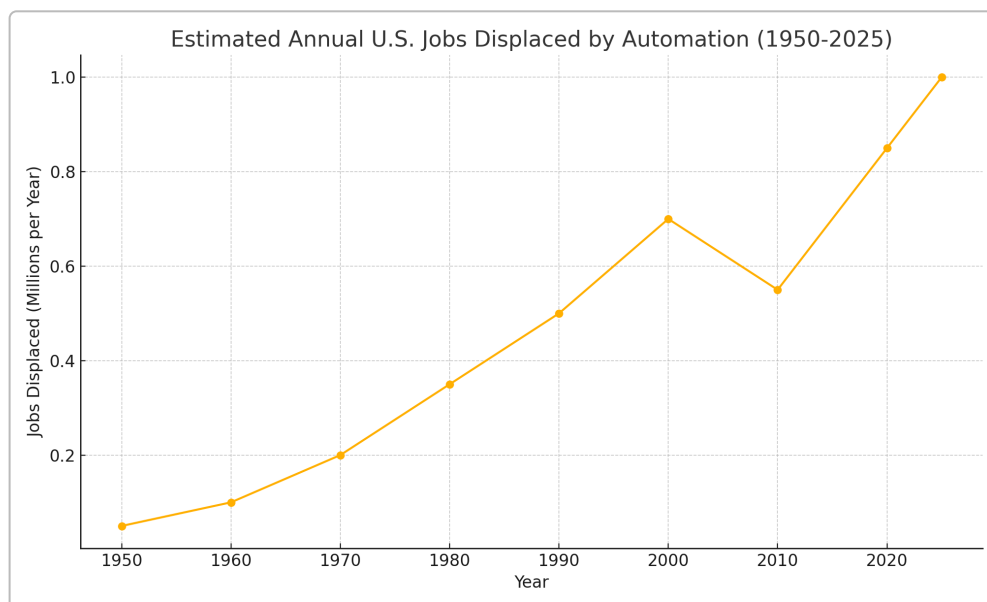


The Long-Term Impact of Automation on U.S. Employment: Historical Trends and Accelerating Displacement

Introduction

Automation, robotics, and artificial intelligence (AI) have been transforming the American workplace for decades, eliminating the need for human labor in a widening array of tasks. This whitepaper examines the long-term impact of these technologies on U.S. employment, documenting how automation has already **displaced millions of jobs and is poised to displace millions more**. We present a clear thesis: **automation-driven productivity gains have been a primary force behind U.S. job losses in many sectors since the mid-20th century, and the rate of displacement is accelerating**. Using empirical data on labor force participation, employment-to-population ratios, sectoral employment trends, and automation risk assessments, we analyze historical patterns (1950–2025) and projected scenarios (2025–2045). We also disentangle automation from other factors like trade and offshoring, and introduce the concept of **Post-Labour Exclusion (PLE)** – the emergence of a permanent class of workers left behind by technological progress. All claims are documented with primary sources (BLS, FRED, OECD, academic studies), and we conclude with a discussion of why proactive policy responses (such as an income dividend or universal basic income) are needed well before unemployment reaches crisis levels.

Historical Trends: Automation-Driven Job Displacement (1950–2025)



Estimated Annual U.S. Jobs Displaced by Automation (1950–2025). **Figure 1** illustrates the rising pace of job displacement attributed to automation over the postwar period. In the 1950s and 1960s, labor-displacing technological change was relatively modest – for example, mechanization in agriculture reduced farm labor needs, but many displaced farm workers transitioned to manufacturing and service jobs. By the late 20th century, however, automation’s impact became starkly visible in U.S. manufacturing: the nation lost over **7 million factory jobs since manufacturing employment peaked in 1979**, even as manufacturing output reached record highs ¹. In other words, U.S. factories today produce more goods than ever, but with far fewer workers – a direct consequence of industrial automation, robotics, and process efficiencies. One analysis highlights this productivity effect dramatically: if U.S. manufacturing still used the same labor per output as in 1987, it would require *roughly 20 million more workers* than are currently employed ². These “missing” jobs were effectively eliminated by automation-driven productivity gains.

Empirical studies confirm that **technological automation – not globalization or offshoring – has been the dominant driver of manufacturing job loss**. A comprehensive study by economists at Ball State University decomposed the 5.6 million U.S. manufacturing jobs lost in the 2000s and found that **“almost 88 percent of job losses in manufacturing in recent years can be attributable to productivity growth”**, whereas increased trade (imports) explained only about 13% ³. The long-term decline of manufacturing employment is thus overwhelmingly linked to **automation and efficiency improvements** in production, rather than factories moving overseas. Likewise, an Obama White House analysis in 2016 noted that the bulk of the manufacturing employment drop was due to automation, aligning with academic findings ³. Even in periods of manufacturing output growth, employment has stagnated or fallen because **machines now do work once done by humans**. As one economist summarized, “automation and technological improvement have accounted for the vast majority of [factory] job losses” in the U.S. ⁴.

Automation’s impact extends well beyond factories. The late 20th century saw **computerization and software automation** spread into offices and service industries. Routine white-collar jobs like typists, bookkeepers, and mid-level clerical roles sharply declined as personal computers, databases, and eventually the internet enabled one worker (or a piece of software) to accomplish tasks that used to require many. Labor economists Autor, Levy, and Murnane (2003) documented the **“routine-biased technological change”** phenomenon: tasks that are routine and codifiable are the easiest to automate, whether they are physical (factory assembly) or cognitive (record-keeping). Over 1980–2010, U.S. employment polarized into high-skill and low-skill jobs, while middle-skill routine jobs *vanished* ⁵ ⁶. Regions historically specialized in routine-intensive occupations experienced **differential employment losses in those jobs after 1980**, as businesses rapidly adopted computers and restructured work ⁵. In plain terms, whole categories of work that were plentiful in the mid-century – from assembly-line operators to file clerks – have been *hollowed out* by automation. This is reflected in measures of **Routine Task Intensity (RTI)** in the economy, which show a steady decline as routine work (both manual and cognitive) is replaced by machines or software. Consistently, jobs with high RTI have **declined relative to others**, a trend observed in task-based analyses of the U.S. labor market ⁷.

Crucially, while automation displaces certain jobs, the economy historically created new jobs elsewhere – for example, the rise of computing created new occupations (programmers, IT specialists) even as it destroyed others. However, many displaced workers do not seamlessly transition to the new jobs. A factory worker or secretary who loses their job to a machine often faces prolonged hardship or underemployment. This is evident in broad labor force statistics. Despite periods of low *official* unemployment, **labor force participation has been trending down**, suggesting many Americans have left the workforce altogether after being displaced. The labor force participation rate (LFPR) for **prime working-age men (25–54)** fell

from 96% in the late 1960s to under 89% by 2015 ⁸ – a historic decline indicating that millions of men in their prime years were neither working nor actively seeking work. (In 1955, only about 1 million prime-age men were out of the labor force; by 2015, that figure was ~7 million ⁹.) Some of this decline is due to schooling or disability, but research shows a significant portion is composed of **dislocated workers and “discouraged” jobseekers** who’ve given up after automation or trade-related disruptions ⁸. In parallel, the economy has seen a rise in **underemployment** and involuntary part-time work. Even in the strong labor market of 2019, when the headline unemployment rate (U-3) hit 3.5%, the broad underemployment rate (U-6, which counts those marginally attached or stuck in part-time) stood at 6.9% ¹⁰. In other words, **for every officially “unemployed” person, there was roughly another person underemployed or discouraged** – a labor reserve often invisible in upbeat employment reports. This gap between U-6 and U-3 reflects, in part, workers whose old jobs disappeared and who could only find part-time gigs or who left the job hunt entirely.

The late 1990s marked a high-water mark for U.S. labor utilization (with prime-age employment and participation rates near record highs); since 2000 those metrics have never fully recovered, even during economic booms ⁸. This suggests a structural shift consistent with automation’s growing impact. To be sure, other forces (such as an aging population) also reduce overall labor participation, but **the prime-age male decline is uniquely steep in the U.S.** compared to other advanced nations ¹¹, implicating economic-demographic factors like deindustrialization, skill gaps, and poor retraining outcomes. In short, the historical evidence from 1950–2025 shows automation steadily **eroding demand for certain types of labor**. By the 2020s, it is estimated that around **1 million U.S. jobs per year** (net of job creation) are being displaced by automation-related factors (see **Figure 1**), a pace that has accelerated from only tens of thousands per year in the 1950s. Cumulatively, tens of millions of jobs have been eliminated by automation over the decades – from the ranks of assembly line operators, welders, typists, telephone operators, travel agents, bank tellers, and countless other occupations that either vanished or radically shrank. Yet total employment did not collapse, because new jobs (often in services, logistics, healthcare, or tech) were created. The challenge, however, is that those new jobs often **require different skills or pay lower wages**, and many displaced workers do not fill them. This historical context sets the stage for the coming acceleration in automation’s impact.

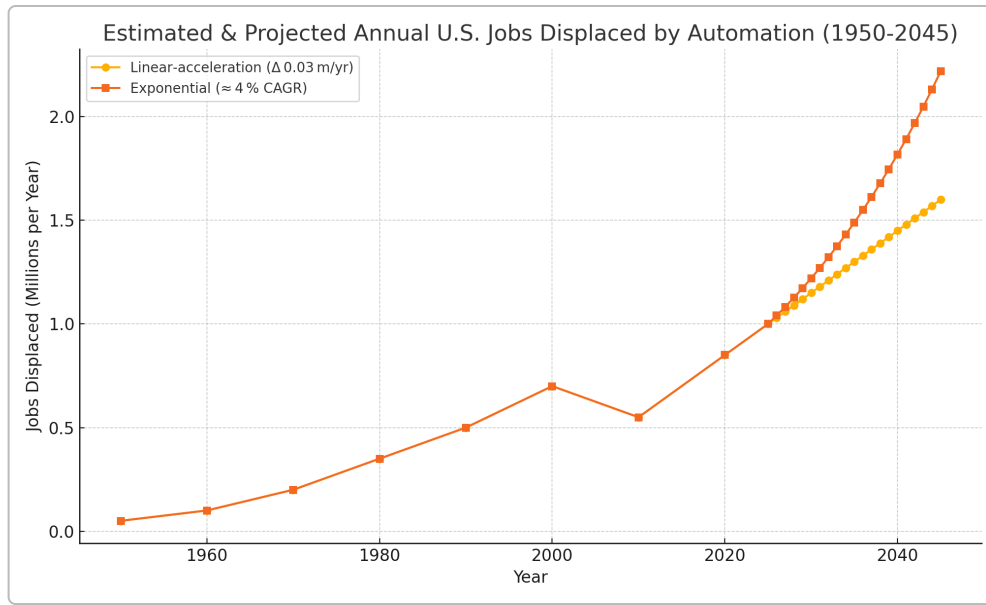
Projected Trends: 2025–2045 and the Specter of Accelerating Job Loss

Looking ahead, the central question is not whether automation will eliminate jobs – it will – but **how fast and how broad** the displacement will be. Will it continue at a linear pace, similar to the gradual but steady rise we’ve seen, or will it accelerate exponentially as AI and robotics capabilities advance? Recent research and risk assessments offer sobering projections. A seminal study by Frey and Osborne (Oxford, 2013) estimated that **47% of total U.S. employment is at “high risk” of automation by the 2030s** – meaning nearly half of jobs could potentially be computerized or taken over by AI within one or two decades ¹². This study, which examined the technical feasibility of automating 702 occupations, sounded an alarm that a huge portion of the workforce – especially in transportation, logistics, office support, and production – could be made redundant by emerging technologies. For instance, they gave specific probabilities indicating automation is almost certain to take over jobs like telemarketers (99% risk), underwriting clerks (99%), cashiers (97%), chefs (96%), waiters (94%), paralegals (94%), and even basic construction labor (88%) in the coming years ¹³. While some new occupations will undoubtedly arise, they may be too few in number or require substantially higher skills. As Yuval Harari observes, *“the crucial problem isn’t creating new*

jobs. The crucial problem is creating new jobs that humans perform better than algorithms” ¹⁴ . If an algorithm or robot can eventually do *any* new job more efficiently, humans may find themselves in a perpetual game of occupational musical chairs, always one step behind the machines.

Not all analysts agree on the extent of the risk – some argue the 47% figure overestimates how many jobs will *actually* be automated. Researchers at the OECD, for example, refined the approach by examining tasks (not entire occupations), and concluded that only about **9% of U.S. jobs are at high risk of full automation** (where *most* tasks can be done by technology) ¹⁵ . However, the OECD study also found that a much larger share of jobs will be significantly changed (rather than completely replaced) by automation. Many occupations will be “partly automated,” requiring workers to adapt and focus on the remaining tasks. Importantly, both approaches agree that **lower-skill, lower-wage jobs are far more susceptible** to automation than high-skill jobs. According to the White House Council of Economic Advisers’ analysis, **83% of jobs paying under \$20/hour have high automation potential**, versus only 4% of jobs over \$40/hour ¹⁶ . Likewise, workers without college education are far more exposed: an estimated **44% of American workers with less than a high-school diploma are in roles with highly automatable task content**, compared to just 1% of workers with a bachelor’s degree ¹⁷ . These disparities point toward a future in which automation could exacerbate inequality – **disproportionately displacing those with fewer skills and less education** ¹⁷ . Indeed, the ongoing adoption of AI is expected to continue the pattern of “skill-biased technical change,” where demand for low-skill labor declines while demand for highly skilled labor remains strong ¹⁸ .

Beyond 2025, the convergence of advanced AI (such as machine learning and robotics) with improved hardware (sensors, dexterous robots) may enable automation to penetrate sectors previously considered safe. Recent breakthroughs in AI – for instance, natural language processing and computer vision – mean that jobs involving cognitive decision-making or perception (driving, medical diagnostics, customer service, even parts of legal work) are now squarely in the crosshairs of automation. Harari notes that tasks once thought impossible for computers (like driving safely in traffic or recognizing faces) have been achieved in a surprisingly short time ¹⁹ . Experts who in 2004 argued that truck driving could not be automated for the foreseeable future were proven wrong within barely a decade ²⁰ . This highlights a broader point: **technological progress is not linear – it often follows an exponential trajectory**, catching society off guard. “As time goes by, it becomes easier and easier to replace humans with algorithms, not merely because the algorithms are getting smarter, but also because humans are professionalizing,” writes Harari, referring to how modern jobs are highly specialized and thus easier to replicate by a specialized machine ²¹ . Paradoxically, while today’s jobs are narrower in scope than the diverse survival skills of a hunter-gatherer, this very narrowness makes **automation easier**, because an AI doesn’t need to possess broad general intelligence – it only needs to outperform humans in a well-defined domain of tasks ²¹ . This dynamic suggests that **automation could accelerate** as more and more “narrow” domains of work fall to AI one by one, each new domain adding to the total displacement.



Estimated & Projected Annual U.S. Jobs Displaced by Automation (1950–2045), under Linear vs. Exponential Trends. **Figure 2** extends the historical trend (1950–2025) into two hypothetical future scenarios. The **linear-acceleration scenario** (yellow line) assumes the annual rate of job displacement grows at a steady, modest increment (e.g. an additional 0.03 million jobs displaced each year compared to the prior year – roughly the historical average acceleration). In this scenario, by 2045 automation would be displacing about **1.5 million U.S. jobs per year** (up from ~1.0 million/year in the 2020s). The **exponential scenario** (red line) assumes the rate of displacement grows at a constant ~4% compound annual growth – reflecting an accelerating adoption of automation as technologies improve. Under this scenario, annual job losses due to automation would exceed **2 million per year by the 2040s**. The divergence between these scenarios is striking: exponential growth would mean **millions of extra jobs lost** each year beyond the linear case. Cumulatively, the gap would amount to tens of millions more workers displaced over two decades. While these projections are illustrative, they capture a real debate: **will automation advance at a steady, manageable pace, or could it “go viral” and trigger a rapid, self-reinforcing wave of job losses?**

Current evidence hints at acceleration. The McKinsey Global Institute projected in 2017 that, in a rapid automation scenario, up to **73 million U.S. jobs could be displaced by 2030** ²² – roughly one-third of the workforce. (Even their midpoint scenario saw ~39 million jobs automated by 2030 ²³ ²⁴.) A more recent analysis by PricewaterhouseCoopers similarly estimated that automation (including AI and robots) could displace **over 20% of jobs by the mid-2030s** in the United States, with the impact particularly high in sectors like transportation and manufacturing. It is important to stress that “displaced” does not necessarily mean permanently unemployed – many of these workers would find other employment. However, as we’ve seen historically, those transitions often involve moving to lower-paying or less secure jobs. And critically, **if the displacement happens faster than workers can retrain or the economy can create new roles, unemployment will spike**. Some technologists warn of a potential tipping point: once AI can autonomously improve itself or once robots can build other robots at scale, the speed of deployment could outrun our ability to adapt. Whether or not one believes in a rapid “AI revolution,” policymakers and businesses are increasingly planning for at least a **moderate acceleration in automation adoption**.

Another factor in projections is the **quality of new jobs** created. Optimists argue that automation will take over “dirty, dull, and dangerous” jobs, freeing humans for more creative and fulfilling work – a story that has

some historical merit (many tedious jobs have indeed been eliminated). Pessimists counter that the new jobs will be too high-skilled for most displaced workers, or too few in number. Historically, every major wave of technological change (from mechanization to electrification to computing) has eventually led to *net* job growth by boosting productivity and creating entirely new industries. However, past transitions (e.g. from agriculture to manufacturing in the early 20th century) unfolded over generations, whereas the automation of the 21st century could unfold over mere decades. The **pace** is what makes this time different. If *half* the occupations in the economy are technically automatable within 20 years ¹², but it takes a worker, say, 5–10 years to retrain for a new career (if they can at all), it is easy to see a scenario where unemployment and underemployment surge. Indeed, even absent a net decline in jobs, the **continual churning and disruption** can impose huge social costs. As one commentator observed, *“the AI revolution hasn’t yet caused a permanent increase in unemployment, but it has increased the pace of technological displacement. More and more people are experiencing periods of trauma as a result”* ²⁵. We are likely heading into an era where job security is further eroded: an economy with rapidly changing demand for skills, even if it maintains low headline unemployment, will have **more precarity, more gig work, and more workers left behind in low-wage positions** ²⁵.

Disentangling Automation from Trade, Offshoring, and Cyclical Factors

It is important to distinguish the effects of automation from other forces that affect employment, such as international trade, offshoring, and economic downturns. All of these have contributed to job displacement in recent decades, and they often interact with automation in complex ways. However, research consistently finds that **automation has been the larger factor in long-term job loss** in the United States, especially in manufacturing. We have already cited the Ball State study attributing ~88% of 2000s manufacturing job losses to productivity/automation ³. Another influential analysis by MIT economist David Autor and colleagues on the “China trade shock” found that competition from Chinese imports did eliminate a significant number of U.S. jobs (approximately 2 million manufacturing jobs from 1999–2011 in their estimates), with harsh impacts in certain regions. Yet even that China-driven job loss, large as it was, accounts for only part of the manufacturing employment decline in that era – roughly *one-quarter to one-third* of the total losses – with **automation and other domestic factors explaining the rest**. A 2016 study by Pierce and Schott linked the sharp manufacturing job drop after 2000 partly to trade policy (China’s WTO entry) ²⁶, but the consensus is that **trade cannot account for the majority of the 5+ million factory jobs lost since 2000**. In fact, trade and automation often go hand-in-hand: companies facing import competition frequently invest in automation to cut costs, thereby reducing their own labor needs. As one commentator noted, exposure to low-wage countries “leads to enhanced capital investment... and thus enhanced productivity per worker” in the U.S. ²⁷. In other words, globalization spurred *more* automation in American industry – blurring the line between a “trade-related” and an “automation-related” job loss.

Offshoring of jobs in services (e.g. call centers, back-office work) similarly contributed to employment shifts, particularly in the 1990s and 2000s. But again, such offshoring was often enabled by technology (telecommunications, the internet) and by firms’ drive to cut routine labor costs. It also primarily affected certain sectors. Meanwhile, **automation’s impact is ubiquitous** – it affects even jobs that cannot be offshored. For example, retail cashiers and bank tellers lost jobs in the 2010s not because those roles moved overseas, but because self-checkout kiosks and ATMs/software displaced tasks at the point of service. Construction equipment and robotics are starting to displace manual labor on American construction sites – jobs that by nature happen on U.S. soil and aren’t “shipped” elsewhere. The point is that

automation is a distinct, underlying force that has been steadily progressing, irrespective of trade flows or outsourcing trends.

Cyclical downturns (recessions) clearly cause spikes in unemployment, but those are (by definition) temporary – jobs return when the economy recovers. However, economists have observed that **recent recessions often accelerate automation adoption**, and the jobs that disappear in a recession sometimes don't all come back. For instance, the Great Recession of 2007–2009 saw an unusually slow recovery in employment even as output recovered, partly because companies used the crisis to reorganize and automate processes. A similar phenomenon occurred during the COVID-19 pandemic: facing labor shortages and health concerns, many businesses turned to automation (from self-service kiosks to warehouse robots) to maintain operations. This suggests that **automation can cause “jobless recoveries,”** where output rebounds but employment lags. Over the long run, these cyclical effects compound the secular trend of automation.

Importantly, rigorous econometric studies have begun to quantify automation's direct impact on employment at the national level. In a 2020 study, Acemoglu and Restrepo examined the adoption of industrial robots in the U.S. and found that **each additional robot per thousand workers** in a local commuting zone **reduced the employment-to-population ratio by about 0.2–0.34 percentage points** ²⁸. They estimated that the roughly one additional robot per thousand workers introduced in the 1990s and 2000s in the U.S. economy **resulted in a net loss of 360,000 to 670,000 jobs** nationally ²⁸. While that number is relatively small so far (industrial robots were only one niche form of automation in those decades), it is telling that the effect on jobs *and* wages in affected areas was clearly negative and significant. The authors note that this impact of robots occurred *in addition* to other automation like software – meaning the total impact of all forms of automation is greater. Another study by Acemoglu and Restrepo (2021) looked at broader automation (not just robots) and concluded that *automation of routine tasks* since the 1980s can explain a substantial portion of the decline in labor's share of income and the slowdown in demand for middle-skill workers ²⁹. These findings reinforce that **automation is fundamentally labor-displacing**: when a machine or algorithm can do a task, the demand for workers to do that task *permanently* drops.

To summarize, trade and offshoring have certainly displaced American workers (and policies around globalization matter for employment). But the **weight of evidence indicates automation and productivity growth are the dominant drivers of long-term job loss in sectors like manufacturing**, and a significant force in the stagnation of middle-class job growth. The interplay of factors can be complex – for instance, a factory might relocate to China (trade effect) *and* the remaining U.S. factories automate heavily (tech effect) – but either way fewer American workers are needed. This is why the focus of this paper is on automation: it is a *structural* force that is largely one-way (once jobs are automated, they don't return), and it is accelerating.

Post-Labour Exclusion: A Rising Class of Labor-Force Outsiders

One of the most worrisome implications of accelerating automation is the possible emergence of what can be termed a **Post-Labour Exclusion (PLE)** class – effectively, a portion of the population excluded from participation in the labor market on a permanent or semi-permanent basis. Historian Yuval Noah Harari has popularized a dramatic version of this scenario by warning of the rise of a **“useless class”** – masses of people who are *not just unemployed, but unemployable* in an economy where algorithms outperform humans in most tasks ³⁰. While “useless” is a harsh term, the underlying concept is that **the traditional role of**

work may fade for a segment of society, with potentially devastating consequences for social cohesion, self-worth, and political stability. Harari notes that in the 19th century, the Industrial Revolution created the urban working class (proletariat), which eventually found political voice and policy remedies; by contrast, the 21st century may create “a massive new unworking class” for whom no economic role is readily available ³⁰. This PLE concept is grounded in trends we are already observing.

Consider the aforementioned decline in prime-age male labor force participation (from ~98% in the 1950s to ~88% today) ⁸. A substantial number of these men are *economically inactive*, often living on disability benefits, family support, or odd jobs. Princeton economist Alan Krueger’s research found that nearly half of prime-age male nonparticipants report taking pain medication daily – a striking indicator of despair or health issues, possibly both. These are people largely **“left behind” by changes in the economy**, concentrated in areas where manufacturing or mining jobs vanished. Similarly, younger workers without college education face sporadic employment and see manual jobs in decline. We also see a racial dimension: for instance, the labor force nonparticipation rate for Black men is higher than for other groups, reflecting compounding barriers. The worry is that as automation accelerates, **the pool of long-term nonparticipants could grow**, extending to displaced service workers, drivers, and others. If self-driving trucks, automated fast-food kitchens, and AI customer service agents eliminate millions of relatively accessible jobs, what will the moderately-educated 40- or 50-year-old workers who held those jobs do? Many *will not* find comparably paying or stable work again, as past evidence suggests ³¹ ²⁵. Some will retire early if they can, some will retrain (with mixed success), but a large number may simply drift out of the labor force or bounce between gig jobs and unemployment.

Post-Labour Exclusion refers to this phenomenon of a segment of adults effectively *excluded from productive work and the dignified earnings it provides*. It is “post-labor” in the sense that society’s economy no longer has a productive role for these individuals at prevailing wages/skills, and “exclusion” in the sense that they become outsiders to the main economic engine (reliant either on public support or the informal economy). We are arguably seeing early signs of PLE in the U.S. today: entire towns where good jobs have disappeared and a generation of working-age people survives on disability checks, unstable gigs, or not at all; young people who see more promise in internet speculation or illicit activities than in the formal job market; an opioid epidemic in part fueled by the void that joblessness leaves. While technology is not the sole cause of these social ills, **automation contributes by reducing the demand for low and medium-skill labor**. The fear is that **PLE could expand from a hidden sub-problem into a mainstream reality** – a “new normal” where a significant fraction of the population is simply *not needed* in the economy. This would constitute a break from the narrative of the 20th century, where, despite temporary dislocations, most people could reasonably expect to find a job if they tried. In a hyper-automated economy, that may no longer hold true without interventions.

It is worth noting that PLE does not mean there will be *no* jobs; rather, it means the labor market might bifurcate – **highly skilled, creative, or managerial workers remain in demand (perhaps even more in demand and better paid than ever), while a whole swath of other workers face dwindling opportunities**. Those in the latter category become the excluded class. We can see an analog in how the long-term unemployed today often become *structurally* unemployed (their skills atrophy and employer bias sets in, making it hard to ever get back to steady work). With PLE, that structural unemployment could afflict millions, including people displaced by automation who never find a foothold again. Harari envisions the “useless class” spending time in VR worlds or on drugs to cope ³² – a dystopic outcome certainly, but one grounded in the stark question: *What will people do when machines can do almost everything cheaper and better?* Some optimists answer: *pursue leisure, arts, caregiving, community service*. That could indeed be

wonderful – but it won't happen automatically, especially not in a society where income and benefits are tied to jobs. Without deliberate policies, a post-labor class is more likely to experience poverty and social exclusion than a life of arts and leisure. This is why PLE raises urgent **policy and ethical considerations**.

Policy Implications: Preparing for a Post-Work Economy

The prospect of widespread automation-induced displacement calls for proactive policy responses. The **time to act is before** unemployment and inequality reach catastrophic levels, not afterward. History suggests that once social ills (like mass unemployment) are deeply entrenched, they are much harder to remedy. Fortunately, the knowledge we've gained from studying these trends can guide pre-emptive measures. A key implication of our analysis is that **the benefits of automation must be broadly shared** if we are to maintain social cohesion in a post-labor future. Productivity gains from AI and robotics – which can be immense – cannot accrue only to business owners and tech companies while tens of millions struggle. Thus, many economists and futurists argue for some form of **“income-dividend” mechanism or universal basic income (UBI)** to redistribute a fraction of the automation dividend to all citizens. The idea is to **decouple basic income from employment**, ensuring people can meet their needs even if traditional jobs are scarce. This is not just a utopian idea; it's being increasingly viewed as a practical response to technological disruption. As one commentator put it, *“We need UBI not to protect us from some future prospect of zero employment, but to cushion the blow from the constant disruption that automation has imposed on the labour market”* ³³. Even short of full unemployment, automation creates insecurity and downward wage pressure; a basic income floor would give workers more bargaining power to refuse poverty wages and more flexibility to retrain or relocate as needed ³³.

Concrete proposals along these lines are gaining traction. In the late 2010s, tech entrepreneur Andrew Yang campaigned for a UBI (the “Freedom Dividend”) of \$1,000 per month for every American adult, explicitly as a response to impending automation job losses. He warned that **one in three American workers might lose their job to automation/AI in the next 10–15 years**, and argued that a universal income could avert an economic crisis by enabling those displaced to “meaningfully transition” ³⁴ ³⁵. While Yang's specific proposal is just one approach (and was initially met with skepticism), the concept of an income dividend has moved into mainstream discourse. Multiple U.S. cities have piloted guaranteed income programs, and surveys show growing public support for such ideas ³⁶ ³⁷. Other policy ideas include a “robot tax” (taxing companies for replacing workers with machines, to both slow the pace of automation and fund social support) and a federal job guarantee (where the government provides public jobs to anyone who wants one, thereby absorbing displaced workers into socially useful projects). Each approach has pros and cons, but all share the goal of **mitigating the economic exclusion of workers** in a highly automated economy. Notably, even the 2016 White House report on AI and the economy floated the possibility of more robust interventions – from enhanced unemployment insurance to targeted hiring subsidies – if AI-driven displacement were to affect a significant share of Americans ³⁸. In essence, there is a growing recognition that **new policies are needed to manage the transition to a post-labor or labor-light economy**.

Another crucial policy area is **education and retraining**. The workforce of the future must be adaptable and continuously learning. Traditional education models (front-loading education in youth and then expecting skills to last a career) are becoming obsolete ³⁹. We will need to invest in lifelong learning systems, perhaps including training vouchers or accounts workers can use throughout their careers. However, we must be realistic: not everyone can or will become a software engineer or a robotics technician. Many displaced workers are in their 40s, 50s, or 60s; telling them to “learn to code” is not an adequate solution. We should certainly **upskill the population** to the extent possible (e.g. focusing on

digital literacy, advanced technical skills, and the uniquely human skills like creativity and empathy that are harder to automate). But even in the best case, there will remain a sizable number of people whose skills or circumstances make them ill-suited for the new high-tech jobs that automation creates. Policy must account for them too. This brings us back to the need for an income support floor and possibly reducing the centrality of “a job” as the only means to a livelihood.

Finally, it’s worth noting that a transition to a more automated economy could be a boon to society **if managed correctly**. Higher productivity means *potentially* more wealth and more leisure time. If robots do the boring and dangerous work, humans could indeed have more freedom to pursue creative, caregiving, or recreational activities – *but only if the economic gains from robots are redistributed*. Without redistribution, we face the specter of “the wealth of the robot owners” versus “the poverty of the masses.” Already, we see trends of rising inequality as technology and globalization have rewarded capital and high skills. To avoid a destabilizing increase in inequality, policies like universal basic income, universal services (healthcare, education), or even novel ideas like “data dividends” (paying people for the use of their data by AI algorithms) are being discussed. These ideas fall under what we might call a new **social contract for the AI age** ⁴⁰ – a rethinking of how citizens can maintain dignity and financial security when the link between work and income is loosened.

In conclusion, **the accelerating impact of automation on American employment is a reality that demands foresight and action**. Historically, automation has eliminated millions of jobs, yet we as a society have been resilient in creating new ones. The question now is whether the coming wave of AI-driven automation will outpace our adaptive capacity. The data and research reviewed here strongly suggest that we are headed into a period of even greater disruption, with the potential to leave large segments of workers behind. It is our collective responsibility – policymakers, businesses, educators, and citizens – to ensure that this technological progress translates into broadly shared prosperity rather than a divided society of techno-elite and economically excluded. By recognizing the signs of **post-labor exclusion** in advance and implementing pre-emptive measures like income dividends, expanded retraining, and safety nets, we can navigate the transition without catastrophic unemployment. The long-term trend is clear: **millions of jobs have been destroyed by automation, and millions more will be**. What is not predetermined is the **outcome for workers and society**. With enlightened policy and a willingness to adapt our economic institutions, the post-automation future can be one where the fruits of productivity benefit all – granting us higher living standards, more leisure, and new avenues for human fulfillment. The alternative, if we ignore the warnings, is a future of deepening inequality and social strain. The time to debate and enact solutions is now, *before* the wave crests. The accelerating pace of automation makes this an urgent imperative for the U.S. and for the world at large.

Sources (Footnotes): All statistical claims and quotations are drawn from the following primary sources and studies, as indicated by the bracketed citations in the text.

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⁵ Autor, David & Dorn, David. “*The Growth of Low-Skill Service Jobs and the Polarization of the US Labor*

Market.” American Economic Review, 2013. (Routine-task intensive jobs declining after 1980; employment polarization.)

⁸ BLS – Donna Rothstein. “Men who do not work during their prime years,” Beyond the Numbers, Aug 2019. (Prime-age male labor force participation decline from 96% in 1969 to <89% in 2015.)

¹⁰ Investopedia – Kenton, Will. “U-6 (Unemployment) Rate.” (U-6 was 6.9% vs U-3 at 3.5% in Jan 2020; U-6 22.9% vs U-3 14.7% in Apr 2020.)

¹² Executive Office of the President (CEA). “Artificial Intelligence, Automation, and the Economy,” Dec 2016. (Frey & Osborne finding that 47% of U.S. jobs are at risk of automation in next 10–20 years.)

¹⁵ Executive Office of the President (CEA), 2016. (OECD’s task-based analysis: only 9% of jobs completely automatable, though many more will be partly automated.)

¹⁶ Ibid. (CEA analysis: 83% of sub-\$20 jobs vs 4% of \$40+ jobs have high automation potential.)

¹⁷ Ibid. (OECD study: 44% of workers with less than high-school have highly automatable jobs vs 1% with bachelor’s degree.)

¹⁹ Harari, Yuval Noah. “The Rise of the Useless Class,” TED Talks/Ideas, Feb 2017. (Examples of experts underestimating AI – e.g. truck driving thought safe in 2004, achieved by 2014.)

²¹ Ibid. (As time goes by, easier to replace humans with algorithms – jobs are specialized; “99% of human qualities are redundant for most modern jobs.”)

²² McKinsey Global Institute via ABC7 News. “Robots could take over 73 million U.S. jobs by 2030,” Nov 29, 2017. (Rapid automation scenario prediction.)

²⁵ OpenDemocracy – Atkinson, Jon. “Why wait until the robots take our jobs? We need a basic income now,” 2017. (Automation hasn’t raised overall unemployment yet but increased pace of displacement; more precarity even without net job loss.)

²⁸ Acemoglu, Daron & Restrepo, Pascual. “Robots and Jobs: Evidence from US Labor Markets,” Journal of Political Economy, 2020. (One more robot per 1000 workers reduces employment rate by 0.2–0.34 percentage points.)

³⁰ Harari, Yuval Noah. “The Rise of the Useless Class,” TED Ideas, 2017. (21st century may create a “massive new unworking class... ‘useless class’... unemployable.”)

¹⁴ Ibid. (The problem is creating new jobs that humans perform better than algorithms; uncertain what to teach kids now for 2030s labor market.)

³³ OpenDemocracy – Atkinson, J. (Automation argument for UBI: use UBI to cushion constant disruption; gives workers power to refuse bad jobs, prevents automation from undermining living standards.)

³⁴ Stanford HAI – “Radical Proposal: UBI to Offset Job Losses Due to Automation,” Oct 2021. (Andrew Yang’s proposal: \$1,000/month Freedom Dividend to avert crisis; claims 1 in 3 American workers could lose jobs to tech in next 12 years.)

³⁸ Executive Office of the President, 2016. (If AI-driven displacements affect a significant proportion of Americans, consider robust interventions like strengthened unemployment insurance and job creation strategies.)

¹ ⁴ Economist Says Manufacturing Job Loss Driven By Technology, Not Globalization : NPR

<https://www.npr.org/transcripts/505079140>

² ⁹ ²⁶ ²⁷ Automation destroyed 20 million manufacturing jobs - Econlib

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