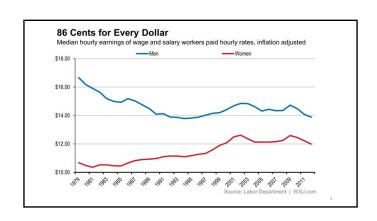
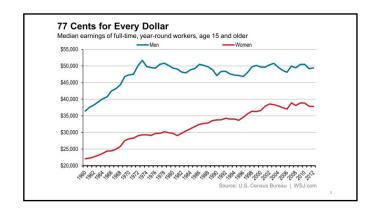
# The Gender Earnings Gap in the Gig Economy

# Gender earnings gap

- Women earn about 20-30% less then men
- 2016 American Community Survey
  - Annual 1% random sample of the population
  - Largest re-occurring survey in the US
  - Ask about work/earnings in the previous year
  - Full-time/full-year workers (30+ hours/week, 50+ weeks/year)
  - Ages 18-50

	Men	Women	Ratio: W/M	
Observations	345,612	262,695		
Mean annual earnings	\$63,844	\$49,971	0.78	
Median annual earnings	\$44,000	\$35,000	0.76	
Mean of hourly wage	\$26.88	\$21.97	0.82	
Median of hourly wage	\$19.61	\$17.16	0.88	





# Questions to consider

- Do wages = productivity?
- Why might this be the case?
- What is discrimination?
- How can we hold all else constant?

Top 10 Industries for Men and Women

% of	Ave annual	
workers	Avg. annual earnings	
3.3%	\$110,60	
2.6%	\$90,883	
1.6%	\$83,71	
3.5%	\$67,45	
1.9%	\$64,90	
1.5%	\$60,44	
2.8%	\$55,69	
10.0%	\$53,45	
1.7%	\$52,12	
3.9%	\$35,14	
	3.3% 2.6% 1.6% 3.5% 1.9% 1.5% 2.8% 10.0%	

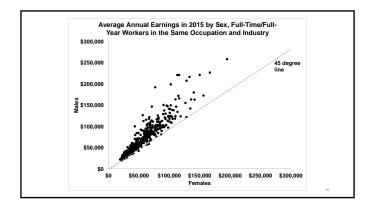
Women					
Industry	% of workers	Avg annual earnings			
Hospitals	9.5%	\$60,226			
Insurance carriers	3.0%	\$59,738			
Offices of Physicians	2.2%	\$59,174			
Banking and related services	2.3%	\$56,681			
Colleges/universities	3.0%	\$55,623			
Justice, order, safety activities	1.9%	\$53,101			
Outpatiernt care centers	2.2%	\$50,333			
Elem/Secondary schools	9.8%	\$47,854			
Nursing care facilities	2.2%	\$37,620			
Restaurants/food service	4.2%	\$28,438			

Top 10 Occupations for Men and Women

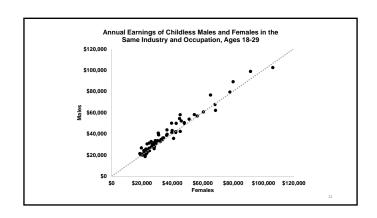
Mer	1		Women			
Occupation	% of workers	Avg. annual earnings	Occupation	% of workers	Avg annual earnings	
Software developer	1.7%	\$114,935	Manager	2.5%	\$81,227	
Manager	3.5%	\$104,540	Accountants/auditors	2.0%	\$67,775	
Sales rep	1.3%	\$90,124	Registered nurse	4.3%	\$63,757	
Supervisor retail	2.4%	\$58,395	Elem/Middle school teacher	4.3%	\$49,636	
Retail sales	1.9%	\$51,455	Supervisor retail	2.3%	\$41,979	
Driver	3.5%	\$47,187	Secretary/admin assistant	3.9%	\$37,913	
Construction laborer	1.9%	\$39,123	Customer service rep	2.5%	\$36,870	
Laborer (moving, stock, freight)	2.1%	\$35,318	Home health aid	2.3%	\$27,558	
lanitor	1.6%	\$34,760	Waiter/waitress	2.0%	\$24,444	
Cooks	1.6%	\$25,642	Cashier	2.7%	\$23,821	

# Hold things constant

- Work 50+ weeks
- Work 30+ hours/week
- Work in the same occupation and industry
  - Cashier/retail stores
  - Laborer/construction
  - Nurse/hospital



#### Ln(annual earnings) regression, 18-45 Independent variable Males (R2=0.565) Females (R2= 0.577) 0.051 (39.0) 0.060 (42.0) -0.00047 (-24.3) -0.00062 (-29.7) 1 child 0.034 (11.5) -0.020 (-7.41) 2 children 0.082 (28.2) -0.007 (-2.62) -0.039 (-10.1) -0.055 (-9.2) 3 children 0.073 (19.6) 0.059 (11.5) 4 children 0.084 (14.0) HS grad 0.141 (34.1) 0.211 (47.8) 0.135 (22.1) Some college BA/BS 0.395 (82.9) 0.347(55.1) Grad degree 0.555 (100.7) 0.531 (78.8) Married 0.102 (42.0) (t-tests in parentheses)



# Gig economy

- Characteristics
  - Divide work into small pieces
  - Independent workers
  - Real-time markets
  - Low barriers to entry
- - 15% do primarily independent work
  - 30% do some independent work
- Very flexible work arrangements

### Uber pay

- Chicago
  - \$1.70/trip
  - \$0.20/mile
- \$0.95/mile
- Some extras
  - Surge pricing
  - Bonus payments for reaching # trip threshold
  - Only 9% of earnings
- Key point: Earnings are productivity in this case

### Hourly pay

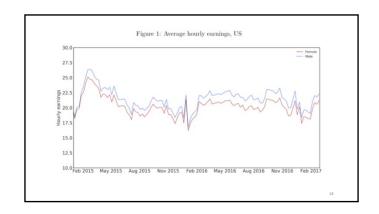
$$p(\cdot) = 60 * \left( \frac{SM \left( r_b + d_1 r_d + 60 * \frac{d_1 r_t}{s} \right) + I}{w + 60 * \frac{d_0 + d_1}{s}} \right)$$

- $r_b r_d r_t$  are base fare, per-mile and per-minute rates • SM = surge multiplier
- $d_0$  = distance between accepts and pickup
- d<sub>1</sub>= distance on trip
- s= speed
- w = wait time
- I =per trip incentive

Data

- Weekly earnings for all drivers, January 2015 March 2017
- UberX and UberPOOL
- Vibera ...
  Size
  741 million trips
  24.9 million driver weeks
  196 cities
  1.877 million drivers
  27.3% females
- For each week
   Total earnings
   Hours worked (total time APP is on)

	All	Men	Women
Weekly earnings	\$376.38	\$397.68	\$268.18
Hourly earnings	\$21.07	\$21.28	\$20.04
Hours per week	17.06	17.98	12.82
Trips per week	29.83	31.52	21.83
6 month attrition rate	68.1%	65.0%	76.5%
Number of drivers	1,873,474	1,361,289	512,185
Number driver/weeks	24,832,168	20,210,399	4,621,760
Number of Uber trips	740,627,707	646,965,269	93,662,438



### $\ln(Earnings_{dt}) = \beta_0 + \beta_1 isMale_d + \rho X_{dt} + \epsilon_d$

Table 2: National gender pay gap

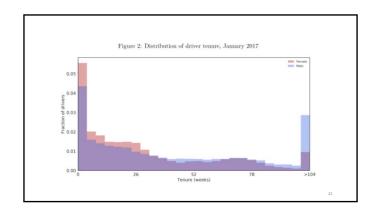
	(1)	(2)	(3)	(4)
	log(weekly earnings)	log(weekly earnings)	log(hourly earnings)	log(hourly earnings)
isMale	0.4142	0.4092	0.0702	0.0653
	(0.002)	(0.002)	(0.001)	(0.001)
Intercept	4.9737	4.9208	2.9280	2.8849
	(0.002)	(0.002)	(0.001)	(0.001)
City	X	X	X	X
Week		X		X
N	24,877,588	24,877,588	24,877,588	24,877,588
Drivers	1,877,252	1,877,252	1,877,252	1,877,252
R <sup>2</sup>	0.125	0.136	0.199	0.239

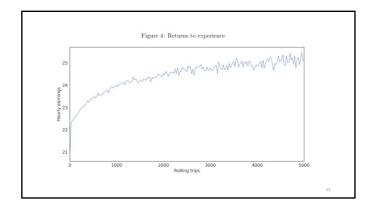
Note: This table documents the gender pay gap for all US cities from January 2015 to March 2017. Data are at the driverweek level; weekly carnings is the entire pay for a given week, while hourly earnings is the pay divided by hours worked in the week. Standard errors (clustered at the driver-level) in parentheses. Data for Chicago

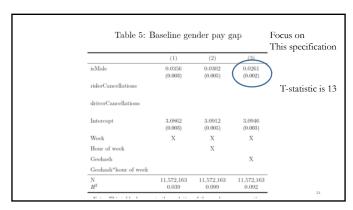
- Hold city and prices constant
- 33 million driver hours
- 120K drivers, 30% are female
- Place driver in a 3 mile x 3 mile geohash where they pick up a passenger
  - Holds location constant
  - Men may work in different geographies than females

20

	Men	Women	Difference
w – Wait time (min)	8.223 (0.008)	8.218 (0.019)	-0.005
$d_0$ – Accepts-to-pickup distance (mi)	0.485 (0.000)	0.500 (0.001)	0.015
$d_1$ – Trip distance (mi)	5,035 (0.003)	4.875 (0.006)	0.160
s – Speed (mph)	19.532 (0.006)	18.760 (0.012)	0.772
SM- Surge multiplier	1.051 (0.000)	1.046 (0.000)	0.005
I – Incentive payout (8)	0.903 (0.001)	0.818 (0.002)	0.085
Total per-trip payout (8)	10.142 (0.004)	9.841 (0.008)	0.301







	(1)	(2)	(3)	(4)	(5)	(6)	
isMale	0.0256 (0.004)	0.0106 (0.002)	0.0101 (0.002)	0.0016 (0.002)	-0.0018 (0.002)	-0.0019 (0.002)	T-statistic is
logSpeed	0.2677 (0.002)	0.4552 (0.001)	0.4623 (0.001)	0.2715 (0.002)	0.4544 (0.001)	0.4616 (0.001)	-0.9 with 11.6 million
Trips completed: 100-500				0.0563 (0.001)	0.0318 (0.001)	0.0321 (0.001)	obsevations
Trips completed: 500-1000				0.0819 (0.002)	0.0460 (0.001)	0.0460 (0.001)	obsevations
Trips completed: 1000-2500				0.1075 (0.003)	0.0599 (0.002)	0.0594 (0.002)	
Trips completed: >2500				0.1519 (0.004)	0.0831 (0.0003)	0.0810 (0.003)	
Intercept	2.3084 (0.003)	1.7704 (0.004)	1.7502 (0.006)	2.2293 (0.005)	1.7346 (0.004)	1.7083 (0.004)	
Week	X	X	X	X	X	X	
Hour of week		X	X		X	X	
Geohash		X	X		X	X	
Geohash*hour of week			X			X	
$\frac{N}{R^2}$	11,572,163 0.101	11,572,163 0.263	11,572,163 0.282	11,572,163 0.111	11,572,163 0.266	11,572,163 0.284	
Note: The table expands on ea	rlier regressions	ov adding sog s	peed as an exp	danatory varial of hourly earni	ic. Speed is too	ed on total trip	