

**Are City Centers Losing Their Appeal?**  
**Commercial Real Estate, Urban Spatial Structure, and Covid-19**

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## Introduction

- This paper considers how the COVID-19 shock has affected cities.
- In order to do so, we obtain some pre-COVID results that speak to the more general issues of agglomeration and urban spatial structure.
- The paper will also have implications for commercial real estate (CRE), a really important and not very well-understood asset class.

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# Why do cities exist?

- Because density is valuable
  - Natural advantage: good locations.
  - Internal scale economies: factory towns.
  - External scale economies / agglomeration economies.
- The same forces explain why we have agglomeration within cities into central business districts (CBDs).
- Equilibrium agglomeration depends on the tradeoff between the benefits of agglomeration and the costs

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# COVID-19

- COVID-19 changes the agglomeration cost-benefit calculus
- Increased costs: disease cost, ventilation, elevators, subways. It is hard to physically distance in an office or retail setting.
- Decreased benefits: empty offices and social distance restrictions reduce interaction; working-from-home has become a viable alternative
- This leads to the big COVID-19 question...

Are city centers losing their attraction to businesses because of COVID-19?

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# Our approach to answering the big question

- Focus on commercial establishments in the retail and office sectors
- We estimate three spatial patterns of rent within cities that address the following questions:
  - How quickly do commercial rents decline with distance to the city center (i.e, commercial rent gradients)?
  - How much higher are commercial rents in high employment density locations?
  - How quickly do commercial rents decline with distance to a rapid transit station (e.g., a subway stop)?

Has COVID-19 affected the answers to these questions?

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## Results preview: Pre-COVID

- Results different between “transit” and “car” cities.
- Downward sloping commercial rent functions, steeper for transit cities.
- Employment density rent premium, larger for transit cities
- Transit station proximity premium.
- All of these show the value assigned to density, broadly conceived, pre-COVID



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## Results preview: Post-COVID

- The commercial rent gradient becomes smaller for transit cities, not car cities.
- The employment density premium becomes smaller for both transit cities and car cities.
- Transit station access premium falls.
- These are consistent with (a) a reduction in the value of city centers, but (b) value remains and (c) effects are heterogeneous

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## Related issues

- Housing affordability
- Productivity

# Theory

- Rent is determined by bidding among potential tenants:
  - Gross profit depends on market interactions:  $\pi(n)$
  - Market interactions depend on distance to CBD:  $n(d)$ ,  $\partial n/\partial d < 0$ .
  - Rent sets profit equal to zero:  $r(n) = \pi(n)$
- COVID-19 impacts both the  $\pi(n)$  and  $n(d)$  relationships, which will be reflected in  $r(n)$ .

# Data

- Data on over 68,000 leases from CompStak.
  - Street address, latitude/longitude
  - Lease type (e.g. new tenant versus renewal); Industry type (e.g. retail versus commercial)
  - Lease term, space leased, date lease executed
  - Effective rent (adjusted for upfront concessions like free months rent)
- Additional data were merged in from local planning authorities and Census. This provides information on ...
  - 2018 employment density in the zipcode in which a lease is located
  - Distance to the closest rapid transit stop.

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## Data

- All leases were executed from January, 2019 through October 31, 2020
- Median lease length is 57 months → Commercial rent encompasses expectations of future value associated with a site, not just current conditions.

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## Organizing leases into cities

- In order to evaluate whether city centers are losing their appeal, we need to specify well-defined centers.
- But cities are multicentric, and for large urban areas, often ringed with important sub-centers.
- To address this, we organize our data into pseudo-monocentric cities using an iterative approach.

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## Organizing leases into core cities

- Step 1: Pool all leases across the country. Pick out the lease in the zipcode with the highest employment density.
- Step 2: Draw a circle of radius 25 miles around the target zipcode centroid. Assign all leases within that circle to the city in which the target zipcode is located.
- Step 3: Repeat steps 1 and 2 leaving out the previously assigned leases. Repeat this until all leases are assigned to a city.
- Step 4: Using the cities defined in Step 3 (89 cities in our data), reassign each lease to the closest city center.
- Step 5: We excluded leases in core cities with  $< 100$  assigned leases

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## Car and transit cities

- Core cities are grouped into two groups. Those that rely heavily on rapid transit (in six MSAs) and all others
- We refer to the first group as transit cities and the second group as car cities
- Transit cities are in the following MSAs
  - NYC, Washington DC, Chicago, Boston, San Francisco, Philadelphia
- Transit cities are populous, dense, and expensive compared to car cities

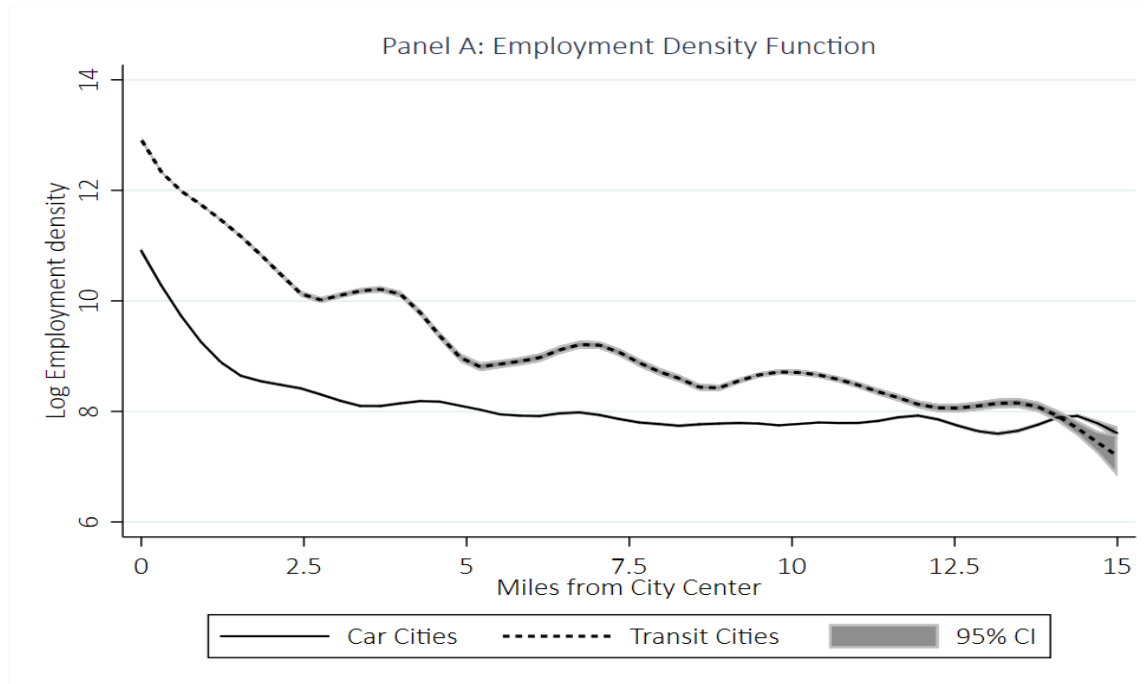


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## Does employment density decline with distance?

- If our iterative approach to grouping leases into cities is successful, we should have created pseudo-monocentric cities.
- In monocentric cities, both rents and density depend on location.
- Is this the case?
- Are transit and car cities different in their density profiles?

# Density declines with distance



**Panel A – Employment density gradient: Dep var = Log zipcode employment density<sup>b</sup>**

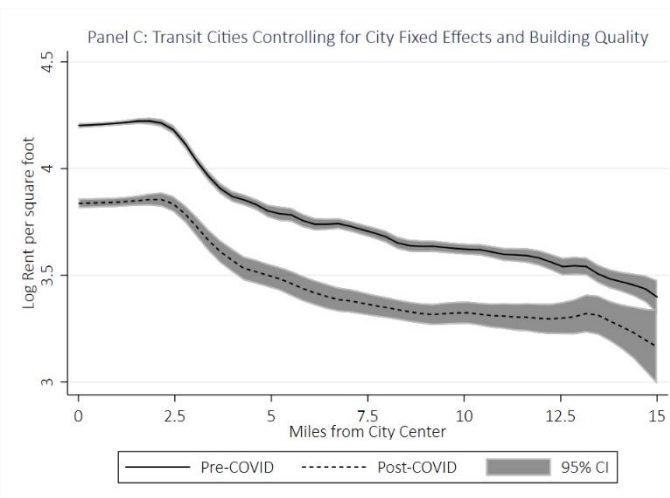
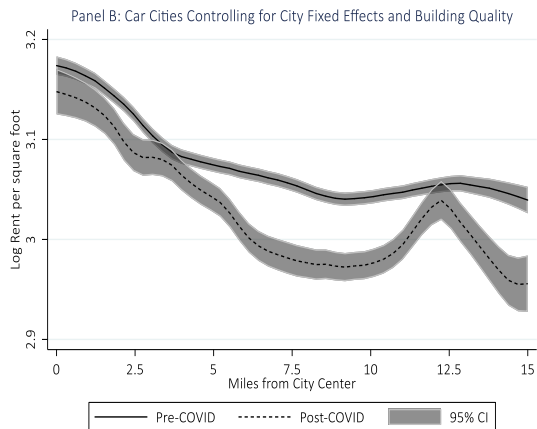
	All Cities	Car Cities <sup>c</sup>	Transit Cities <sup>c</sup>
Distance (miles) to CBD ( $D_{CBD}$ )	-0.1989	-0.1372	-0.3783
	(-8.59)	(-10.14)	(-18.26)
Core city fixed effects <sup>d</sup>	89	83	6
Observations	63,886	48,590	15,296
R-squared	0.336	0.225	0.661

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## Density gradient is larger in transit cities

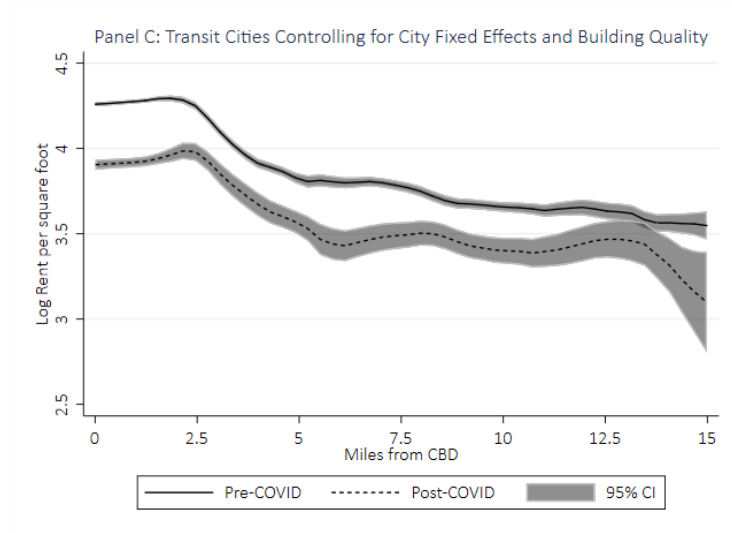
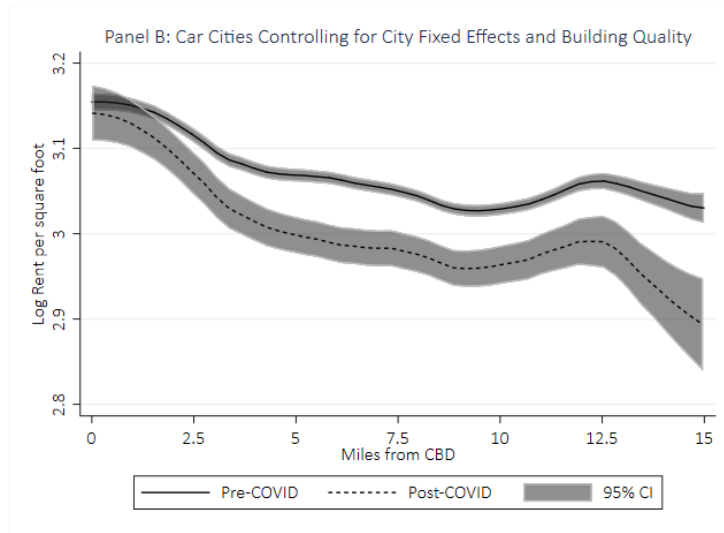
- The density gradient is larger in the transit cities: 38% versus 14%
- Theory suggests that the rent gradient should also be larger in transit cities

# Rent declines more rapidly in transit cities



<b>Panel B – Distance to CBD: Dep var = Log rent<sup>b</sup></b>	<b>All Cities</b>	<b>Car Cities<sup>c</sup></b>	<b>Transit Cities<sup>c</sup></b>
Post Covid (April 1 – Oct 31, 2020) <sup>d</sup>	-0.0395 (-1.47)	0.0007 (0.04)	-0.0858 (-2.32)
Distance (miles) to CBD ( $D_{CBD}$ )	-0.0227 (-3.54)	-0.0092 (-2.46)	-0.0633 (-5.00)
$D_{CBD}$ * Post Covid	0.0031 (0.96)	-0.0016 (-0.75)	0.0094 (2.20)
Core city fixed effects <sup>d</sup>	89	83	6
Observations	68,638	52,490	16,148
R-squared	0.157	0.142	0.246

# But COVID-19 reduces the gradient in transit cities



## Panel B – Distance to CBD: Dep var = Log rent<sup>b</sup>

Post Covid (April 1 – Oct 31, 2020)<sup>d</sup>

Distance (miles) to CBD ( $D_{CBD}$ )

$D_{CBD} * \text{Post Covid}$

Core city fixed effects<sup>d</sup>

Observations

R-squared

	All Cities	Car Cities <sup>c</sup>	Transit Cities <sup>c</sup>
Post Covid (April 1 – Oct 31, 2020) <sup>d</sup>	-0.0057 (-0.19)	0.0400 (2.19)	-0.0662 (-2.30)
Distance (miles) to CBD ( $D_{CBD}$ )	-0.0212 (-3.12)	-0.0088 (-2.19)	-0.0631 (-4.21)
$D_{CBD} * \text{Post Covid}$	0.0020 (0.53)	-0.0030 (-1.32)	0.0116 (2.18)
Core city fixed effects <sup>d</sup>	109	102	7
Observations	53,092	40,838	12,254
R-squared	0.139	0.148	0.237

# Rent also increases with local density but once again, more so in Transit cities pre-COVID

<b>Panel C – Valuing employment density: Dep var = Log rent<sup>b</sup></b>	<b>All Cities</b>	<b>Car Cities<sup>c</sup></b>	<b>Transit Cities<sup>c</sup></b>
Post Covid (April 1 – Oct 31, 2020) <sup>d</sup>	0.1280 (2.61)	0.1021 (1.93)	0.1194 (2.98)
Employment per square foot ( $D_{EmpDen}$ )	0.0835 (5.34)	0.0455 (6.77)	0.1338 (4.82)
$D_{EmpDen}$ * Post Covid	-0.0177 (-3.07)	-0.0143 (-2.45)	-0.0169 (-3.73)
Core city fixed effects <sup>a</sup>	89	83	6
Observations	68,638	52,490	16,148
R-squared	0.157	0.142	0.246

# All cities see a weakening of the local density-rent relationship post-COVID

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## So far...

- Flatter rent functions in transit cities post-COVID.
- Smaller local employment rent premium post-COVID.
- Overall, we see
  - The attraction of downtown is weakened but ...
  - Not everywhere and...
  - Some attraction remains.



# Extension for Transit Cities: Are the post-COVID patterns simply driven by retail? No!

	(1)	(2)	(3)	(4)
	Retail	Office	Retail	Office
Post Covid (April 1 – Oct 31, 2020) <sup>d</sup>	-0.2673 (-3.19)	-0.0444 (-2.04)	0.1154 (0.62)	0.0734 (1.49)
Distance (miles) to CBD ( $D_{CBD}$ )	-0.0942 (-3.11)	-0.0571 (-8.20)	-	-
$D_{CBD}$ * Post Covid	0.0260 (2.56)	0.0057 (2.40)	-	-
Employment per square foot ( $D_{EmpDen}$ )	-	-	0.2546 (4.39)	0.1164 (7.60)
$D_{EmpDen}$ * Post Covid	-	-	-0.0230 (-1.22)	-0.0099 (-2.19)
Core city fixed effects	6	6	6	6
Observations	2,772	12,524	2,889	13,259
R-squared	0.246	0.361	0.310	0.326

Both retail and the office sectors are strongly affected by COVID-19

# Extension for Transit Cities: New arrival lease versus lease renewal

	(5)	(6)	(7)	(8)
	New Lease	Renewal Lease	New Lease	Renewal Lease
Post Covid (April 1 – Oct 31, 2020) <sup>d</sup>	-0.0589 (-1.48)	-0.1197 (-2.51)	0.0674 (1.26)	0.1757 (2.35)
Distance (miles) to CBD ( $D_{CBD}$ )	-0.0655 (-4.55)	-0.0617 (-5.38)	-	-
$D_{CBD}$ * Post Covid	0.0057 (1.53)	0.0118 (2.22)	-	-
Employment per square foot ( $D_{EmpDen}$ )	-	-	0.1364 (4.44)	0.1319 (5.27)
$D_{EmpDen}$ * Post Covid	-	-	-0.0098 (-2.17)	-0.0250 (-3.01)
Core city fixed effects	6	6	6	6
Observations	6,438	8,858	6,798	9,350
R-squared	0.269	0.256	0.256	0.242

Pre-COVID new and renewal estimates are very similar

# Extensions for Transit Cities: New arrival lease versus lease renewal

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Core city fixed effects	6	6	6	6
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Post-COVID estimates: for renewals, rent discounts for “known, safer” tenants

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# Why are COVID-19 effects larger in transit cities?

- Some possibilities include ...
  - Transit cities are big and dense, making them more vulnerable.
  - Transit cities were hit earlier and harder by COVID-19. Possibly this contributed to more cautious social behavior and stricter lockdowns.
  - Transit cities are culturally different, possibly implying different behavior and consequent differences in the spread of COVID-19
  - Transit cities also rely heavily on rapid transit which exposes riders to more risk than car travel.

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# COVID-19 and the built environment

- Has COVID-19 reduced the premium for locations close to a rapid transit station?  
  
→ Estimate rent function with distance from transit stop pre- and post-COVID.

## Partial linear regression

- Log rent depends linearly on location fixed effects  $\beta_j$  and  $X$ , and nonlinearly on  $z$  as  $m(z)$

$$\text{Log Rent} = \beta_j + \beta_1 X + m(z) + \varepsilon$$

- We estimate  $m(z)$  using the semipar routine in Stata which uses Robinson's (1988) double error approach.

## Partial linear regression

- Write the partially linear model as:  $y = X\beta + m(z) + \varepsilon$
- Take expected values conditioning on  $z$ :  $E(y|z) = E(X|z)\beta + m(z) + E(\varepsilon|z)$
- Differencing  $\rightarrow y - E(y|z) = (X - E(X|z))\beta + \varepsilon$
- By estimating  $E(y|z)$  and  $E(X|z)$  nonparametrically and replacing them in the above equation, it is possible to estimate  $\beta$  consistently without modelling  $m(z)$  since it differences out.

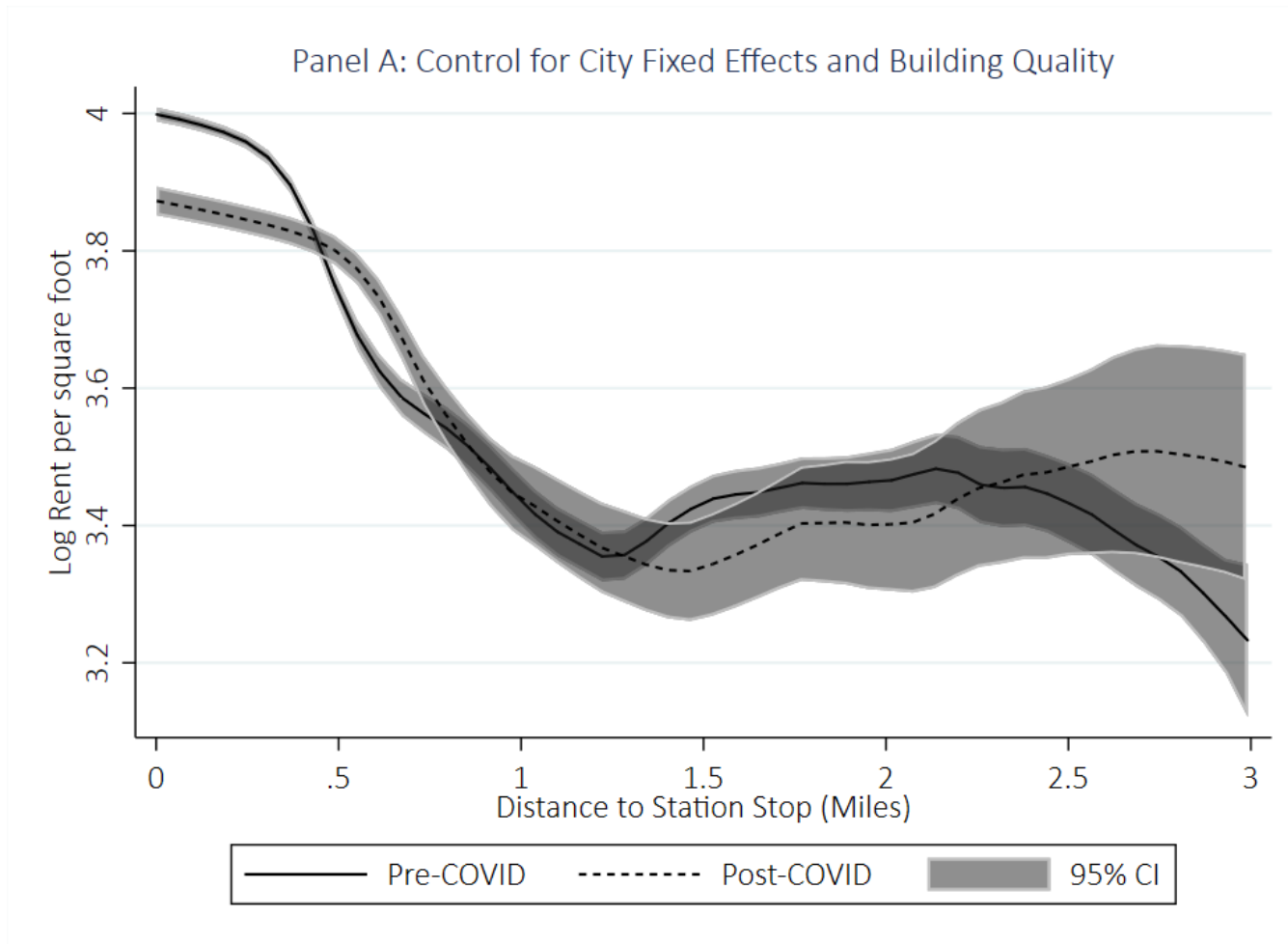
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## Partial linear regression

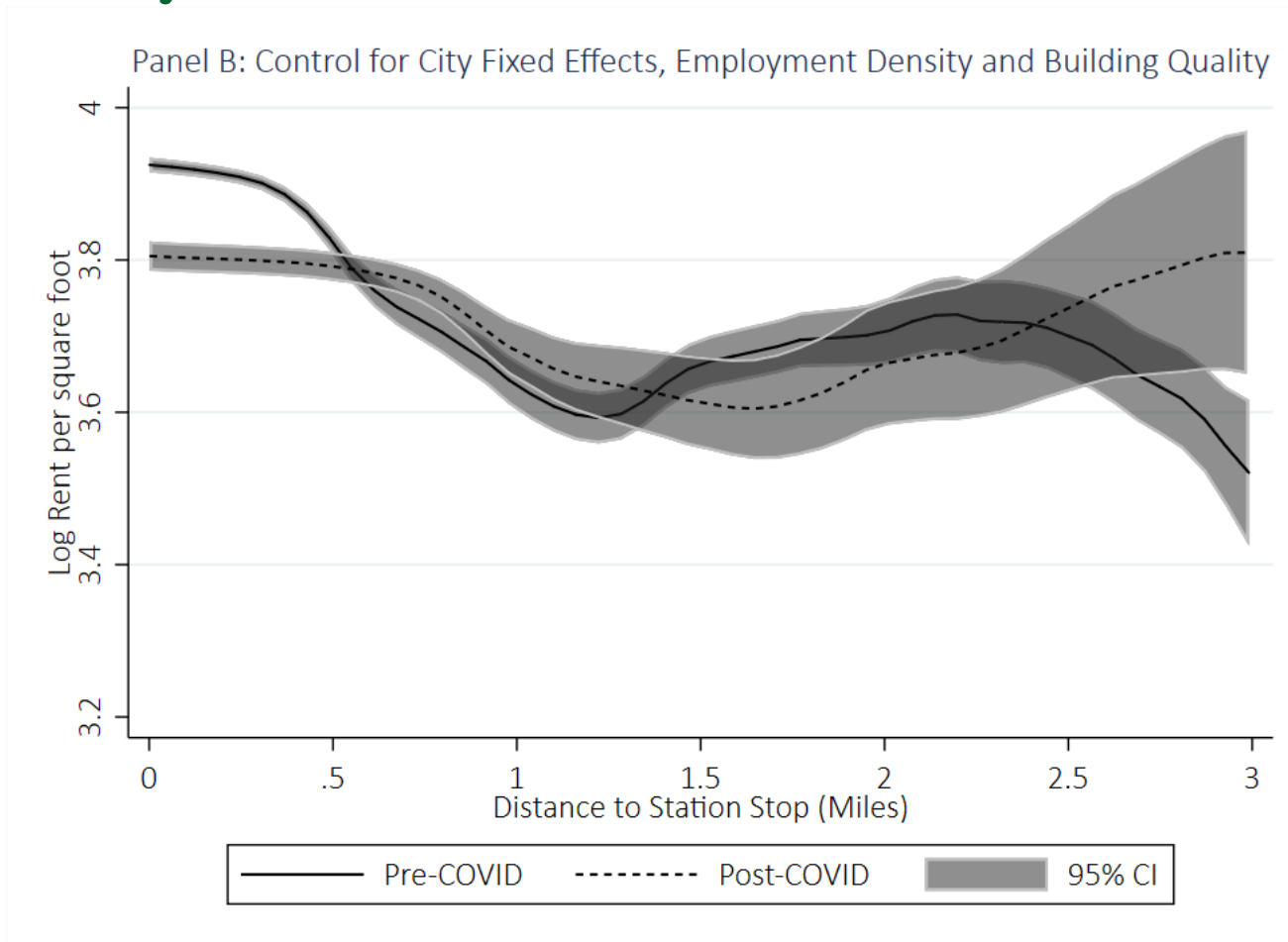
- Having estimated  $\beta$ ,  $m(z)$  can be estimated by regressing  $(y - X\beta)$  on  $z$  nonparametrically.



# Rent declines with distance to a transit station

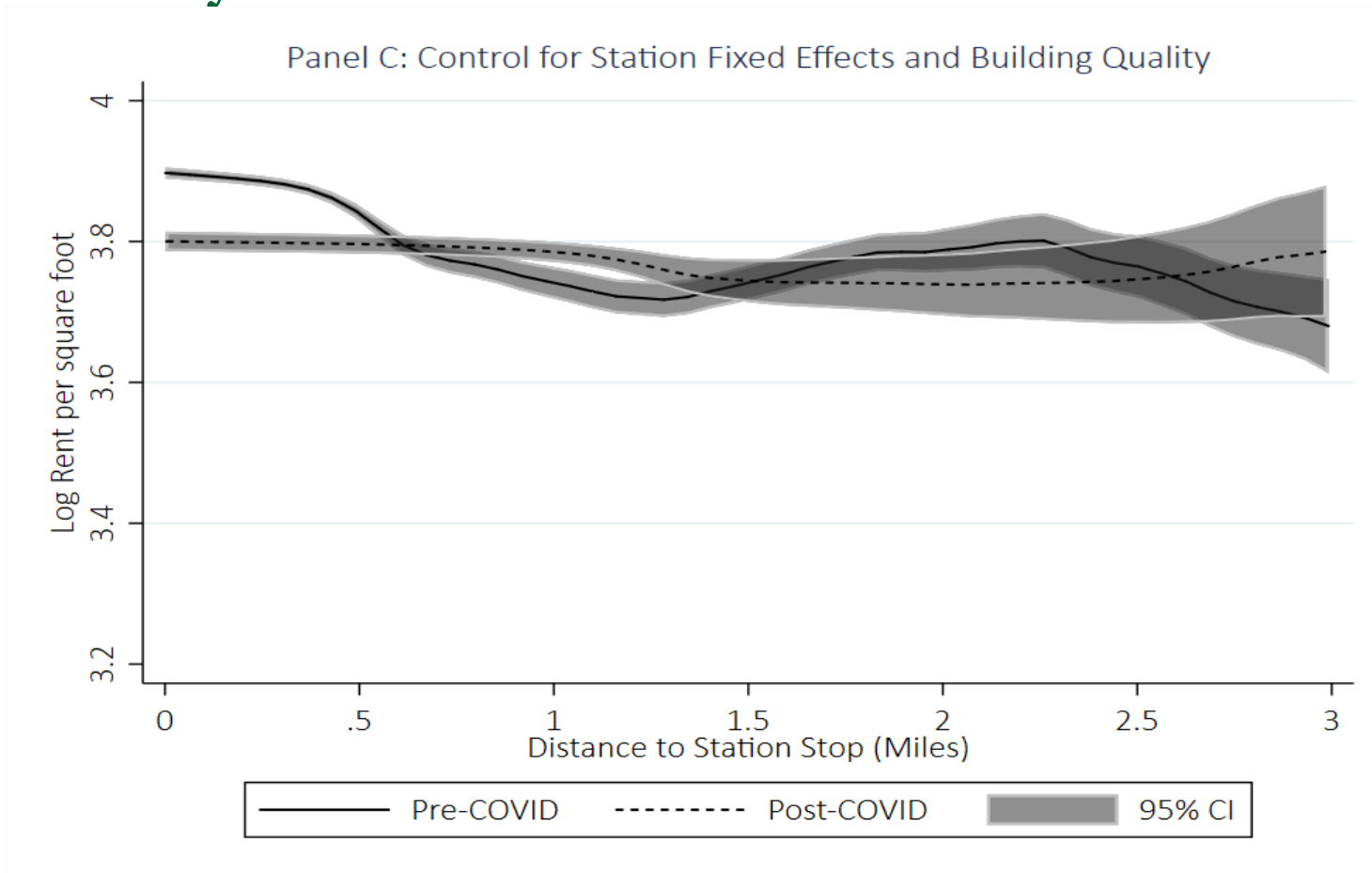


# The rent-transit station relationship persists with local density controls



Controlling for local density mutes but does not eliminate the pattern

# The rent-transit station relationship persists with local density controls



Controlling for local transit stop fixed effects further mutes the pattern

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# Conclusions

- Commercial rent data are increasingly available and provide a new way to evaluate the degree to which companies value different locations, including density.
- Pre-COVID, we obtain anticipated patterns
  - Rent declines with distance from the CBD
  - Rent is higher in high employment density locations
  - Rent declines with distance to a transit station that provides fast access to business centers

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## Conclusions

- COVID-19 has not affected all cities similarly.
  - Among transit cities, central locations and density are less valued, as is proximity to rapid transit.
  - Among car cities, we see only the local density effect.
- Will the effects on transit cities persist? While it is not possible to be certain, it is true that
  - Working from home has become established.
  - COVID-19 is unlikely to disappear.
  - The circumstances for the creation of novel viruses are present, and they do not seem to be weakening.