor-market mobility. The second property is homogeneity. Two job seekers (or two vacancies) should be close enough that they can be considered identical by employers (or workers). This means that the definition of segments should be precise enough. If the partition is too

⁴ <https://www.brookings.edu/interactives/still-searching-job-vacancies-and-stem-skills/>

⁵ The text of the job ad can be parsed, and skill requirements can be extracted. The analytics company, Burning Glass Technologies, for example, culls information on ads from several thousands of American online job boards. Each job listing corresponds with a set of phrases representing features of the job, including skill requirements. See Deming and Kahn (2017) for an example.

coarse, the measure of mismatch will underestimate the true phenomenon. If submarkets are too small, mismatch may be overestimated. What is the right partition of the labor market? This question first arose in the literature on geographic mismatch, where the submarkets are geographic areas.

ACCOUNTING FOR MOBILITY ACROSS GEOGRAPHIC AREAS: To measure the impact of local fiscal policy and transport improvements on neighboring areas, Manning and Petrongolo (2017) assess “how local” a given labor market is. How far will a local stimulus propagate thanks to worker mobility? They estimate the extent to which job seekers tend to apply for further-away jobs by combining data on flows into and out of unemployment in England and Wales at the Census-ward level, and a structural model of job searching and matching. Overall, they find little mobility and modest ripple effects of local policies.

Marinescu and Rathelot (2016) apply a similar approach to measure mismatch in the United States. Instead of specifying ad hoc the level of analysis (states might be too coarse and ZIP Codes too fine), they allow job seekers to apply everywhere, with different probabilities. They use data from the website CareerBuilder.com to observe the locations (at the ZIP Code level) of job seekers, vacancies and applications. They estimate a measure of “distaste for distance,” which captures the reluctance of job seekers to apply for far-away vacancies. They inject this parameter in a model à la Manning and Petrongolo (2017), in which job seekers decide to apply somewhere based on two criteria: (i) the distance to the vacancy, and (ii) how many job seekers compete for a given vacancy. They find that 10 more miles decreases the probability of applying for a job by around 35%. They use this information to predict how many matches will result from a given allocation of job seekers across ZIP Codes, and they draw conclusions about the gulf between a “perfect” situation – one that would maximize the number of hires – and the situation that occurs when taking into account these geographical and competitive realities. They find the share of unemployment that is due geographic mismatch is around 5% only. Reassigning workers over space to maximize the number of hires would only increase those by 5%.

ACCOUNTING FOR MOBILITY ACROSS SKILLS: Measuring skills mismatch suffers from the same dilemma as geographic mismatch in terms of the difficulty in defining sensible submarkets. Even the Standard Occupational Codes that categorize the nature of certain kinds of work are imperfect vessels for this analysis. For instance, one might surmise that workers in the same four-digit occupation codes can move to and from the same positions with relative ease. But, for instance, would this mean that former post-secondary English teachers can fill post-secondary physics teachers’ positions? By contrast, workers slotted into different two-digit occupation codes might seem “unmovable.” Yet, could tax examiners and tax examiners logically segue to and from positions in their differently coded jobs? Ideally, a measure of skills mismatch should account for the fact that job seekers move, to some extent, across occupations.

The U.S. Department of Labor has sponsored the creation of an impressive database, O*NET OnLine, which provides a detailed mapping between six-digit SOC occupations and the usual tasks and skill requirements associated to jobs in this occupation. Autor and Handel (2013) motivate this effort by showing that tasks explain a substantial amount of wage heterogeneity within occupations. However, research using these data is just beginning.

Most studies on mismatch use three- or four-digit occupations, or even industries, to operationalize the concept of a skill-based labor market segment (Lazear and Spletzer, 2012, Sahin et al. 2014, Herz and van Rens 2014). The results of these studies are likely to be revisited, and may very well be overthrown as researchers find ways to use the opportunities in the O*Net data.

DATA SOURCES. A separate issue regards what data to use. In the United States, job seekers are counted using the Current Population Survey (CPS), which assumes that the industry and the occupation of a job seeker are both the same from one job to another. For vacancies, two U.S. data sources are available. The Job Openings and Labor Turnover Survey (JOLTS), which serves as a source of demand-side indicators of labor shortages at the national level, and allows researchers to compute the number of vacancies by industries (at a two-digit level). The Conference Board's Help Wanted OnLine (HWOL) is made of the universe of unique online job vacancies in the United States; these are collated into counts of vacancies by occupations. These two sources (JOLTS/HWOL) provide the most straightforward measure of the vacancy-unemployment ratio across industries and occupations. Many papers use these sources (Lazear and Spletzer 2012, Sahin et al. 2014) at the price of working on a relatively short time window. Other papers make assumptions on the matching technology to back out the vacancy-unemployment ratio from the job-finding probabilities measured in the CPS. This technique results in a much longer time series (Herz and van Rens 2014). Luckily, findings look very similar in research relying on these different sources, lending additional credibility to the results.

CAUSES: When it comes to the causes of mismatch, the evidence is very thin indeed. We expect the thinking on this issue will progress substantially as further research sheds light onto the mechanisms and trade-offs behind wage determination.

Summary and policy advice

Skills mismatch is an important cause of productivity loss and unemployment. Thus, policymaking tools that diminish its presence and persistence hold potential to benefit economies, firms, and people who are unemployed and underemployed.

However, in the context of a proposed “radical rethink on how education and training systems can deliver the skills needed by the labor market,”⁶ reform of education and training systems may be neither needed nor desired. The most striking conclusion from current research is that worker mobility frictions may not be the main contributor to labor market mismatch. Yet, almost all proposed solutions to the skills gap treat the phenomenon as a problem of the education system. Such interventions in education and training are likely to be expensive, and, at the same time, may not be as effective as expected.

Why would increasing the emphasis on “scarce” skills in schools and universities fail to guarantee that skills mismatch will be reduced? The reason is simply that students choose, first, what skills to acquire in school and university, and, then, whether and how to use these skills in the labor market. If wages do not reward certain skills, students will either

⁶ EC press release IP/12/1233

choose either not to pursue such skills, or, pursue the skills but seek employment in other, higher-paying occupations. The STEM “gap” offers the most obvious example of this problem. While firms complain about a shortage of qualified physicists and engineers on the labor market, a very large number of graduates in these fields work in the financial sector, where they use only a subset of their STEM skills, but earn more money than they would in the shortage niches. Encouraging universities to educate more physicists and engineers will not solve the mismatch problem if these graduates look for – and find – better-paying jobs in investment banks.

These questions underline the need for additional research to understand the forces that foster and perpetuate mismatch. While the message from the current research should not be interpreted as a call to “do nothing,” greater knowledge is needed to guide policymakers in devising effective solutions.

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