

Re-assessing the Spatial Mismatch Hypothesis

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The spatial mismatch (SM) hypothesis is that the combination of residential segregation and an uneven distribution of jobs limits the employment opportunities of Black workers (Kain 1968, Ellwood 1986, Wilson 1987). We evaluate this hypothesis in the 2010s using data from the Longitudinal Employer-Household Dynamics (LEHD) program at the U.S. Census, which has detailed residential and workplace locations for nearly all U.S. workers. We also introduce a new measure of job quality to the SM literature: the establishment earnings premium, estimated from an Abowd-Kramarz-Margolies (1999) (hereafter AKM) earnings decomposition.

We focus on two groups of cities: (1) older industrial cities in the Northeast and Midwest; and (2) newer Sunbelt cities in the South and Southwest.¹ We use an AKM model to decompose the Black-white earnings gap in these cities into components attributable to permanent worker effects and to pay premiums of employers. Our first key finding is that virtually none of the Black-white gap reflects differences in average establishment pay premiums. This surprising result is the opposite of what we would expect if distance to high-paying jobs was a major driver of the gap.

Next, we examine the geographic distribution of workers and jobs. We show that Black workers' homes are, if anything, closer to potential workplaces in their city, and to workplaces with high earnings premiums, than are white workers' homes.

Ellwood (1986) argues that in an equilibrium model with mismatch, Blacks will have longer commute distances than whites. Our third key finding is that Black workers' average commutes are *shorter* than those of white workers. Among both Blacks and whites longer

¹ The older industrial cities in our analysis are: Philadelphia, Detroit, Pittsburgh, Cleveland, Newark, Buffalo, Baltimore, Chicago,

Minneapolis and St. Louis. The newer Sunbelt cities are Los Angeles, Houston, Atlanta, Miami, Dallas, San Diego and Phoenix.

commutes are associated with higher-paying jobs (i.e., jobs at establishments with higher AKM pay premiums), as one might expect if workers trade off pay and commuting distance. However, this association is no stronger for Black workers, as would be expected if SM limited Black workers' access to high-paying jobs. Last, in the older industrial cities, jobs near Black neighborhoods offer *higher* average pay premiums than do jobs near white neighborhoods. The relationship is weaker, though still positive, in the Sunbelt cities.

Overall, we find no evidence that SM is a primary contributor to Black-white earnings gaps in the 2010s. We close by suggesting some directions for further research on the micro-geography of employment and earnings.

I. The Spatial Mismatch Hypothesis

The idea of spatial mismatch originated in a 1965 paper by John Kain and was expounded in Kain (1968). A subsequent literature – reviewed in the online Appendix – has found decidedly mixed evidence regarding the SM hypothesis. A secondary concern in the literature (e.g., Ellwood, 1986) is that the theoretical underpinnings of SM are weak. In the online Appendix we sketch a partial equilibrium model of worker choices over jobs in some potential “opportunity set” that includes jobs with different pay premiums at

different commute distances. In that framework it is useful to characterize the relative number of higher- and lower-premium jobs in a city that are within a given commute distance of a typical Black or white worker. We also discuss how comparisons between the regression coefficients for Blacks and whites relating AKM pay premiums to commuting distances can be interpreted as evidence of differences in their average opportunity sets. Our framework takes worker and job locations as given, and is only a first step in incorporating AKM-style wage-setting features in a spatial model.

II. Data

Much of the existing evidence on SM has utilized data from only a few cities with historically high rates of residential segregation – for example Chicago (in the case of Ellwood, 1986), Detroit, or both (in the case of Kain, 1968). In an effort to expand the evidence we present results for two groupings of larger commuting zones (CZs). One group consists of ten older CZs in the Northeast and Midwest. These have a traditional urban structure, with Black neighborhoods concentrated near the city center and largely white suburbs and exurbs. The second group is seven CZs in the South and Southwest that emerged as major centers later in the 20th century, and tend to have a less monocentric structure.

Characteristics of these two sets of CZs, and two other groupings of cities that make up the balance of the 30 largest CZs in the country, are presented in the online Appendix

To study mismatch in these CZs we use quarterly earnings data from 2010Q1 to 2018Q2 from the LEHD, linked to detailed residential and workplace location information. In this period the LEHD covers about 95% of private sector employment, as well as state and local government employees. We limit attention to non-Hispanic Black and white workers who are likely full-quarter, full-time workers with a single employer in the quarter; see the Appendix for details.

We measure commute distance as the distance between a worker's residential location and that of the establishment where he/she works. Because CZs vary so much in their scales and commute distance distributions, in many of our analyses we rescale distances to set the 75th percentile commute in each CZ to equal 16 miles.

In most states, the assignment of workers to establishments in multi-establishment firms is probabilistically imputed using information on the worker's residence and the locations of the firm's establishments. We rely on these imputations, and note below one case in which results appear sensitive to them.

Changes in residential patterns since the 1960s may have reduced SM. Unfortunately, data like ours do not exist for earlier periods.

III. Decomposition of the Black-white Earnings Gap

A core part of the mismatch hypothesis is that Black workers have access to worse jobs than do similarly-skilled White workers, by virtue of their residential locations. Thus, we might expect that employed Black workers work at worse firms than do employed white workers.

We begin with a simple AKM decomposition of log quarterly earnings (y_{it} , for person i in quarter t) into components representing permanent person effects, α_i , common effects for the establishment $f(i,t)$ at which the worker is employed, $\delta_{f(i,t)}$, a few time-varying controls (X_{it} , calendar time dummies and a cubic in age), and a residual, ϵ_{it} :

$$(1) \quad y_{it} = \alpha_i + \delta_{f(i,t)} + X_{it}\beta + \epsilon_{it}.$$

We estimate this separately for each CZ, normalizing the restaurant industry to have zero average pay premium. We then average the estimates from (1) by CZ and race to decompose the White-Black log earnings gap in CZ c into parts reflecting mean difference by race in the person effects, the establishment effects, and the covariates:

$$(2) \quad \bar{y}_{cw} - \bar{y}_{cb} = (\bar{\alpha}_{cw} - \bar{\alpha}_{cb}) + (\bar{\delta}_{cw} - \bar{\delta}_{cb}) + (\bar{X}_{cw} - \bar{X}_{cb})\beta.$$

The third term is negligible so this yields a two-part decomposition.

Table 1 presents this decomposition, averaged across CZs in each grouping. In both sets of CZs the quarterly white-Black earnings gap is about 35 log points – similar to the pay gaps for full-year earners in the American Community Survey (see the online Appendix). The AKM decomposition attributes virtually all of this to differences in the person effects. While differences in workplace pay premiums account for over 10% of the variation in wages across workers, there is no appreciable difference in the mean pay premiums received by Blacks versus whites. This is an initial indication that access to high-paying establishments is not a primary determinant of the Black-white earnings gap.²

The absence of a racial gap in pay premiums is potentially surprising given the tendency for assortative matching between high-skilled workers and high-premium workplaces, and the fact that Black workers are estimated to have around 35% lower person effects than whites. It is potentially consistent, however, with longstanding evidence that Blacks are

more likely to work in unionized jobs (see the Appendix for further discussion).

[Insert Table 1 Here]

IV. Geographic Distribution of Workers and Jobs

Figure 1 begins our analysis of the location of jobs relative to workers. For each worker and for varying radii r , we compute the share of all jobs in the CZ that are located within radius r . We average this across Black and White workers separately in each CZ, then average across CZs in our two groupings.

[Insert Figure 1 Here]

The figure shows that, for every r , there are more jobs within r miles of the typical Black worker than of the typical White worker in the older industrial CZs. The same is true for the newer Sunbelt CZs for any $r > 1$. Moreover, the jobs near Black workers are of no lower quality (as measured by the establishment premium) than those near White workers. Dashed lines show the share of jobs at establishments with δ_f in the top tercile within radius r ; these tell a similar story as the all-jobs series. There is no indication that there is a systematic shortage of

² This does not rule out that employers as a whole are discriminating against Black workers either in employment or wage setting. Insofar as

this discrimination is similar across firms, this could appear as reduced person effects for Black workers rather than in the δ term.

jobs, or of good jobs, within a reasonable commuting distance of Black workers. The Appendix presents an analysis that distinguishes jobs held by college and non-college workers; this does not affect the result.

Another way to measure access to good jobs is via the correlation between the fraction of residents at a location who are Black and the average premium of all establishments within a short commuting range of that location. We assign workers to locations defined by a 0.5-mile-by-0.5 mile grid, and measure the average δ_f of all establishments within 2.5 miles of each location. The correlation of this measure of nearby job quality with location fraction Black is 0.26 for the older industrial CZs and 0.10 for the newer sunbelt CZs – if anything, jobs near Black neighborhoods are *better* than those near white neighborhoods.

Ellwood (1986) proposes what we would now call an “outcomes test” for spatial mismatch: If Black workers have less access to nearby jobs, we should see longer commutes for this population in equilibrium. In the older industrial CZs, the mean Black worker commutes only 86% as far as the mean white worker, and the white commute distribution stochastically dominates that of Blacks. In the newer Sunbelt CZs, mean white and Black commute distances are about the same (see Appendix Figure A-2 and Table A-3).

Figure 2 shows how job quality relates to commute distance. In both groups of cities and for both races, longer commutes are associated with better establishment quality (δ_f), as would be expected if workers trade off wages against commute time in job search. In the older industrial CZs Black workers with commute distances up to 10 miles commute to better (higher δ_f) establishments than do White workers with similar commutes, while in the newer Sunbelt CZs establishment quality conditional on commute distance is similar for the two races. In both sets of CZs, among workers with the longest commutes Black workers commute to worse establishments. This implies that the establishment quality-commute distance tradeoff is less steep for Black workers, the opposite of what we would expect under spatial mismatch.

[Insert Figure 2 Here]

One notable aspect of Figure 2 is that, while job quality is generally increasing in commute distance, the pattern reverses for commutes longer than about 20 miles. Appendix Figure A-3 presents an analysis of this relationship that distinguishes by firm size and number of establishments. The downward slope is driven by workers at multi-establishment firms. We suspect that this is an artifact of the algorithm for imputing a worker’s establishment, which

is based on establishment locations but not on their relative wage effects. Figure 2 suggests that workers may be particularly unlikely to commute past a nearby establishment to work at another one with a lower wage premium; if so, this would create bias for long commutes at multi-establishment firms. The upward slope through most of the distribution in Figure 2 appears to be robust to this, however, as it holds even for workers at single-establishment firms.

As a final exercise, we estimate the elasticity of earnings with respect to commuting distance, as well as elasticities of the individual and establishment AKM wage components. Again, spatial mismatch should imply a steeper gradient of earnings with respect to commute distance for Black workers, driven by the establishment wage component. As shown in the Appendix, we see no evidence of this in the LEHD or (using commute time) in the ACS.

V. Discussion

Our findings suggest that geographic proximity to “good” jobs – as measured by the AKM pay premiums offered at different workplaces – is not a major source of racial wage gaps in large cities in the U.S. today (though it may have been in the 1960s). More research is needed to understand whether the negligible difference in average pay premiums received by Black and white workers masks

differences for particular employers or sectors. In ongoing work we are also investigating differences between males and females, and between Hispanic and non-Hispanic workers. Finally, it is important to note that our analysis focuses on racial gaps conditional on working full time: whether spatial mismatch has more relevance for other margins (such as employment versus non-employment) is clearly an important remaining question.

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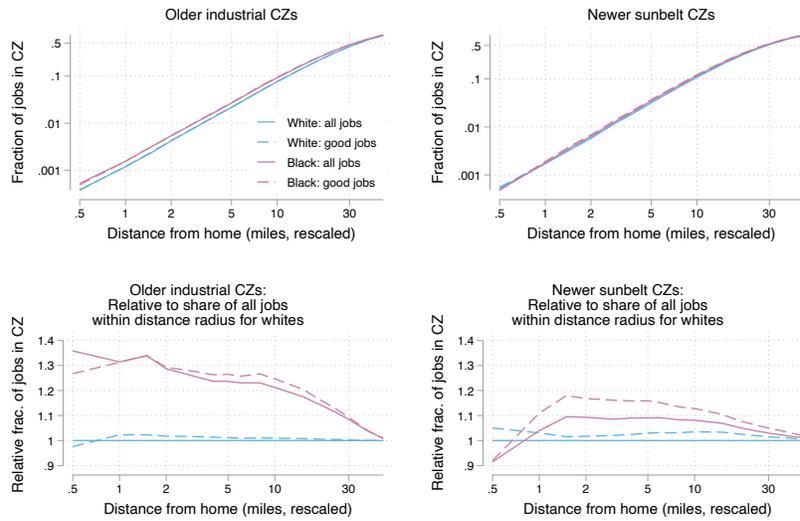


FIGURE 1. FRACTION OF ALL JOBS AND JOBS AT GOOD ESTABLISHMENTS WITHIN COMMUTING DISTANCE OF AVERAGE BLACK AND WHITE WORKER

Note: Distances for each CZ are rescaled to set the 75th percentile commute distance to 16 miles. “Good jobs” are those at establishments with AKM establishment effects in the top tercile.

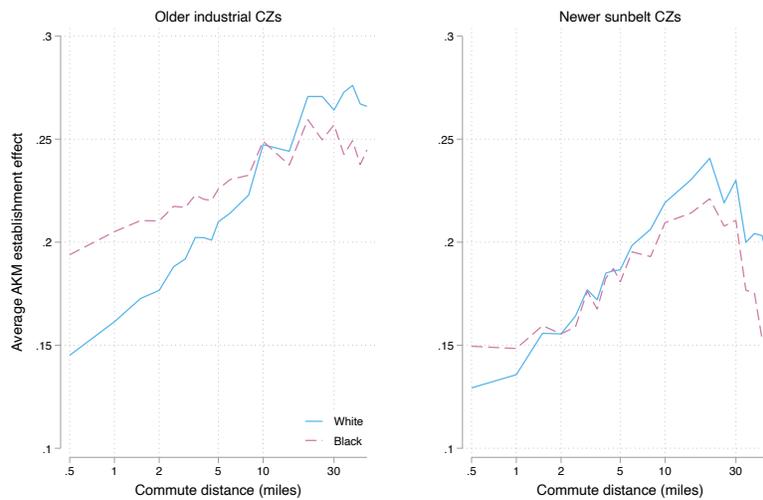


FIGURE 2. AKM ESTABLISHMENT EFFECT AND COMMUTE DISTANCE, BY CZ GROUP AND RACE

Note: Distances for each CZ are rescaled to set the 75th percentile commute distance to 16 miles.

| TABLE 1—AKM DECOMPOSITION OF BLACK-WHITE QUARTERLY EARNINGS GAP | | |
|---|----------------------|-------------------|
| | Older industrial CZs | Newer sunbelt CZs |
| Geometric mean earnings (quarterly, nominal dollars) | | |
| White | 14,271 | 15,835 |
| Black | 10,067 | 10,993 |
| Log earnings gap | 0.35 | 0.36 |
| Share of gap attributable to (%) | | |
| Person effects | 105.1 | 104.1 |
| Establishment effects | -0.8 | 1.2 |

Notes: Author calculations from LEHD data. Sample excludes person-quarters with multiple employers or earnings below \$3800, as well as the first and last quarter of each spell. AKM model is fit separately for each CZ, with the average establishment effect in the restaurant industry normalized to zero. Results are then pooled across CZs in each group. Geometric mean earnings are the exponentiation of mean log earnings across person-quarters in the AKM sample.

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ONLINE APPENDIX

A. The spatial mismatch literature

In his paper introducing the spatial mismatch (SM) hypothesis, Kain (1968) presented a series of models that related the share of jobs held by black workers in a given “work zone” in Chicago or Detroit to the Black share of residents in that zone and the distance from the zone to the nearest “ghetto.”¹ He then performed a counterfactual analysis, assuming that Black residential shares were equalized across zones (eliminating the ghetto areas) but that the model coefficients remained unchanged. The results suggested that the elimination of residential segregation would lead to a 10% rise in Black employment in Chicago and a 3-7% rise in Detroit.

Two criticisms were raised almost immediately about these findings. First, the conclusions of Kain’s counterfactual analysis were reversed if the model was relaxed slightly to allow a nonlinear relation between Black residential shares and Black employment shares (Offner and Saks, 1971). Second, it is not obvious how to derive Kain’s specification from an underlying model of the choices of Black and white workers about where to work. Many subsequent studies tried to incorporate broader city-wide measures of the decentralization of jobs (e.g., Mooney, 1969; Masters 1974) or of the average distance between residential locations and potential work locations (e.g., Hutchinson, 1974). While some studies in this vein found support for SM, others did not. In a careful review of the literature up to 1990, Holzer (1991, p. 118) concluded that “... the evidence remains very contradictory.” Glaeser, Hanushek and Quigley (2004, p. 76) are less sympathetic, asserting: “It is not generally true that Blacks live further from jobs than Whites do, and it is hard to believe that the physical costs of getting to jobs are really responsible for the pathologies of the ghetto.”

¹ The precise definition of a zone is unclear, but maps in Kain (1964), which presents the same empirical evidence as Kain (1968), suggest the zones are related to Census tracts. Similarly, Kain does not discuss how he defines the boundaries of the ghettos in Chicago and Detroit in the mid-1950s.

A useful theoretical perspective on the SM hypothesis is provided by Ellwood (1986), who discusses the conditions under which the employment outcomes of two groups of equally productive workers are affected by their residential locations. He argues that the clearest way to test the SM hypothesis is relate outcomes of workers (or potential workers) in a neighborhood to measures of job proximity for people in that neighborhood. This is the idea that guides our analysis.

There have been important changes in residential patterns since the 1960s that may have reduced the importance of SM. Black households are much less concentrated in formerly redlined neighborhoods in central cities than they were in the 1960s, and the Black population has suburbanized to a significant extent. Moreover, many more workers commute by car rather than public transportation. However, residential segregation remains high, preserving the plausibility of SM as a potential explanation for racial differences in labor market outcomes.

B. A Simple Model

i) Basic Setup

In this section we sketch a simple model of wage outcomes for workers in a spatially differentiated labor market. The model explicitly builds in an AKM-style model of wage setting in which each establishment offers a proportional wage premium that raises or lowers wages of any worker who is employed there relative to other workplaces in the market. Similar to the monopsonistic competition model in Card et al. (2018) the model focuses on worker's preferences over available jobs, ignoring frictions in the matching process.

Specifically, assume that person i gets utility from employer j :

$$u_{ij} = \alpha_i + \delta_j - \beta_i d_{ij} + \epsilon_{ij},$$

where δ_j is the pay premium offered by employer j , d_{ij} is a measure of the commute distance for i to get to workplace j , and ϵ_{ij} is a match effect. If worker i takes a job at employer j her observed wage is:

$$\ln w_{ij} = \alpha_i + \delta_j + v_{ij},$$

where (as in a standard AKM model) α_i is a common component of wages for i across all jobs, and the residual term v_{ij} is assumed to be independent of d_{ij} and ϵ_{ij} .

Next, assume that a worker who is searching for a job has an “offer set” O_i representing the potential set of job opportunities that are available. She takes the job with the highest utility in the set:

$$j^*(i) = \operatorname{argmax}_{j \in O_i} [\delta_j - \beta_i d_{ij} + \epsilon_{ij}],$$

and we observe the combination of the wage premium and commute distance $(\delta_{j^*(i)}, d_{ij^*(i)})$ for that worker.

ii) Comparing Job Opportunities of Different Groups

Suppose there are two groups of workers G_1 and G_2 . If the joint distributions of (δ_j, d_{ij}) in the offer sets are the same for the two groups, and they have the same distribution of β_i 's, then they will have the same probability distributions over $(\delta_{j^*(i)}, d_{ij^*(i)})$. In particular, the conditional expectation of the wage premium, given commute distance

$$E[\delta_{j^*(i)} | d_{ij^*(i)} = d]$$

will be the same for the two groups. This provides the basis for a simple “outcome test”: if two groups have the same access to jobs, and the same preferences for wages versus commuting distance, then we would expect the observed relationship between wage premiums and commute distances to be the same for the two groups.

To facilitate comparisons between workers with different offer sets, suppose that commute distances are discrete, $d_{ij} \in \{d_1, d_2, \dots, d_N\}$, and that wage premiums are also discrete, $\delta_j \in \{\delta_1, \delta_2, \dots, \delta_M\}$. In this case the offer set for a given worker is summarized by which particular combinations of (δ_u, d_v) are available (i.e., the support of the joint distribution of wage premiums and commute distances). For example a high-wage premium job at close proximity may not be available in a given worker's choice set.

Suppose that the offer sets for individuals in group G_1 have the property that jobs with wage premiums $\delta_j \in \{\delta_1, \delta_2, \dots, \delta_M\}$ are available at every commute distance, while the offer sets for individuals in group G_2 have the property that jobs with wage premiums $\delta_j > \bar{\delta}$ are only available with commute distances $d_{ij} > \bar{d}$. In this case we would say that the job opportunities of group G_2 are negatively affected by their residential locations, relative to

group G_1 . In particular we would expect that the observed wage premiums for workers in G_2 with relatively short commuting distances would be lower than the premiums for workers in G_1 in the same range of commute distances. We would also expect that the slope of the conditional expectation of the wage premium, given the commute distance, will be higher for the disadvantaged group.

C. LEHD sample and comparisons to American Community Survey

(i) LEHD Sample

We use data from Census Bureau’s Longitudinal Employer-Household Dynamics (LEHD) program. These data are derived from quarterly earnings reports provided by employers to state unemployment insurance (UI) agencies. The core data set includes total wages paid by a given employer to each worker in a quarter. This is supplemented with information on employers and workers derived from other sources (e.g., the decennial census and ACS files) – see Abowd et al. (2009). The LEHD covers about 95% of private sector employment, as well as state and local government employees, but excludes federal employees, members of the armed services, and self-employed workers. From 2010 forward it includes data from all 50 states.

Our sample construction follows Card et al. (2023). We begin with person-employer-quarter (PEQ) observations from 2010Q1 to 2018Q2 where the worker is between 22 and 62 years of age. To help screen out part-time jobs and/or partial-quarter job spells we exclude PEQs with earnings below \$3,800 (roughly the earnings from a full-time job at the federal minimum wage), quarters where an individual had multiple jobs, and all *transitional* quarters (the first or last quarter of any person-employer spell). We also drop PEQs with an unknown industry and/or establishment location. Finally, we drop individuals with fewer than 8 quarters of earnings that satisfy the previous restrictions over our 8½ year sample window, and individuals who are neither white non-Hispanic or Black non-Hispanic. We assign individuals to 1990 Commuting Zones (CZs) (Tolber and Sizer, 1996) based on the county of their establishment.

The upper rows of **Table A-1** reports summary statistics for non-Hispanic white and Black workers in three groups of CZs – the “older industrial” and “newer Sunbelt” CZs discussed

in the main paper, and a residual group consisting of all other CZs in the top-200. (We discuss the lower rows of the table below.) Not surprisingly, the first two groups – which are drawn exclusively from the largest 30 CZs – have somewhat higher earnings for both white and Black workers than does the latter group. They also have larger white-Black gaps in log earnings – 0.37 and 0.38, respectively, vs. 0.25 in the other CZs.

(ii) Comparisons to American Community Survey (ACS)

In this appendix, we also report some results for a sample constructed from the 2010-2018 ACS. We select people age 22-62 inclusive from the ACS with at least 1 year of Mincer experience (i.e., age-education-6>0). For our analyses of earnings outcomes we further limit attention to “full year earners” with annual wage and salary earnings of \$15,200 or higher (a threshold 4x higher than the quarterly threshold for full time work we impose on the LEHD). We assign 1990 CZs to Public Use Micro Areas (PUMA’s) identified in the ACS using PUMA-county population files for the 2000 and 2010 Census created by David Dorn.² For PUMA’s that contain observations from multiple CZs we probabilistically assign one CZ based on the relative share of the PUMA population in that CZ. Finally, we limit attention to individuals in the 30 largest CZs (based on counts of person quarter observations in our LEHD samples) and group the CZs into 4 groups: (1) Older industrial cities; (2) Newer Sunbelt cities; (3) Northeast Corridor; (4) Other CZs (which are mainly in the West). The resulting sample of workers in large CZs contains 6.49 million observations, representing a weighted population of roughly 721 million 22-62 year olds (80.2 million per sample year). In this sample 52% are white non-Hispanic and 13.7% are Black non-Hispanic; roughly 61.8% are classified as full year earners.

Table A-2 reports summary statistics for the working age populations in all larger CZs and in the four groups of cities, as well as statistics for the subset of full-year earners. (Note that in contrast to the statistics in Table A-1, these results include people of all ethnicities and racial groups). We note first that the population share of white and Black non-Hispanics varies across the four CZ groups, being relatively low in the older industrial cities and higher in the

² See <https://www.ddorn.net/data.htm>. We downloaded two files from this site: [E5] 2000 Census and 2005-2011 ACS Public Use Micro Areas to 1990 Commuting Zones; and [E6] 2010 Census and 2012-ongoing ACS Public Use Micro Areas to 1990 Commuting Zones.

Sunbelt cities. Mean years of education and the share of people with a BA or higher also vary somewhat across CZ groups and are particularly low in the Sunbelt (largely driven by the high share of Hispanics in these cities). Average employment rates (based on having earnings in the previous year) are fairly similar across city groups, ranging from 79 to 82%; the fraction of full-time earners varies a little more and is particularly low in the Sunbelt cities.

The middle rows of the table show characteristics of full year earners (i.e., with at least \$15,200 of earnings last year). This group is 12-15% Black non-Hispanic except in the Western CZs, and is about 45% female. Mean annual earnings range from 60,000 in the Sunbelt cities to 75,000 in the Northeast corridor; mean hourly wages show a similar range. On average about 86% of full-time earners in the top 30 CZs commute to work in their own car: this rate is higher in the Sunbelt (around 95%) and lower in the Northeast corridor (67%). Mean commute times average about 30 minutes (one way), but are a little higher in the Northeast corridor (partly reflecting the fact that commuters by bus and rail have relatively long average commute times).

Finally, the bottom three rows of the table show mean log hourly earnings of white non-Hispanics, Black non-Hispanics, and the Black-white earnings gap. The mean gap is about 32 log points and is slightly lower in the older industrial cities than the Sunbelt cities or the Northeast corridor. Importantly, the magnitudes of the Black-white gaps in annual earnings in our ACS sample are similar to the gaps in quarterly earnings in our LEHD sample. About one-eighth of the gap in earnings for full-year earners appears to be due to a lower hours among Black workers – the Black-white gap in hourly wages for full-year earners in the largest 30 CZs is 28 log points. This gap, in turn, is not too different than the 26 log point gap in log hourly wages for 2010-2018 reported by Wilson and Darity (2022), based on hourly or weekly wages reported in the monthly Current Population Surveys.

Table A-3 reports some statistics on each of the 30 CZs in our ACS sample. Most of the older industrial cities have around a 20% share of Black workers (the exceptions are Minneapolis and Pittsburgh) while nearly all the newer Sunbelt cities have a relatively high share of Hispanics (the exception is Atlanta). Average one-way commute time are pretty similar across CZs, but higher in New York, and to a lesser extent Washington DC.

(iii) Imputation of Establishment Locations in LEHD

The UI data in the LEHD contain an identifier for the employing firm and the state, but not for the specific establishment if the firm has multiple establishments in the same state. The Census Bureau uses data on workers' residential addresses and the locations of establishments owned by the firm to impute establishments for workers employed at multi-establishment firms (Vilhuber 2018). We use the first of the multiple imputations available to assign PEQs to establishments. For some analyses in this appendix, we classify establishments by the characteristics of the firm to which they belong. We measure firm size as the largest number of PEQs associated with all of that firm's establishments in any quarter in our period. We use this to define three strata of firm size: Ten or fewer workers; 11-276 workers; and greater than 276 workers. We further divide firms in the latter two size categories into those with just a single establishment and those with multiple establishments, yielding a total of five firm categories. Note that we use State Employer Identification Numbers (SEINs) to define firms, so strictly speaking a "firm" for our purposes is a firm-by-state combination.

(iv) Coding of geographic locations in LEHD

The Census Bureau assigns geographic locations at a highly granular level to workers' residences and establishment locations, at an annual frequency. We use the latitude and longitude of home and workplace compute the as-the-crow-flies commute distance for each worker, in miles. To analyze the number of jobs within a radius r of each worker, we coarsen the locations of firms and workers to a set of grid points spaced 0.5 miles apart in both the North-South and East-West locations. Commute distances computed based on this grid are extremely highly correlated with those that use the original uncoarsened locations, so we do not believe much precision is lost with this coarsening, but it dramatically reduces computational burden. Because CZs differ dramatically in their scales, to make multiple-CZ averages meaningful we standardize distances across CZs by rescaling so that the 75th percentile commute distance in each CZ equals 16 miles. So if the 75th percentile in a particular CZ is 12 miles, we multiply all distances by $4/3$, whereas if the 75th percentile is 24 miles, we multiply all distances by $2/3$.

D. AKM Model and the role of establishment pay premiums in the racial wage gap

(i) AKM Model

Using our LEHD sample for each CZ, we fit an AKM model with worker and establishment fixed effects and time-varying observables:

$$(A-1) \quad y_{it} = \alpha_i + \delta_{f(i,t)} + X_{it}\beta + \epsilon_{it}.$$

The variables included in X are a full set of calendar quarter indicators and a cubic in worker age. We estimate (A-1) separately for each CZ pooling Black and white workers but limiting to the largest connected set in the CZ (which typically includes well over 95% of PEQs in the CZ). We normalize the average pay premium ($\delta_{f(i,t)}$) of all firms in the restaurant industry in each CZ to zero. Thus, $\hat{\delta}_f$ can be interpreted as the average pay premium at establishment f relative to the average pay premium at restaurants in the same CZ.

Post-estimation we average the left-hand and right-hand sides of (1) by CZ and race, then take the difference between whites and Blacks in each CZ, yielding:

$$(A-2) \quad \bar{y}_{cw} - \bar{y}_{cb} = (\bar{\alpha}_{cw} - \bar{\alpha}_{cb}) + (\bar{\delta}_{cw} - \bar{\delta}_{cb}) + (\bar{X}_{cw} - \bar{X}_{cb})\hat{\beta},$$

where \bar{y}_{cw} and \bar{y}_{cb} represent the means of log earnings for white and Black workers in CZ c , respectively, $\bar{\alpha}_{cw}$ and $\bar{\alpha}_{cb}$ represent the means of the estimated person effects for white and Black workers in that CZ, $\bar{\delta}_{cw}$ and $\bar{\delta}_{cb}$ represent the means of the estimated establishment effects for the two groups, and \bar{X}_{cw} and \bar{X}_{cb} represent the mean vectors of covariates. This is equation (2) in the text.

Let s_{fcw} and s_{fcb} represent the shares of all PEQ's of white and Black workers in CZ c that worked at establishment f . Then

$$(A-3) \quad \bar{\delta}_{cw} - \bar{\delta}_{cb} = \sum_{f \in C} (s_{fcw} - s_{fcb}) \hat{\delta}_f.$$

Thus, the second term in equation (2) can be interpreted as measure of the differential sorting of whites relative to Blacks to workplaces with a higher estimated pay premium. If Black workers are less likely than whites to be employed at such workplaces, this term will be negative.

The lower rows of Table A-1 report the average values of the three terms on the right hand side of equation (A-2) for the three groups of larger CZs. Table 1 of the paper reports these components expressed as a percentage of the average Black-white gap in mean log earnings. Our estimated models imply that differences in the average pay premiums received

by Black workers relative to white workers account for close to zero of the racial wage gap, and that the differential sorting of Black and white workers to high-paying establishments is very small.

In Card et al. (2024) we present a simple decomposition of establishment pay premiums into the mean by industry, and the deviation of the establishment premium from the average for its industry, which we call a within-industry “hierarchy effect”. The bottom rows of Table A-1 use this approach to decompose the Black-white difference in mean establishment pay premiums into the difference in mean industry wage effects and the difference in mean hierarchy effects. Interestingly, for the two major CZ groups that are the focus of this paper these have opposite signs: Black workers work in slightly lower-paying industries than whites (particularly in the older industrial cities), but within a given industry they are employed at slightly higher-paying establishments.³

(ii) Interpretation

The fact that estimated average pay premiums for white and Black workers are nearly the same in our LEHD sample is surprising. Gerard et al. (2021) find that the under-representation of Black workers relative to whites at higher-paying workplaces in Brazil explains about 20% of the Black-white pay gap in that country. Moreover, our estimated AKM models, like the models in most other recent studies, imply that there is tendency for workers with higher values of α_i to work at establishments with higher pay premiums. In particular, our estimates imply that, other things equal, a 10% increase in α_i is associated with about a 1% increase in $\delta_{f(i,t)}$ within a CZ. Given the 30-35% gap in the mean of α_i between Black and white workers in our setting, one might have expected a roughly 3 log point gap in average pay premiums between Blacks and whites just because of assortative matching, rather than the 0 that we observe.

We do find that Black workers are slightly less likely to work in higher-paying industries, but this is offset by the tendency to be employed at higher-premium workplaces within a given

³ We also find a tendency for whites to work in higher-premium industries in the ACS. Using estimated premiums for 295 4-digit industries (discussed below) we find that the average industry premium is 3 log points higher for whites than Blacks.

industry. This pattern is potentially consistent with a longstanding fact about the U.S. labor market, which is that Black workers are more likely to be covered by unions than whites (e.g., Ashenfelter, 1972). Data from the unionstats.com website shows that the ratio of the Black to white union coverage rate for male workers was around 130% in the late 1970s, and averaged about 120% in years 2010-2018. Similarly, the relative coverage rate of Black females relative to white females was around 160% in the late 1970s and averaged about 120% in years 2010-2018.

One possibility that we are exploring in ongoing research is that Black workers are actually closer to higher-premium jobs than white workers in many CZs – rather than further away, as is suggested by the Spatial Mismatch hypothesis – and that this relative proximity offsets the tendency we would expect from assortative matching for Blacks to work at lower-paying jobs. It is an open question whether this can explain the interesting pattern of between- and within- industry wage differences we see in the LEHD.

E. Job access and commuting patterns by race

i) Job access

Figure 1 of the main paper shows the share of all jobs, or of all good jobs, within radius r of the average Black and white worker in each group of CZs. We interpret this as showing that Black workers are not systematically farther from jobs or from good jobs than are white workers. But this may be misleading about access to jobs if not all jobs are available to all workers. To try to assess this, we separate workers by their level of education, high school or less vs. college or more. We similarly separate jobs by whether they are held by high school or college workers, assuming that jobs are only available to workers of the same education level as their incumbent workers. For each education group, we compute the share of all jobs for that education group that are within radius r of the average white and Black worker of that education group. Results are shown in **Figure A-1**. The general story is similar to that seen in Figure 1 – in both the high school and college labor markets, Black workers tend to live closer to jobs, and closer to good jobs in particular, than do white workers.

ii) Commute distances and job quality

Figure A-2 shows the distribution of commute distances for Black and white workers in our two groups of CZs. This is constructed from estimates of the kernel density of log commutes, which we then convert to CDFs. **Table A-4** reports selected quantiles for various groups of CZs. For comparison, we also show the quantiles of 1-way commute times (not distances) from the ACS. The LEHD data on commute distances show that in the older industrial cities, Black workers have generally shorter commute distances than whites (e.g., a median distance of 7.8 miles for Blacks versus 9.5 miles for whites), whereas in the Sunbelt cities Black workers have longer commute distances at quantiles up to the median (e.g., a median commute distance of 9.9 miles for Blacks versus 9.2 miles for whites). Looking at commute times in the ACS, we see identical quantiles for Blacks and whites in the older industrial cities, but slightly higher commute times at the median for Black workers in the Sunbelt. Since Black workers are more likely than whites to use buses and subways, and these modes have longer average travel times, we interpret the ACS travel time data as being consistent with the LEHD travel distances.⁴

Figure 2 shows the relationship between commute distance and the establishment wage premium for Black and white workers in each group of CZs. One concern is that errors in the imputation of establishments within multi-establishment firms may create bias in this relationship. To assess this, we estimate the relationship separately for workers in five groups of firms – the smallest firms, with no more than 10 workers in any quarter; larger firms that are still below median in size, separately by whether they have one or multiple establishments; and above-median firms, again separately by whether they have one or multiple establishments. Any imputation error would affect only the workers at multi-establishment firms. **Figure A-3** shows that the relationship is generally similar for all groups of firms.

However, the decline in establishment premiums for workers with the longest commutes is seen only among those at multi-establishment firms. This is consistent with the

⁴ In the 30 largest CZs 88% of whites commute by car versus 81% of Blacks. Mean commute times are about 20 minutes longer for commuters who use buses or other transit modes (except walking) relative to car.

hypothesis that it might be spurious, due to imputation errors. The establishment imputation model does not take account of establishment pay premiums, and thus will not reflect that workers may be more likely to commute past one establishment to another when the farther establishment offers better pay.

Another interesting relationship that our framework lets us examine is the relationship between worker skill and commute distance. This is *ex ante* ambiguous – commute time has higher opportunity cost for high ability workers, but these workers may consume some of their higher income via more housing space, requiring a move to the suburbs. **Figure A-4** shows the distribution of commuting time by worker type (deciles of α_i), race, and CZ grouping. In both groups of CZs, high-skill workers commute modestly farther than do low-skill workers, though the dispersion within decile is quite large. In the older industrial CZs, the white commute distance distribution stochastically dominates the Black distribution in each ability decile, except possibly among the lowest-skill workers. In the newer Sunbelt CZs, Black workers tend to commute farther than similarly-skilled white workers, but this reverses for the lowest-skilled workers at the highest commute lengths.

Table A-5 presents the estimated elasticities of earnings, and the components of earnings attributable to person effects and workplace pay premiums, with respect to commute distance by CZ group and race. The elasticity of earnings is slightly higher for whites than Blacks in the older industrial cities, but lower for whites than Blacks in the Sunbelt cities and in the remainder group of CZs. Interestingly, however, when we look separately at the parts of earnings attributable to worker skills and establishment wage premiums, we see that the elasticity of the personal skills component is uniformly higher for Black workers, whereas the elasticity of workplace premiums is uniformly lower for Black workers. This suggests that the positive relationship between potential earnings capacity and commute distance is stronger for Blacks than whites, but that access to better-paying jobs, conditional on worker skills, is if anything better for Black workers (consistent with the main findings in our paper).

Table A-6 presents a parallel set of estimates based on our ACS samples. Since the ACS only reports the average (one-way) time taken for commuting, we include a set of dummies for the mode of transit. The elasticities of earnings with respect to commute time range from 0.08

to 0.10 for whites and 0.06 to 0.07 for Blacks. In three of the four city groups we estimate that the elasticity is *lower* for Blacks than whites, though in the Sunbelt cities the elasticity is very slightly higher for Black workers (0.071 versus 0.069).

We cannot decompose earnings in the ACS into person and workplace effects. As an alternative, we estimated a relatively rich cross-sectional wage model (separately by race) that included 295 4-digit industry effects. This allows us to decompose an individual earnings observation into a part attributable to the industry of employment, a part due to other observed covariates, and an unexplained part. We then regressed the industry component on commute time and obtained the set of elasticities shown in the second row of each panel in Table A-6. For both Blacks and whites we estimate that longer commute times are associated with employment in higher-paying industries: the elasticities are in the range of 0.024-0.028 for whites and 0.018 to 0.026 for Blacks – not too different from the elasticities of workplace pay premiums with respect to travel distance we obtained in the LEHD. Again, the elasticity of industry pay premiums tends to be slightly lower for Blacks, suggesting that if anything Black workers have slightly *better access* to higher-paying industries, except in the Sunbelt cities where whites and Blacks have similar access.

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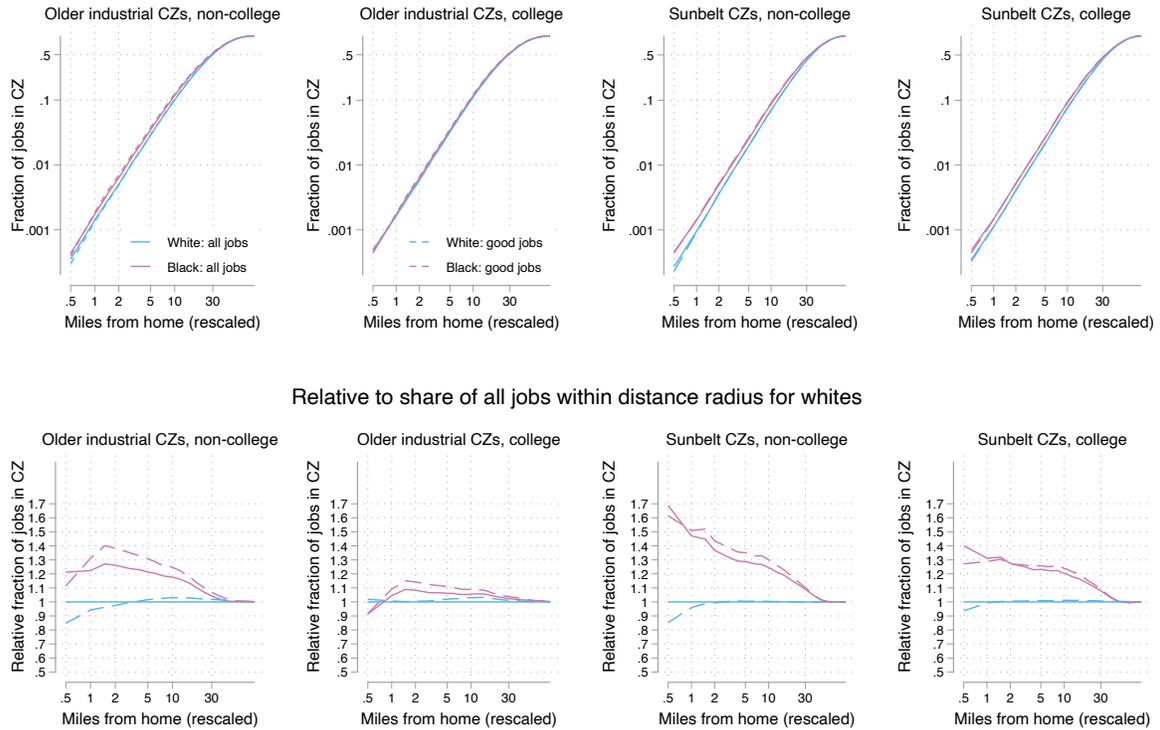
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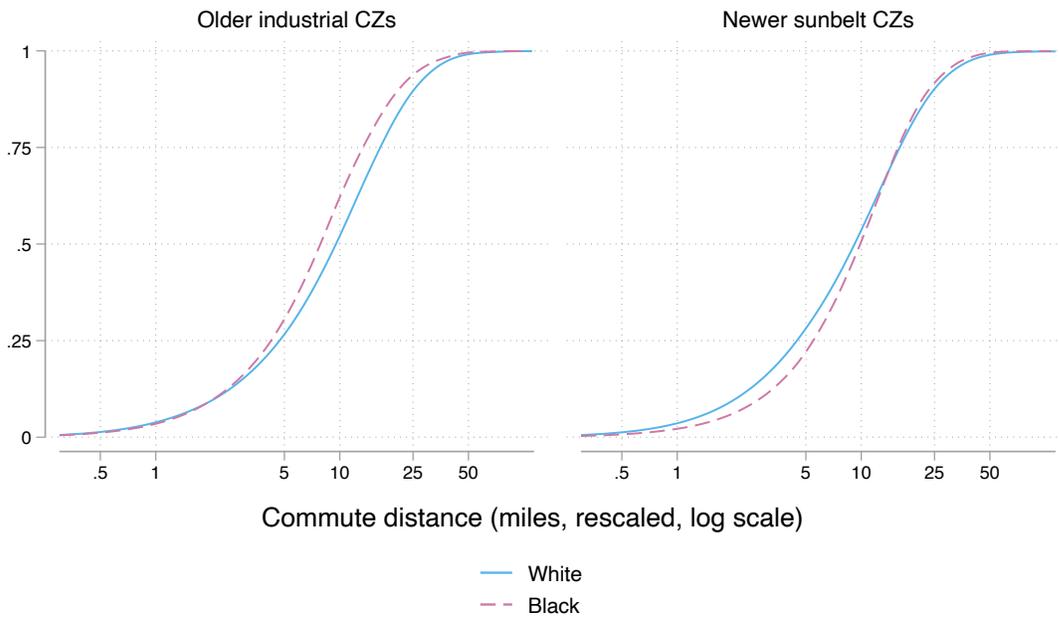
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Figure A-1. Job access in college and non-college labor markets



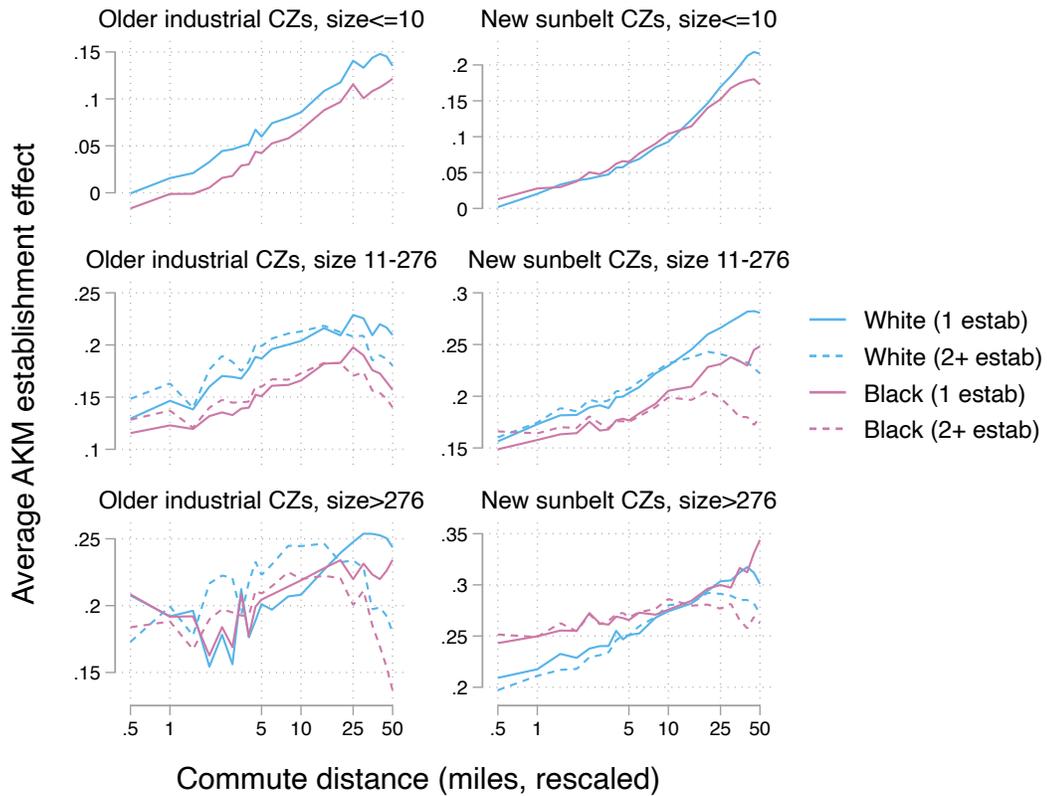
Notes: Distances for each CZ are rescaled to set the 75th percentile commute distance to 16 miles. “Good jobs” are those at establishments with AKM establishment effects in the top tercile. Bottom panels show fraction of jobs in a CZ relative to fraction of jobs available within distance radius for whites with the same level of education.

Figure A-2. CDFs of commute distribution, by CZ group and race



Notes: We estimate kernel densities of log standardized commute time using pooled samples across CZs in each group, separately by race. Commute distances are standardized to set the 75th percentile commute distance in each CZ to 16 miles. Kernel density estimates use an Epanechnikov kernel and Stata's default bandwidth for the white sample, and are evaluated at each multiple of 0.1 miles up to 10 miles, and then at each mile up to 100 miles. We linearly interpolate the estimated densities and construct CDFs from them.

Figure A-3. AKM establishment effect and commute distance, by firm type, CZ group, and race



Notes: Commute distances are standardized to set the 75th percentile commute distance in each CZ to 16 miles. Rows categorize firms by the maximum number of workers observed at the firm in any quarter; within each panel, series distinguish firms with a single establishment vs. more than one.

Figure A-4. Commute distances by AKM worker effect decile, race, and CZ group

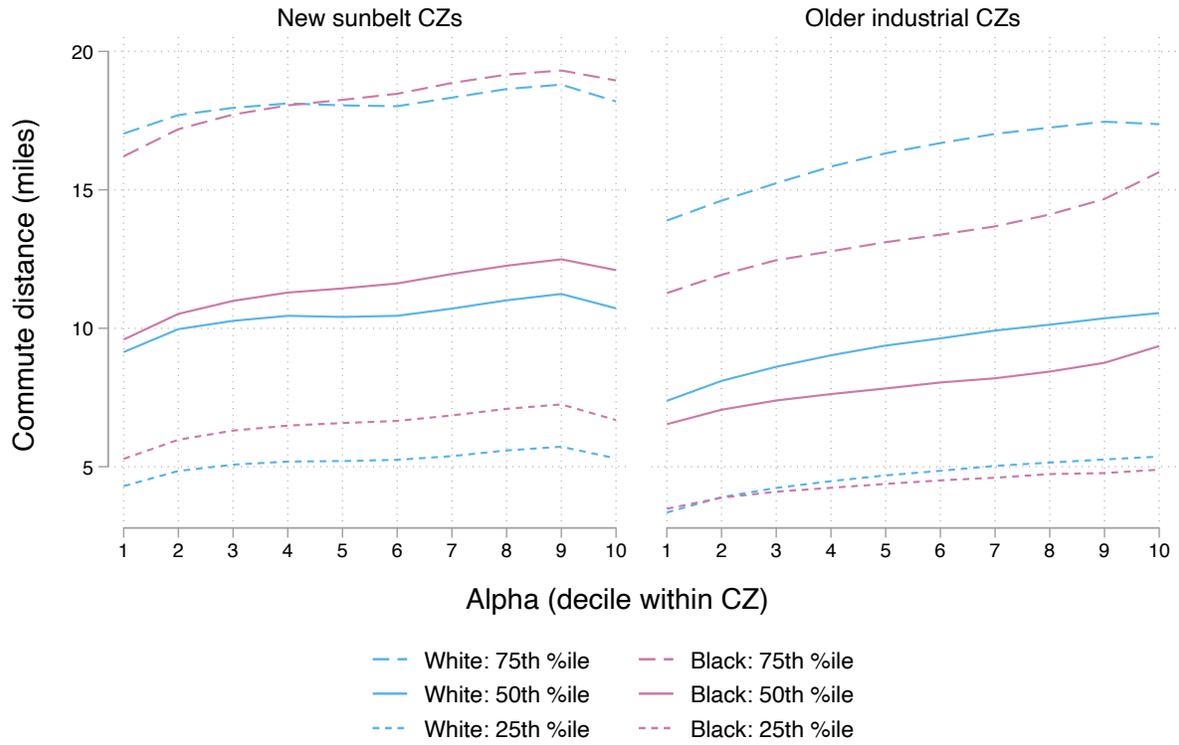


Table A-1. LEHD summary statistics and AKM decomposition

| | Older industrial CZs | Newer sunbelt CZs | All other CZs among largest 200 |
|-----------------------------------|----------------------------|-------------------------|---------------------------------------|
| <u>All workers</u> | | | |
| Geometric mean earnings | 13,521 | 14,458 | 11,673 |
| SD of log earnings | 0.64 | 0.69 | 0.58 |
| No. of person-quarters (millions) | 311.4 | 203.1 | 1077.0 |
| <u>White workers</u> | | | |
| Geometric mean earnings | 14,271 | 15,835 | 12,064 |
| No. of person-quarters (millions) | 262.1 | 156.6 | 931.7 |
| <u>Black workers</u> | | | |
| Geometric mean earnings | 10,067 | 10,993 | 9,396 |
| No. of person-quarters (millions) | 49.2 | 46.5 | 145.3 |
| <u>White-Black gap</u> | | | |
| Log earnings | 0.35 | 0.36 | 0.25 |
| Components of AKM decomposition | | | |
| Person effect | 0.37 | 0.38 | 0.28 |
| Establishment effect | 0.00 | 0.00 | -0.01 |
| Xb | -0.01 | -0.02 | -0.01 |
| Decomposition of estab. effects | | | |
| Industry effect | 0.02 | 0.01 | 0.00 |
| Within-industry effect | -0.02 | -0.01 | -0.01 |

Notes: Source is 2010-2018 LEHD. Sample includes only white and Black non-Hispanic people with quarterly earnings above \$3,800, and excludes the first and last quarter of each employment spell. Person effects and establishment effects are obtained from AKM model (equation 1) with controls for age (cubic) and calendar quarter (indicators).

Table A-2. Summary Statistics for Four Groups of Larger CZ's

| | Top 30 CZ's | Older Industrial | Newer Sunbelt | Northeast Corridor | Remainder (mostly West) |
|---|----------------|---------------------|------------------|-----------------------|----------------------------|
| Share of Obs (%) | 100.0 | 28.6 | 32.4 | 19.0 | 20.0 |
| <u>Demographics of Working Age Population</u> | | | | | |
| White non-Hispanic (%) | 52.0 | 64.1 | 38.7 | 51.9 | 56.4 |
| Black non-Hispanic (%) | 13.7 | 16.6 | 13.2 | 16.7 | 7.5 |
| Hispanic (%) | 22.6 | 11.3 | 36.3 | 19.0 | 20.0 |
| Asian non-Hispanic (%) | 11.8 | 8.1 | 11.8 | 12.5 | 16.3 |
| Mean Years of Education | 13.6 | 13.8 | 13.1 | 14.0 | 13.7 |
| BA or higher (%) | 36.3 | 37.1 | 31.2 | 43.2 | 36.8 |
| Employed (%) | 80.5 | 81.0 | 79.0 | 81.8 | 81.0 |
| Full-time Earner (%) | 61.8 | 63.3 | 58.4 | 64.7 | 62.2 |
| Mean Earnings (with 0's) | 41,976 | 42,197 | 36,913 | 49,322 | 42,878 |
| <u>Characteristics of Full Time Wage and Salary Earners</u> | | | | | |
| Black non-Hispanic (%) | 12.3 | 13.6 | 12.7 | 15.7 | 6.6 |
| Female | 45.1 | 46.0 | 43.7 | 46.8 | 44.4 |
| Mean Years of Education | 14.2 | 14.4 | 13.8 | 14.6 | 14.2 |
| BA or higher (%) | 44.4 | 45.1 | 39.2 | 51.7 | 44.2 |
| Mean Earnings | 66,064 | 64,812 | 60,986 | 74,618 | 67,170 |
| Mean Hourly Wage | 31.33 | 30.62 | 29.14 | 35.09 | 31.99 |
| Use Car to Commute (%) | 86.0 | 88.4 | 94.3 | 67.2 | 88.8 |
| Mean Commute Time (mins) | 30.9 | 29.9 | 30.0 | 35.3 | 29.4 |
| <u>Earnings of Full Time Wage and Salary Workers</u> | | | | | |
| Mean Log Earnings - White NH | 10.96 | 10.90 | 10.97 | 11.09 | 10.93 |
| Mean Log Earnings - Black NH | 10.64 | 10.61 | 10.61 | 10.74 | 10.61 |
| Black-white Gap | -0.32 | -0.30 | -0.36 | -0.35 | -0.32 |

Source: 2010-2018 ACS public use files. Adult population includes people age 22-62 with age> education+6. Full time earners have annual earnings above \$15,200. Older industrial CZs are Philadelphia, Detroit, Pittsburgh, Cleveland, Newark NJ, Buffalo, Baltimore, Chicago, Minneapolis, and St. Louis. Newer Sunbelt CZs are Los Angeles, Houston, Atlanta, Miami, San Diego, Phoenix, and Dallas. Northeast Corridor CZs are New York, Washington DC, Boston, and Hartford. Remaining CZs are San Francisco, Seattle, Denver, Sacramento, San Jose, Portland, Tampa, Orlando and Fort Worth.

Table A-3. Characteristics of Commuting Zones (CZ's) in Four Major Groups of Larger CZ's

| | Working Age Population x1000 | Working Age Adults (22-62 with Positive Experience) | | | | | BA or more | Full Year Earner | Full Year Earners Only | |
|------------------------------------|---------------------------------|---|----------|----------|----------|-----------|---------------|---------------------|--------------------------------|-----------------------|
| | | White NH | Black NH | Hispanic | Asian NH | Immigrant | | | Annual Wage & Sal. Earnings | One-way Comm. Time |
| <u>Older Industrial Cities:</u> | | | | | | | | | | |
| Chicago | 4,764 | 53.4 | 16.4 | 21.5 | 8.8 | 25.6 | 38.9 | 63.7 | 66,155 | 33.3 |
| Newark | 3,429 | 48.6 | 13.8 | 23.4 | 14.2 | 36.6 | 42.3 | 65.0 | 75,162 | 34.2 |
| Philadelphia | 3,237 | 63.2 | 19.5 | 9.4 | 7.9 | 14.0 | 36.3 | 62.4 | 65,846 | 30.7 |
| Detroit | 2,827 | 69.2 | 20.8 | 3.8 | 6.2 | 11.1 | 30.9 | 57.2 | 60,445 | 27.8 |
| Minneapolis | 1,871 | 77.7 | 7.7 | 5.3 | 9.3 | 14.2 | 41.8 | 70.1 | 65,041 | 25.8 |
| Baltimore | 1,517 | 57.8 | 29.1 | 5.2 | 7.9 | 13.2 | 39.0 | 66.7 | 67,591 | 31.9 |
| Cleveland | 1,377 | 74.2 | 18.0 | 3.6 | 4.2 | 6.6 | 31.5 | 61.1 | 56,743 | 25.2 |
| St Louis | 1,336 | 73.4 | 19.4 | 2.6 | 4.6 | 6.5 | 34.9 | 63.6 | 59,286 | 26.2 |
| Pittsburgh | 1,334 | 87.1 | 7.7 | 1.5 | 3.7 | 4.2 | 35.4 | 63.1 | 57,925 | 28.0 |
| Buffalo | 1,257 | 79.3 | 10.9 | 5.2 | 4.7 | 7.4 | 32.7 | 62.3 | 54,583 | 22.2 |
| <u>Newer Sunbelt Cities:</u> | | | | | | | | | | |
| Los Angeles | 10,272 | 32.6 | 6.5 | 44.6 | 16.4 | 40.3 | 29.1 | 55.7 | 61,037 | 31.0 |
| Houston | 3,446 | 38.0 | 17.3 | 35.3 | 9.4 | 32.4 | 30.2 | 59.8 | 64,352 | 30.3 |
| Atlanta | 2,822 | 46.5 | 34.8 | 10.5 | 8.2 | 20.1 | 38.3 | 61.7 | 62,790 | 32.3 |
| Dallas | 2,562 | 46.0 | 16.6 | 27.7 | 9.7 | 28.4 | 34.8 | 63.7 | 63,323 | 28.8 |
| Miami | 2,577 | 24.2 | 20.3 | 51.6 | 3.9 | 53.1 | 29.7 | 57.1 | 53,723 | 30.2 |
| Phoenix | 2,421 | 57.5 | 5.3 | 29.0 | 8.2 | 20.2 | 28.4 | 59.4 | 57,027 | 27.0 |
| San Diego | 1,826 | 47.7 | 5.0 | 31.5 | 15.8 | 30.6 | 35.1 | 59.8 | 62,988 | 26.0 |
| <u>Northeast Corridor Cities:</u> | | | | | | | | | | |
| NYC | 6,993 | 41.6 | 18.3 | 25.5 | 14.6 | 40.5 | 39.5 | 60.8 | 73,448 | 39.2 |
| Washington DC | 3,296 | 45.9 | 25.4 | 15.1 | 13.5 | 30.3 | 50.4 | 70.9 | 78,587 | 35.4 |
| Boston | 2,967 | 72.2 | 7.3 | 10.4 | 10.1 | 23.0 | 46.8 | 67.0 | 73,401 | 32.0 |
| Hartford | 1,946 | 68.3 | 10.1 | 14.9 | 6.7 | 18.5 | 39.0 | 64.9 | 73,128 | 27.1 |
| <u>Remaining Cities in Top 30:</u> | | | | | | | | | | |
| San Francisco | 2,981 | 41.6 | 7.8 | 21.6 | 29.1 | 36.8 | 45.5 | 63.5 | 81,992 | 33.1 |
| Seattle | 2,616 | 68.9 | 5.1 | 8.7 | 17.4 | 20.1 | 37.8 | 64.8 | 67,805 | 30.2 |
| Denver | 1,765 | 68.4 | 4.9 | 20.1 | 6.7 | 15.8 | 42.7 | 67.0 | 65,303 | 27.6 |
| Sacramento | 1,711 | 52.1 | 6.7 | 23.6 | 17.6 | 25.3 | 27.3 | 55.3 | 60,366 | 29.4 |
| Tampa | 1,571 | 64.3 | 11.7 | 18.3 | 5.7 | 17.0 | 28.8 | 59.4 | 54,652 | 27.8 |
| San Jose | 1,488 | 35.5 | 2.4 | 31.1 | 31.0 | 46.3 | 43.4 | 62.5 | 83,949 | 27.6 |
| Ft. Worth | 1,311 | 55.7 | 13.4 | 24.2 | 6.7 | 19.8 | 28.0 | 62.3 | 57,988 | 28.6 |
| Orlando | 1,326 | 50.4 | 15.2 | 27.8 | 6.6 | 21.9 | 29.8 | 58.9 | 51,650 | 28.5 |
| Portland | 1,286 | 75.5 | 2.8 | 10.4 | 11.3 | 16.8 | 37.1 | 62.3 | 61,163 | 26.7 |

Source: 2010-2018 ACS Public Use Files. Working age population includes people 22-62 with positive experience. Full year earnings have at least \$15,200 in annual wage and salary earnings. Size of working age population is based on average weighted count of ACS sample in 2010-2018. Commuting zones are based on 1990 CZ definitions.

Table A-4. Quantiles of commute distance or commute time by CZ group and race

| | Older Industrial CZs | | Newer Sunbelt CZs | | Northeast Corridor CZs | Remaining CZs |
|--|----------------------|------------------|-------------------|------------------|---------------------------|------------------|
| | Miles (LEHD) | Minutes (ACS) | Miles (LEHD) | Minutes (ACS) | Minutes (ACS) | Minutes (ACS) |
| <u>Percentiles for White Non-Hispanics</u> | | | | | | |
| 10 | 2.1 | 10 | 2.1 | 10 | 10 | 10 |
| 25 | 4.7 | 15 | 4.5 | 15 | 15 | 15 |
| 50 | 9.5 | 25 | 9.2 | 25 | 30 | 25 |
| 75 | 16.6 | 40 | 16.2 | 40 | 45 | 40 |
| 90 | 25.3 | 60 | 24.9 | 60 | 60 | 55 |
| <u>Percentiles for Black Non-Hispanics</u> | | | | | | |
| 10 | 2.1 | 10 | 2.8 | 10 | 15 | 10 |
| 25 | 4.2 | 15 | 5.5 | 15 | 20 | 15 |
| 50 | 7.8 | 25 | 9.9 | 30 | 30 | 25 |
| 75 | 13.4 | 40 | 15.9 | 40 | 55 | 40 |
| 90 | 20.9 | 60 | 23.3 | 60 | 70 | 60 |

Notes: Miles represent distances from home to work, from LEHD. Distances are standardized across CZs to set the 75th percentile commute distance in each CZ to 16 miles. Minutes represent commute times, from ACS.

Table A-5: Elasticity of Earnings and Earnings Components w.r.t. Commute Distance

| | Older Industrial | Newer Sunbelt | Remainder (Top 200) |
|---------------------------|---------------------|------------------|------------------------|
| <u>White non-Hispanic</u> | | | |
| Log Quarterly Earnings | 0.059 (0.000) | 0.024 (0.000) | 0.031 (0.000) |
| Person Effects | 0.027 (0.000) | 0.001 (0.000) | 0.007 (0.000) |
| Firm Premium | 0.030 (0.000) | 0.024 (0.000) | 0.023 (0.000) |
| <u>Black non-Hispanic</u> | | | |
| Log Quarterly Earnings | 0.056 (0.000) | 0.048 (0.000) | 0.056 (0.000) |
| Person Effects | 0.035 (0.000) | 0.026 (0.000) | 0.032 (0.000) |
| Firm Premium | 0.016 (0.000) | 0.019 (0.000) | 0.020 (0.000) |

Source: 2010-2018 LEHD. Sample includes only people age 22-62 with quarterly earnings above \$3,800. Person effects and firm premium represents estimated person and establishment earning effect obtained from AKM model (equation 1) with controls for age and calendar quarter. Coefficient estimates in table are obtained from specifications that regress log of quarterly earnings and AKM components on log of commute distance with CZ controls.

Table A-6: Elasticity of Annual Earnings w.r.t. Commute Time

| | Top 30 CZs | Older Industrial | Newer Sunbelt | Northeast Corridor | Remainder (mainly West) |
|--------------------------------------|------------------|---------------------|------------------|-----------------------|----------------------------|
| <u>White non-Hispanic</u> | | | | | |
| Log Annual Earnings | 0.090 (0.001) | 0.103 (0.001) | 0.070 (0.002) | 0.099 (0.002) | 0.079 (0.002) |
| Industry Premium (295 industries) | 0.028 (0.001) | 0.030 (0.001) | 0.026 (0.001) | 0.028 (0.001) | 0.027 (0.001) |
| <u>Black non-Hispanic</u> | | | | | |
| Log Annual Earnings | 0.069 (0.002) | 0.063 (0.003) | 0.077 (0.003) | 0.066 (0.003) | 0.069 (0.005) |
| Industry Premium (295 industries) | 0.023 (0.001) | 0.021 (0.001) | 0.028 (0.001) | 0.021 (0.001) | 0.023 (0.002) |

Source: 2010-2018 ACS public use files. Sample includes only people age 22-62 with positive experience (age-education>6) and annual earnings above \$15,200. Industry premium represents estimated industry wage effect received by worker, obtained from model fit by gender to all 30 of the largest CZ's, with controls for education, experience, race, immigrant status and CZ effects. Coefficient estimates in table (with robust standard errors) are obtained from specifications that regress log of annual earnings on log of commute time with controls for gender, mode, and CZ.