# You Can't Sit with Us: How Locals and Tourists Compete for Urban Amenities

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# Paris Town Hall (2019): "Are there too many tourists in Paris?"



#### There were reasons for concern. In 2019:

- France was the most visited country in the world
- Paris was the third most visited city in the world
- ► The number of foreign tourists to France had more than doubled over the previous 15 year
- During the year, 35.4 million tourists stayed in the city's hotels, which is approximately 16 times more than the population of the city.

# Anti-Tourism Protests across Europe



#### And then...

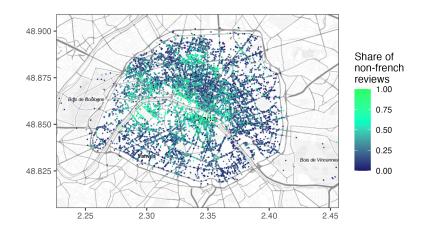


#### Research Question

How does tourism affect locals' satisfaction with amenities? We focus on restaurants as a classic example of amenities to answer this question.

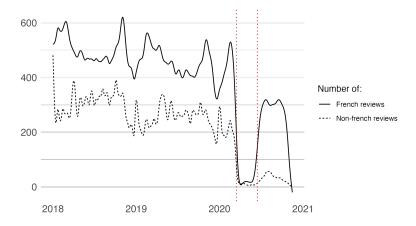
- We test three mechanisms:
  - Overcrowding
  - Supply-side change
  - Social frictions, such as xenophobia towards tourists
- ⇒ We draw on two episodes of exogenous drop in tourism:
  - ▶ November 2015 Paris terrorist attacks
  - ► First wave of COVID-19 pandemic

## Map of Restaurants by Share of Non-French Reviews

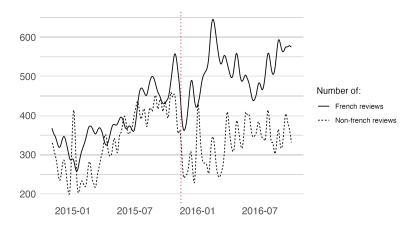


Grid map Grid map: restaurants density

# Daily Number of Restaurant Reviews in Paris (Pandemic Shock)



# Daily Number of Restaurant Reviews in Paris (November 2015 Attacks Shock)

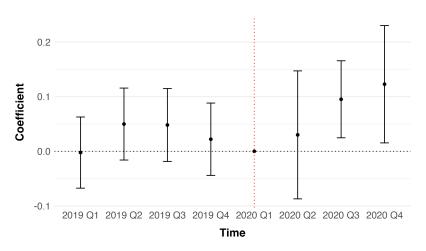


## Empirical Strategy: Difference in Difference

$$Y_{jt} = \beta \times Post-Shock_t \times Tourism_j + \gamma_j + \delta_t + \theta_{tn} + \epsilon_{jt}$$
 (1)

- $ightharpoonup Y_{jt}$  is an outcome of restaurant j in month t
- Post-Shock $_t$  a binary variable indicating whether month t belongs to the period after a shock (attack or pandemic)
- Tourism<sub>j</sub> to what extent restaurant j is frequented by tourists
- γ<sub>j</sub> restaurant fixed effects
- $ightharpoonup \delta_t$  month fixed effects
- $\bullet$   $\theta_{tn}$  month  $\times$  neighborhood fixed effects
- ▶ We cluster standard errors at the neighborhood level

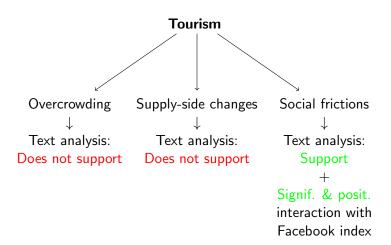
# Event Study Plot: Touristic Restaurants Have Relative Improvement in Ratings After Pandemic, Restaurant-Level Specification



# Tourism and Restaurant Ratings (Review-Level Analysis)

	Rating by Parisian			
	(1)	(2)	(3)	(4)
Variables				
Tourism Share $\times$ Post-Lockdown	0.0691***	0.0470*	0.0656**	0.0847**
	(0.0209)	(0.0240)	(0.0298)	(0.0389)
Fixed-effects				
Restaurant	Yes	Yes	Yes	Yes
Month	Yes	Yes		
User		Yes	Yes	
Month  imes Neighborhood			Yes	Yes
User × Post-Lockdown				Yes
Fit statistics				
Observations	120,314	120,314	120,314	120,314
$R^2$	0.28145	0.73488	0.74564	0.76153
Dependent variable mean	0.71999	0.71999	0.71999	0.71999

#### Potential Mechanisms



## **Textual Outcomes**

	Tourists (1)	Low Food Quality (2)	Too Expensive (3)	Too Noisy (4)	Long Wait (5)
Panel A: restaurant-level					
Variables					
Tourism Share $\times$ Post-Lockdown	-0.0646*** (0.0112)	-0.0032 (0.0190)	0.0044 (0.0142)	0.0093 (0.0109)	-0.0132 (0.0123)
Fixed-effects					
Restaurant	Yes	Yes	Yes	Yes	Yes
Month × Quarters	Yes	Yes	Yes	Yes	Yes
Fit statistics					
Observations	75,997	75,997	75,997	75,997	75,997
$R^2$	0.24881	0.23065	0.19966	0.18782	0.19802
Dependent variable mean	0.02306	0.07168	0.04727	0.02365	0.02561
Panel B: review-level					
Variables					
Tourism Share × Post-Lockdown	-0.0891***	-0.0032	-0.0334	0.0145	-0.0332
	(0.0222)	(0.0311)	(0.0278)	(0.0265)	(0.0223)
Fixed-effects					
User-Post-Lockdown	Yes	Yes	Yes	Yes	Yes
Restaurant	Yes	Yes	Yes	Yes	Yes
Month × Neighborhood	Yes	Yes	Yes	Yes	Yes
Fit statistics					
Observations	111,756	111,756	111,756	111,756	111,756
$R^2$	0.56827	0.60988	0.53738	0.47727	0.53808
Dependent variable mean	0.02274	0.07506	0.05095	0.02816	0.02702

# Textual Outcomes and Social Proximity

	Tourists (1)	Low Food Quality (2)	Too Expensive (3)	Too Noisy (4)	Long Wait (5)
Variables					
Tourism Share	-0.0491***	0.0197	0.0295	0.0043	-0.0162
$\times$ Post-Lockdown $\times$ High SCI	(0.0177)	(0.0334)	(0.0241)	(0.0130)	(0.0153)
Tourism Share	-0.0816***	-0.0221	0.0077	0.0171	-0.0135
imes Post-Lockdown $ imes$ Low SCI	(0.0160)	(0.0247)	(0.0183)	(0.0120)	(0.0135)
Fixed-effects					
Restaurant	Yes	Yes	Yes	Yes	Yes
$Month  \times  Quarter$	Yes	Yes	Yes	Yes	Yes
Fit statistics					
Observations	62,079	62,079	62,079	62,079	62,079
$R^2$	0.24497	0.22017	0.18684	0.18442	0.18753
Dependent variable mean	0.02580	0.07424	0.04878	0.02452	0.02618

#### Conclusion

- We document that during the pandemic a drop in tourism caused an increase in Parisians' satisfaction with restaurants and other amenities
- We document a similar effect for another shock in tourism caused by 2015 Paris attacks
- We consider three mechanisms overcrowding, supply-side changes and social frictions – and find support for the social frictions

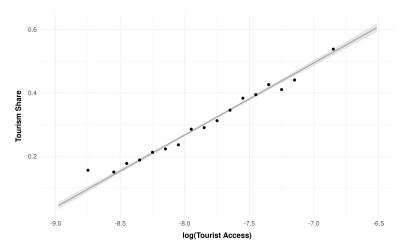
#### Data

- ➤ **Tripadvisor:** We collect data on restaurants reviews. We construct unique and highly detailed panel that reflects city's restaurant consumption across space and time. The final sample consists of around 15,000 restaurants and 2 million reviews.
- ▶ 'Dans ma rue' Mairie de Paris: application that allows users to write and geolocate complaints in Paris, e.g.
  - ► Abandoned bulky objects, Waste & dirt, Damaged road, Inconvenient parking, Graffiti, Overflowing litter bin, Rats

#### Facebook Social Connectedness Index

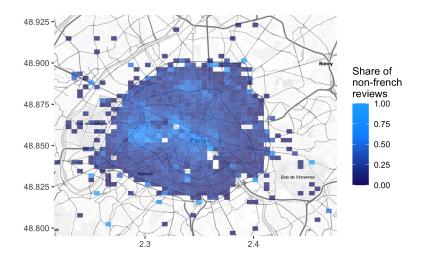
 Measures the density of network connections between users from different countries

# Tourist Access vs Tourism Proxy



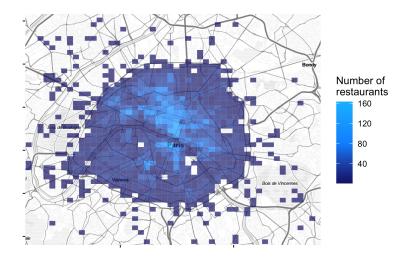
Tourist Access<sub>i</sub> = 
$$\sum_{j} \frac{\text{Visitors}_{j}}{\text{Distance}_{ij}}$$

## Grid Map of Restaurants by Share of Non-French Reviews



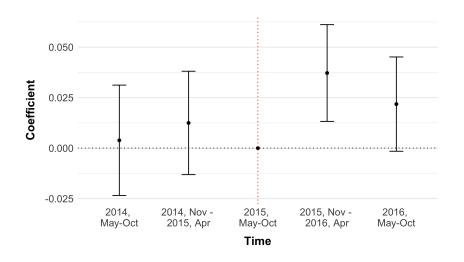


## Grid Map of Restaurants Density





# Event Study Plot: Touristic Restaurants Have Relative Improvement in Ratings After November 2015 Attack, Restaurant-Level Specification



# Tourism Decreases Resident's Satisfaction with Urban Amenities (Pandemic Shock)

Natural experiments:	Before and After First Pandemic Locko (Post = Post-Lockdo		
Dependent variables:	Avg. Rating (1)	by Parisians (2)	
Share of Non-French Reviews prior to observation period (by Restaurant) $\times$ Post	0.0752*** (0.0197)	0.0811*** (0.0238)	
Fixed-effects Restaurant	Yes	Yes	
Month	Yes		
$Month\timesNeighborhood$		Yes	
Fit statistics			
Observations	75,876	75,876	
$R^2$	0.35637	0.38035	
Dependent variable mean	0.71498	0.71498	
Dependent variable SD	0.3094	0.3094	

# Tourism Decreases Resident's Satisfaction with Urban Amenities (November 2015 Attacks Shock)

Natural experiments:	Before and After Terrorist Attack - November 2015 (Post = Post-Terrorist Attack)				
Dependent variables:	Avg. Ratin (3)	g by Parisians (4)	Avg. Rating (5)	by Non-Parisians (6)	
Share of Non-French Reviews prior to observation period (by Restaurant) × Post	0.0384*** (0.0094)	0.0335*** (0.0107)	0.0078 (0.0090)	0.0069 (0.0101)	
Fixed-effects					
Restaurant Month	Yes Yes	Yes	Yes Yes	Yes	
$Month  \times  Neighborhood$		Yes		Yes	
Fit statistics					
Observations	41,611	41,611	60,309	60,309	
$R^2$	0.36487	0.38716	0.33306	0.34983	
Dependent variable mean	0.68987	0.68987	0.73798	0.73798	
Dependent variable SD	0.2808	0.2808	0.2255	0.2255	

#### Other Results

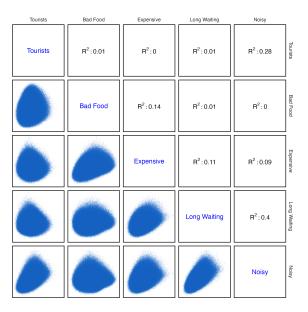
- Posit. and signif. for the streets anomalies Dans ma rue
- Robust to:
  - Different measures of tourism
  - Different aggregation periods

## "Dans Ma Rue" Complaints

	# Complaints				
	(1)	(2)	(3)	(4)	
Variables					
Share Tourism	-0.6570***	-0.2581*			
imes Post-Lockdown	(0.2272)	(0.1364)			
Top 25% Most Touristic			-0.3527***	-0.1504**	
× Post-Lockdown			(0.1213)	(0.0726)	
Fixed-effects					
Restaurant	Yes	Yes	Yes	Yes	
Month	Yes		Yes		
$Month  \times  Quarter$		Yes		Yes	
Fit statistics					
Observations	366,930	305,332	366,930	305,332	
$R^2$	0.48157	0.68477	0.48024	0.68481	
Dependent variable mean	0.40114	0.48207	0.40114	0.48207	



### Word Embedding Cosine Distances: Correlation Matrix



#### Social Connectedness Index

- We want to test whether the origin of tourists has an impact on locals' perception of them
- ➤ To proxy for cultural and social proximity between foreign countries and France we rely on the Social Connectedness Index (SCI) published by Facebook
- It is based on the number of Facebook friendships between users located in a pair of countries. More precisely, it is computed as:

Social Connectedness<sub>ij</sub> = 
$$\frac{\text{FB Friends}_{ij}}{\text{FB Users}_i \times \text{FB Users}_i}$$
,

# Social Proximity

	Avg. Rating by Parisian			
	(1)	(2)	(3)	(4)
Variables				
Tourism Share × Post-Lockdown	0.3073** (0.1206)			
Tourism Share $\times$ Post-Lockdown $\times$ High SCI	, ,	0.1623 (0.1506)		
Tourism Share $\times$ Post-Lockdown $\times$ Low SCI		0.3379*** (0.1209)		
Top 25% Most Touristic × Post-Lockdown		, ,	0.0865 (0.0571)	
Top 25% Most Touristic $ imes$ Post-Lockdown $ imes$ High SCI			(* * * * * * )	0.0384
Top 25% Most Touristic $\times$ Post-Lockdown $\times$ Low SCI				0.1209* (0.0637
Fixed-effects				
Restaurant	Yes	Yes	Yes	Yes
Month × Neighborhood	Yes	Yes	Yes	Yes
Fit statistics				
Observations	62,050	62,050	62,050	62,050
$R^2$	0.36701	0.36705	0.36696	0.36698
Dependent variable mean	3.8055	3.8055	3.8055	3.8055

# Spillovers

Dependent Variable:	Avg. Rating by Parisian					
Model:	(1)	(2)	(3)	(4)		
Variables						
Tourism Share $\times$ Post-Lockdown	0.3053*** (0.0836)	0.2790*** (0.1007)	0.3095*** (0.1020)	0.2775*** (0.1036)		
Touristic Area ( $<$ 100m) $\times$ Post-Lockdown	(5.5557)	-0.1396	()	0.0018		
Touristic Area (100m-300m) × Post-Lockdown		(0.1512) 0.4084*		(0.1551) 0.4558*		
Touristic Area (300m-500m) × Post-Lockdown		(0.2432) 0.0834		(0.2657) 0.1179		
,		(0.2977)		(0.3427)		
Touristic Area (500m-1000m) × Post-Lockdown		-0.3662 (0.2911)		0.0816 (0.4458)		
Fixed-effects						
Restaurant Month	Yes Yes	Yes Yes	Yes	Yes		
Month × Quarter	Yes	Yes	Yes	Yes		
Fit statistics						
Observations	63,410	63,410	63,410	63,410		
$R^2$	0.34439	0.34445	0.37327	0.37333		
Dependent variable mean	3.8157	3.8157	3.8157	3.8157		