

Automated **Digitization** of the Censuses of Housing **Block Statistics**, 1940-1970

Jeffrey Lin, Dan Moulton, Isaac Rand & Robyn Smith Federal Reserve Bank of Philadelphia

August 2024

Disclaimer

The views expressed here are those of the authors and do not necessarily represent the views of the Federal Reserve Bank of Philadelphia or the Federal Reserve System.



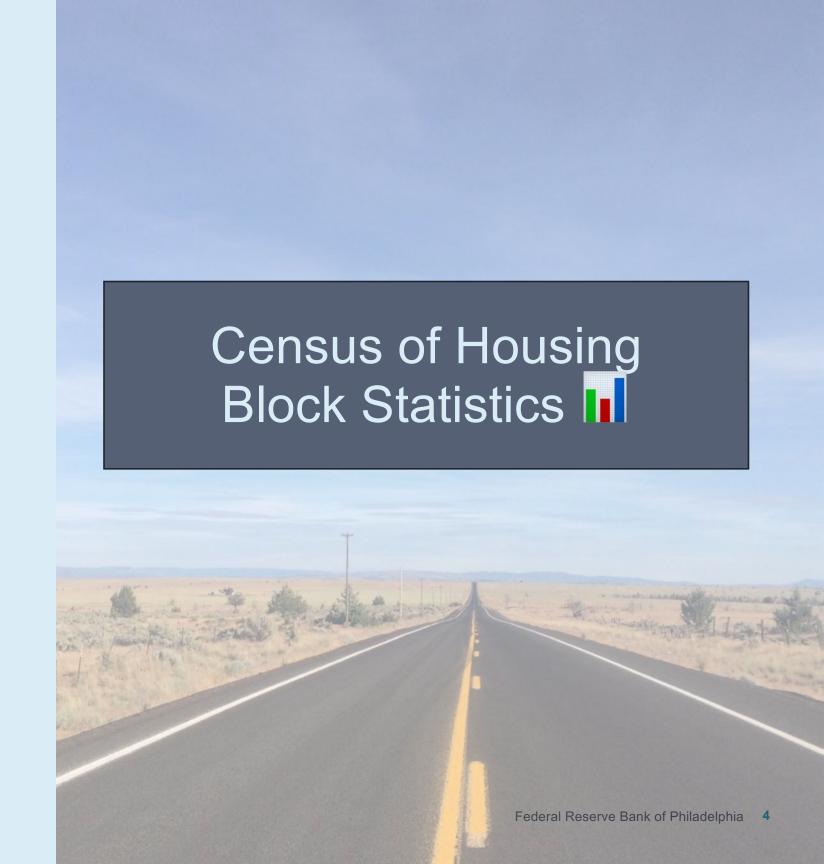
Digitizing Block Statistics

- What
- Why
- Goals
- Tasks and Challenges



Digitizing Block Statistics

- What
- Why
- Goals
- Tasks and Challenges



HOUSING—BLOCK STATISTICS

Table 3.—CHARACTERISTICS OF HOUSING FOR CENSUS TRACTS BY BLOCKS: 1940—Con.

Cen- sus tract	Block	Total struc- tures	ALL DWELLING UNITS BY OCCUPANCY AND TENURE					ALL DWELLING UNITS BY YEAR BUILT					OCCUPIED DWELLING UNITS				ALL DWELLING UNITS BY STATE OF REPAIR AND PLUMBING EQUIPMENT				OWNER-OCCU- PIED UNITS BY MORTGAGE STATUS		ALL DWELLING UNITS BY CONTRACT OR ESTIMATED RENT	
			Total dwell- ing units	Owner occu- pied	Ten- ant occu- pied	Va- cant, for sale or rent	Va- cant, other	Number report- ing	1930 to 1940	1920 to 1929	1900 to 1919	1899 or before	Total occu- pied	Occu- pied by non- white	Num- ber	sons room 1.51 or more	Number report- ing	Needing repair or no private bath	Need- ing re- pair	No pri- vate bath	Number report- ing	Mort- gaged	Number report- ing	Average monthly rent (Dollars)
3-A	24 25 26 27 28 30 31 33 33 33 33 40 41 42 43	21 41 33 4 26 11 24 12 28 22 23 30 4 43 27 28 28 28 28 28 28 28 28 28 28 28 28 28	3436894258334 901136441603	11 200 16 13 18 12 11 16 3 3 3 7 7 7 4 4 12 2 11 10 4 10 4 10 10 10 10 10 10 10 10 10 10 10 10 10	17 20 19 24 30 120 23 20 27 66 23 220 28 47 25 43 3	6311111332	1	342 3637 494 2538 334 90 311 426 444 716 493	2 1 1 2 2	3 2 3	442339 2 22 2 2	3 4 2 6 2 3 3 4 3 8 2 9 7 8 3 3 9 6 4 1 1 3 4 9 3	28 400 355 377 48 244 355 313 39 39 39 40 68 35 347 37	19 1 15 12 1 2 11 1	289 337 48244 3283 990 33440 6354 3	2 1 4 3 4 1 4 6	4 0 3 6 3 7 4 8 8 2 4 4 3 8 8 3 2 2 3 4 9 3 0 2 7 4 3 6 4 4 4	175244 155333 296198 14477145 277145	1 1 3 3 3 3 3 3 3 3 1 0 1 1 1 1 1 1 1 1	190 175 248 150 150 193 200 114 200 223	1 0 5 3 1 0 3 5 4 7 1 5 1 5	9 11 6 8 6 2 5 8 5 1 3 2 4 1 4 6	436 347 422 228 331 34 90 312 444 716	2424 1821 1822 23.73 21.98 22.55 17.86 20.66 20.42 18.71 20.00 21.57 17.32 22.14 18.04 28.32 24.70 17.80 18.67



Census of Housing Block Statistics III

- Most granular, earliest, extant Census spatial data on housing.
- 1940-1970.
- Tens of thousands+ of scanned pages of tables and maps.

What's in it?

- Tenure, occupancy, structure age and condition, rents and values, race of occupants.
- All houses, not just occupied ones.
- High level of spatial detail: Usually, a city block.
- Small size (Pop. ~50 vs ~4,000 for ED/Tract).
- Coverage of large section of cities over time.
- 191 cities in 1940 \rightarrow All 1970 urbanized areas.

What's it good for?

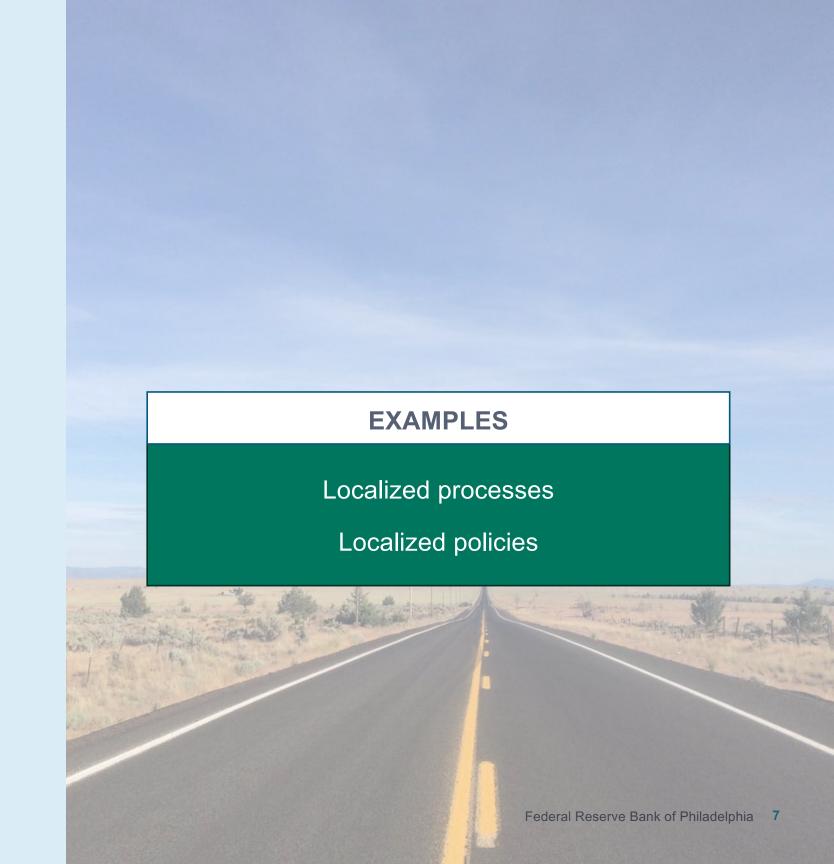
Studies of housing investment and maintenance and long-run urban dynamics.

Studies of policies and processes that occur at extremely localized spatial scales.

Studies of many cities, or a single city's history.

Digitizing Block Statistics

- What
- Why
- Goals
- Tasks and Challenges



Localized Processes

Residential segregation in South Philadelphia, 1970 at Census Tract scale



Percent of population Black

No Pop

Localized Processes

Residential segregation in South Philadelphia, 1970 at Census Block scale



Percent of

population Black

No Pop

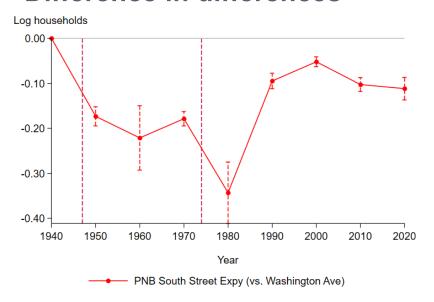
Localized Policies

Runner-up design

- "Expecting an Expressway" (Brinkman, Lin & Mangum).
- Two proposed routes for the Crosstown Expressway in South Philadelphia.



Difference in differences



Digitizing Block Statistics

- What
- Why
- Goals
- Tasks and Challenges



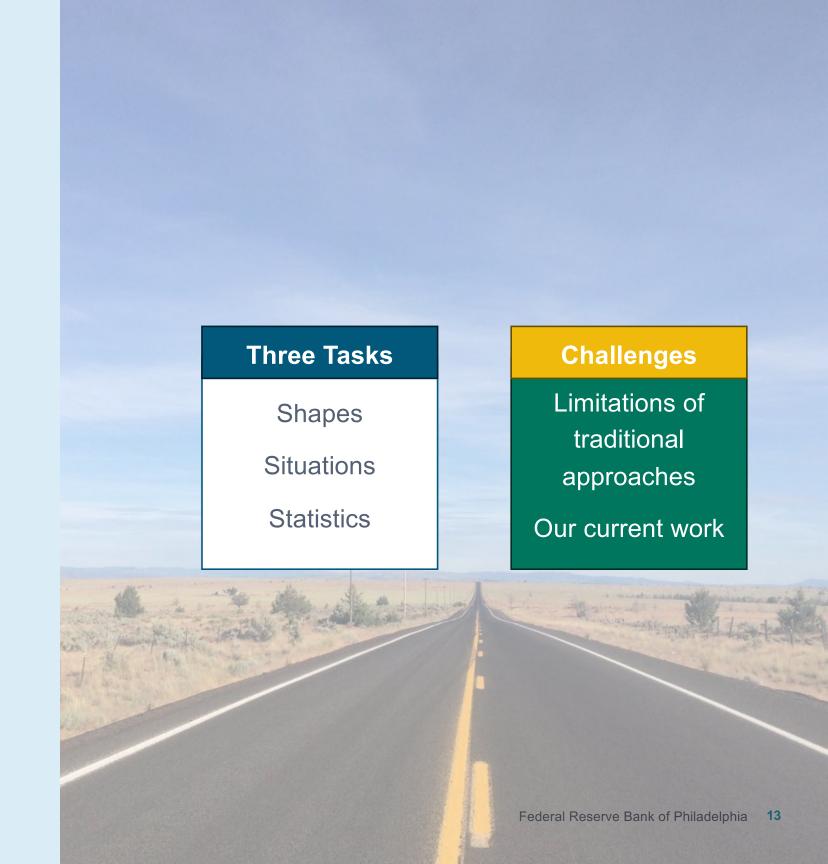
Boston Philadelphia Pittsburgh Columbus Washington Baltimore San Francisco Los Angeles Atlanta Houston

Our Goals

- Block data for **16 cities**, 1940-1970.
- Training and validation data.
- · Code and methods.
- Freely distributed for use and re-use.

Digitizing Block Statistics

- What
- Why
- Goals
- Tasks and Challenges



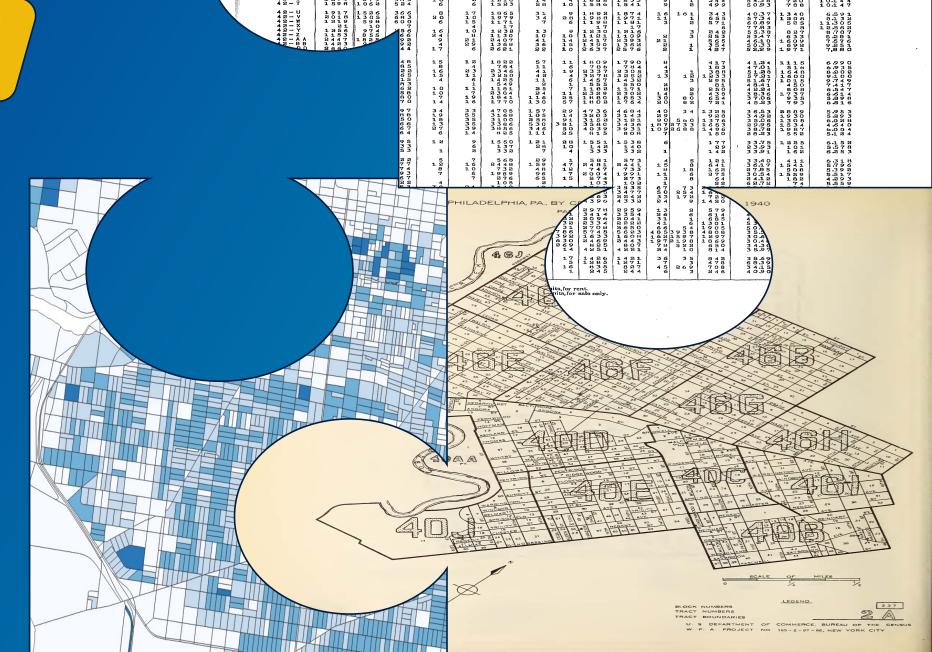


Shapes

Blocks need to know their:

2 Situations

3 Statistics



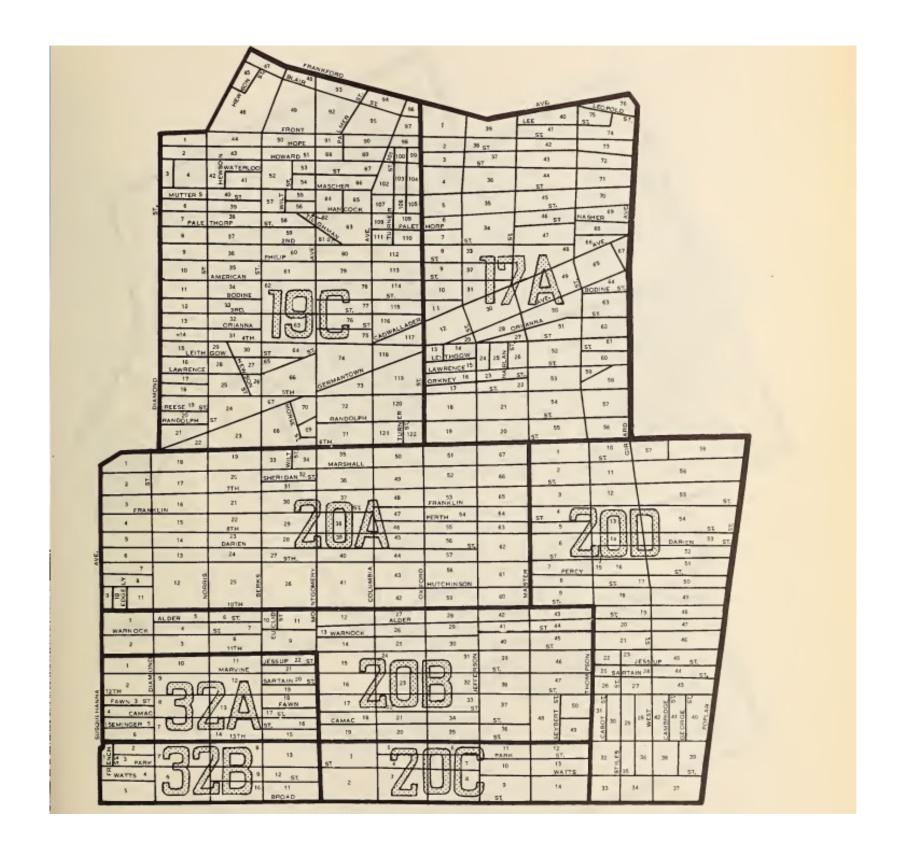
3 Tasks, 3 Pieces of Data

Shape 2 Situation 3 **Statistics**

Segmenting Block Shapes from Maps

Our Ideal Process Has Only Three Steps

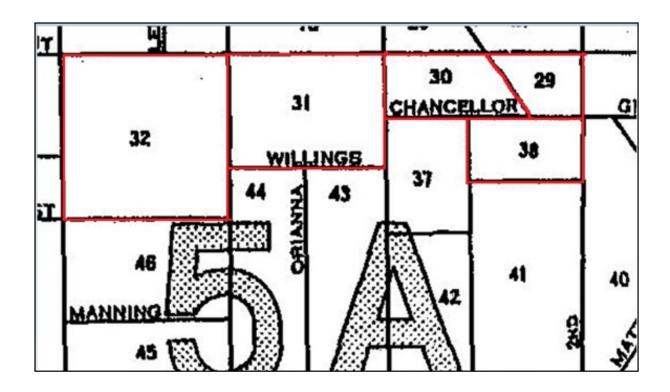
- 1. Identify closed loops of black ink.
- 2. Call them all blocks.
- 3. Declare victory.

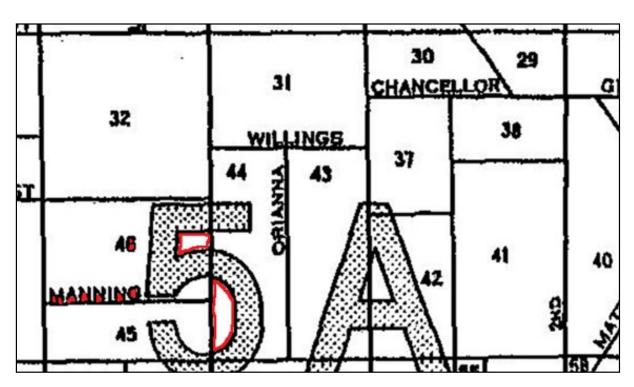


Unfortunately, This Process Fails Spectacularly

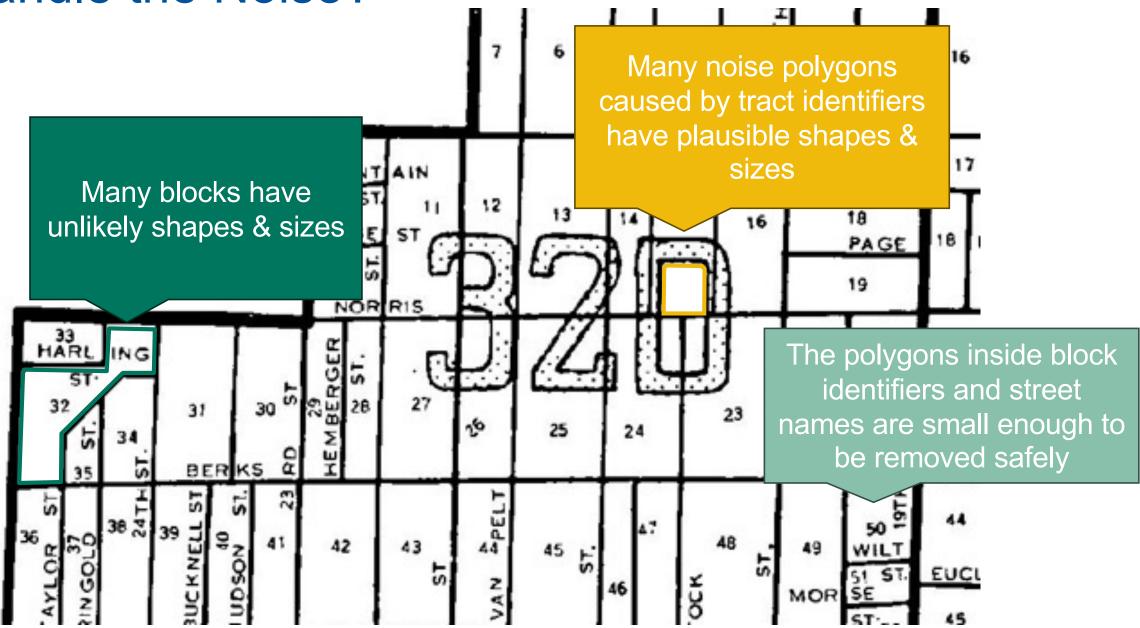
While many closed loops of black ink are blocks....

Many closed loops of black ink are not blocks 🕾



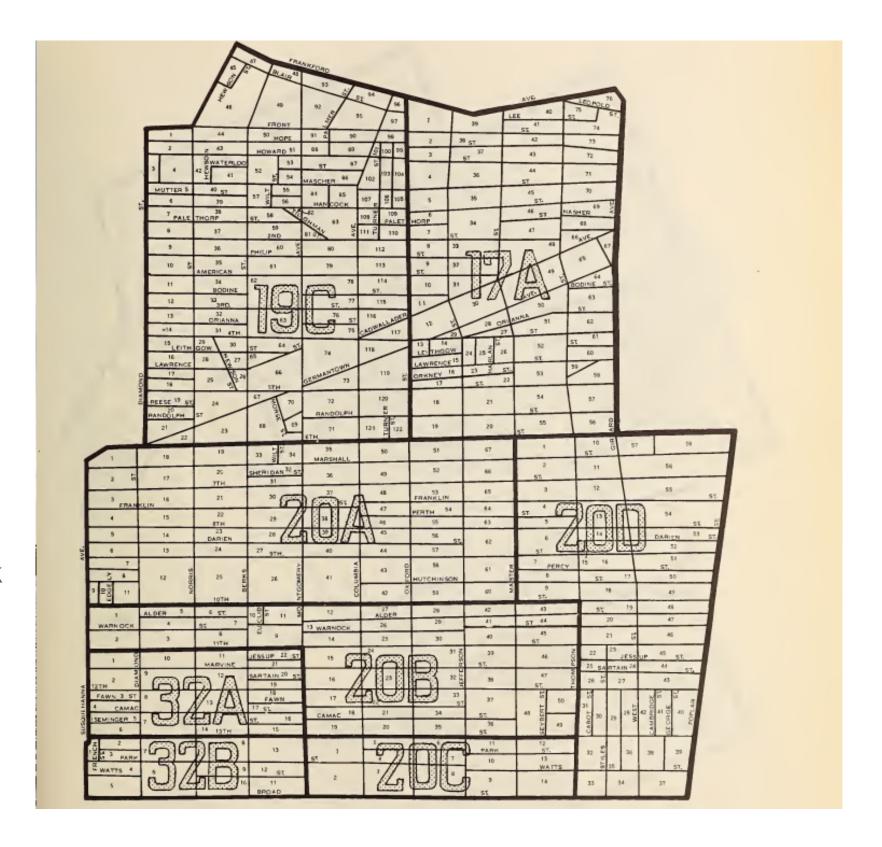


Can We Handle the Noise?



Our Ideal Process Has Only Three Four Steps

- 1. Remove the tract identifiers from the page
- 2. Identify remaining closed loops of black ink
- 3. Call any reasonably large loops blocks
- 4. Declare victory



How can we remove tract identifiers?

Traditional Method 1: Matching Large Shapes

Issues

Arbitrary Rotations, Inconsistent Scale, Shape, and Font, Noise/Interference from other features.

Low-confidence matches

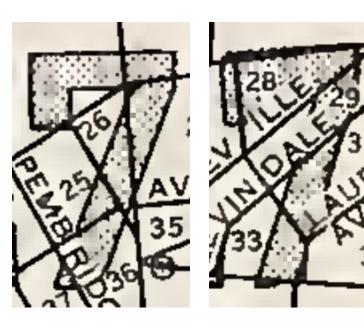
- If we accept low confidence matches of enough templates, everything starts to look like a tract identifier.
- Especially problematic with blocky characters like 1 and E.





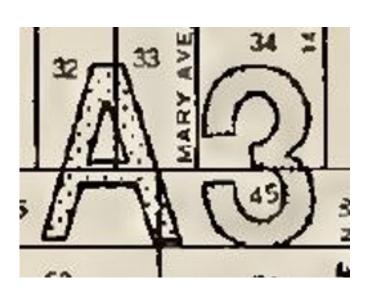


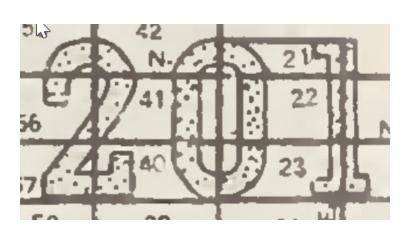




Traditional Method 2: Matching Patterns

- Sometimes speckled
- Sometimes no fill
- Block boundaries still a problem
- Sometimes inked













Current Work: CNN

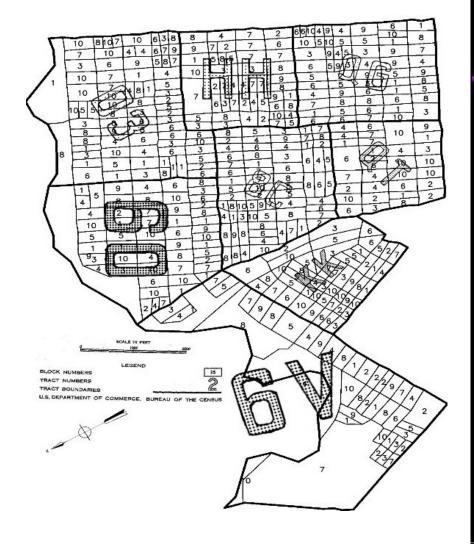
More holistic

- Consider properties of block boundaries as well as properties of tract identifiers.
- Focus on identifying block boundaries, not removing tract identifiers.

More flexible

- Can learn more patterns than we can with shape template matching.
- Can address partial shapes.
- Important because of intersections between boundaries and identifiers.







Simulated Map

Training Mask

Creating **Training Data**

- Hand annotations are expensive; Simulating maps is cheap.
- Sample 1990 Census block and tract boundaries from NHGIS.
- Sample tract and block identifiers from real 1940 maps.
- Randomly assign speckle density to tract identifiers.

Can the model trained on simulated maps generalize to real ones?





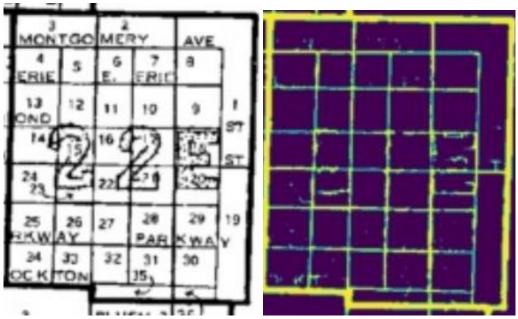
What's Next?

Model and training improvements

- Better simulated maps.
- Augment with hand annotations.

Add more steps

- Inpainting lines erased by CNN.
- Suggestions?





And Now For Something Completely Different

(1970 maps)

Promises

Tract boundary segmentation is somewhat easier.

Pitfalls

- Which block is this? Block identifiers are inconsistently located, look like street names.
- Too much detail: Block boundaries look like streets.
- "Fishhooks" are important and omnipresent.

Our current approach

We are relying on hand annotations for training and validating CNN.



3 Tasks, 3 Pieces of Data

1 Shape

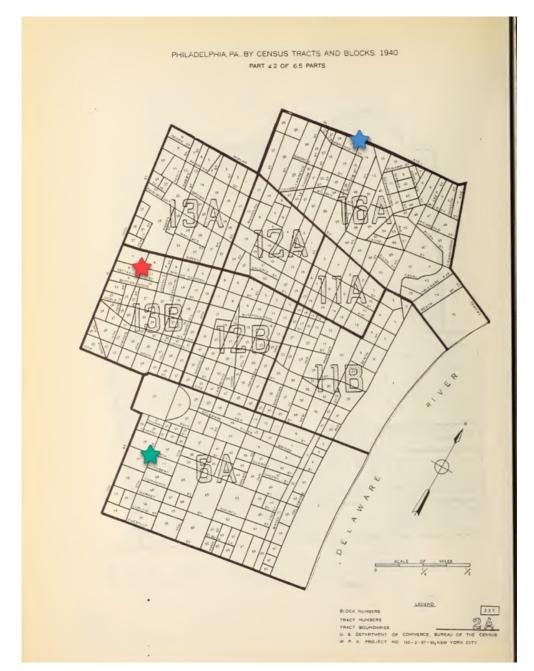
2 Situation

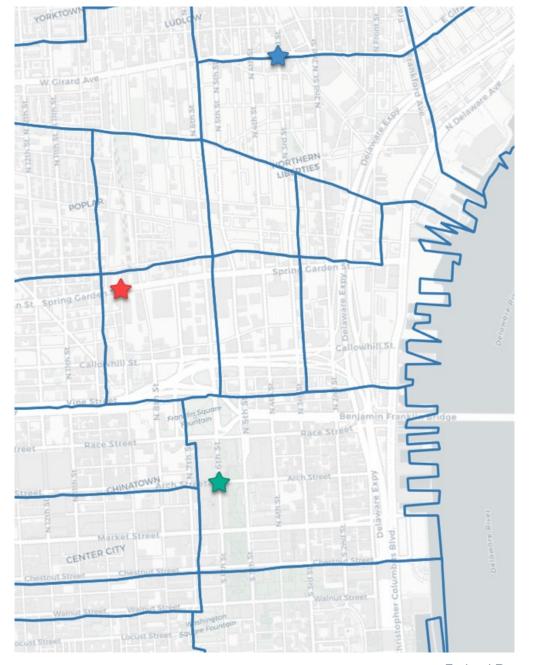
Geo-Referencing Maps

3 Statistics

Just Keep Clicking











A Georeferenced Map ©

We know where this picture goes now!

But...

- Slooooooow.
- Only 3 points doesn't handle map inaccuracies well.
- Doesn't match (blue) reference NHGIS shapefile.



How Can We Get Better?

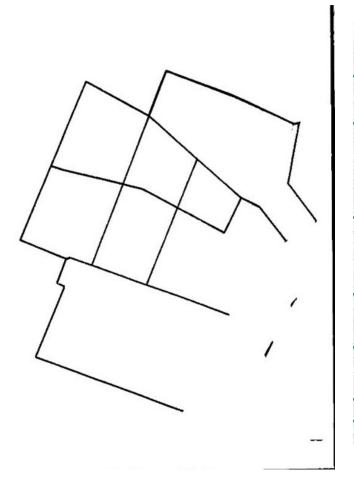
- Faster!
 - We want to process many maps.
- More accurate!
- How much of this can a computer do for us?

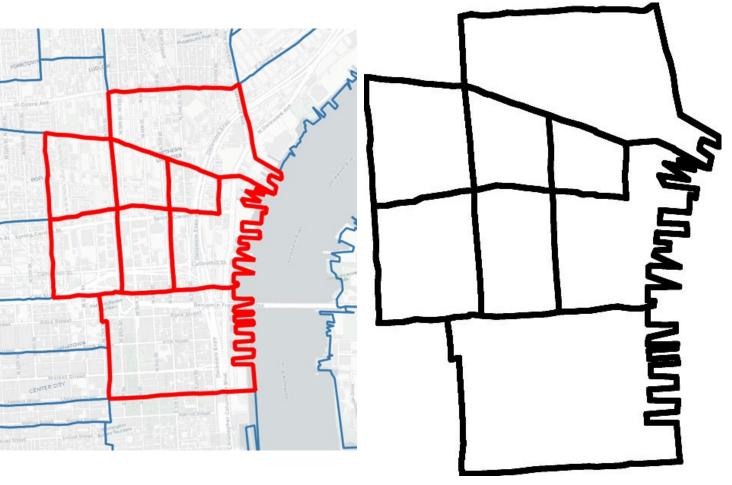
Simplifying the Problem





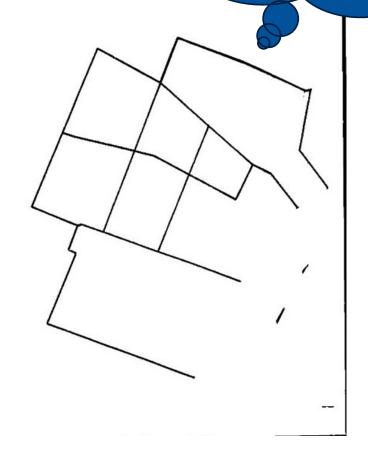


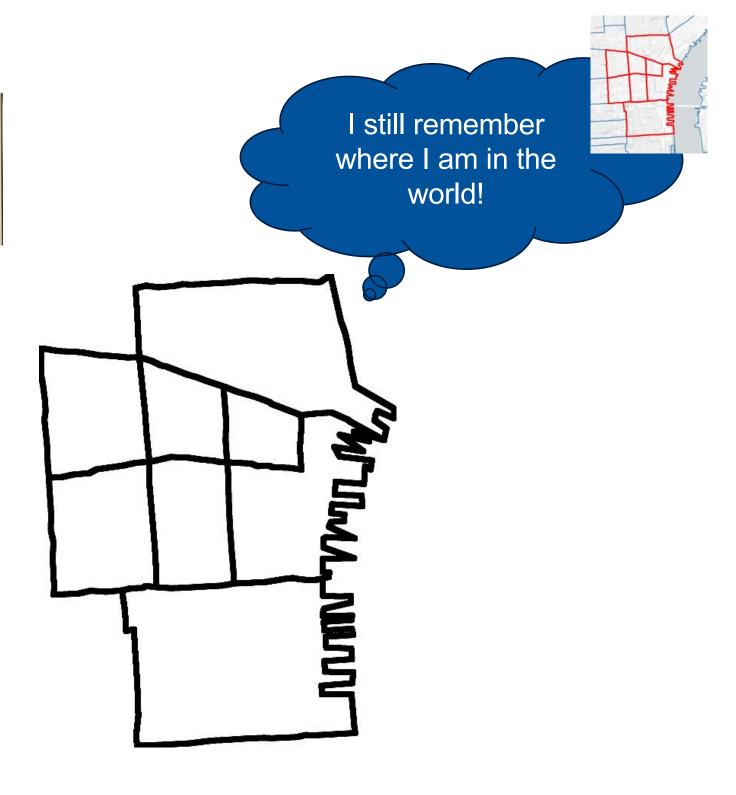




How Hard Could This Be?

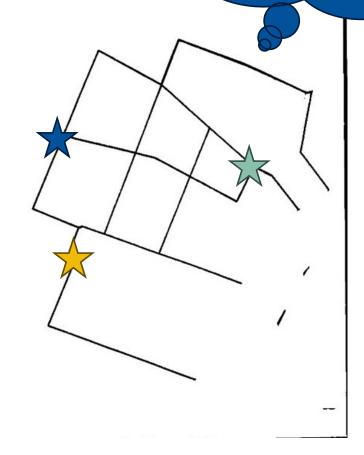
I still remember the block boundaries!



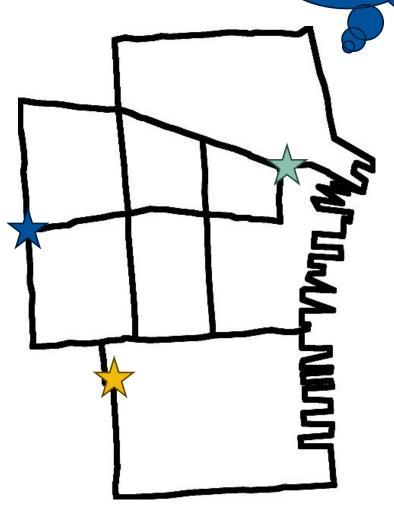


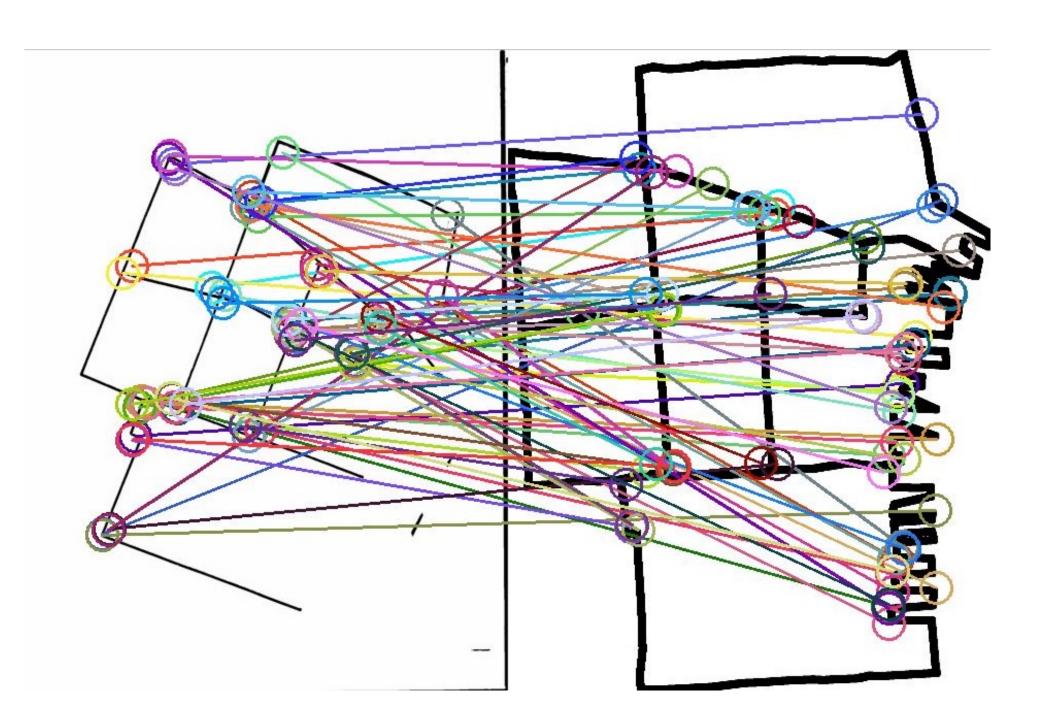
How Hard Could This Be?

I still remember the block boundaries!









Super. Duper. Hard.

- Too many matches.
- Several per corner!
- We need to narrow these down.

The Basic Steps 🦥 🔭

1 Make guess

Randomly select internally consistent links.

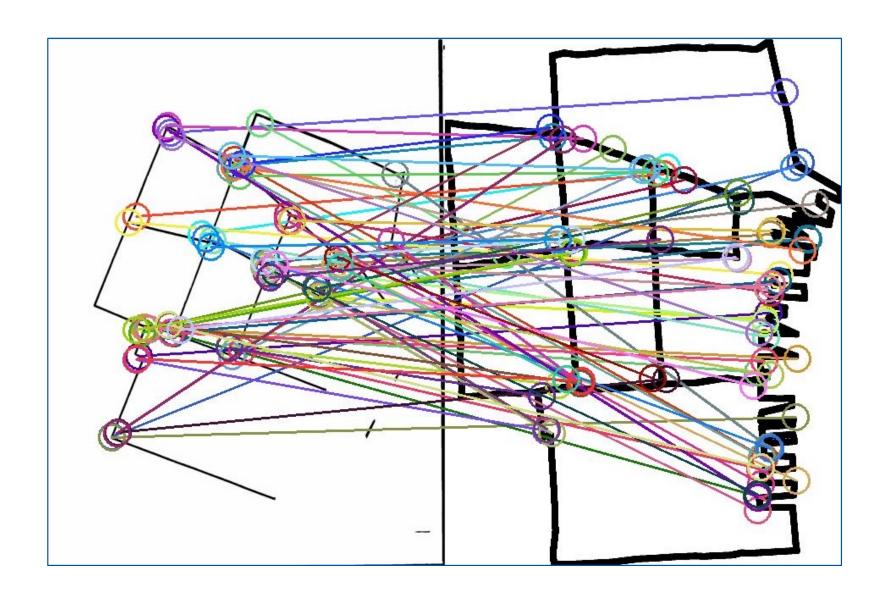
2 Evaluate guess

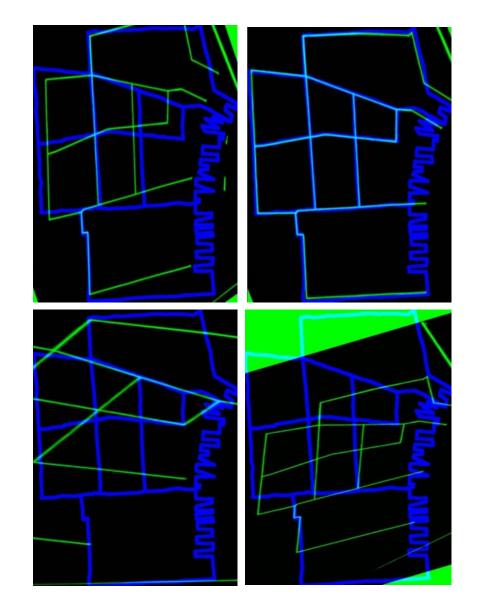
For selected links, how much do maps overlap?

3 Repeat

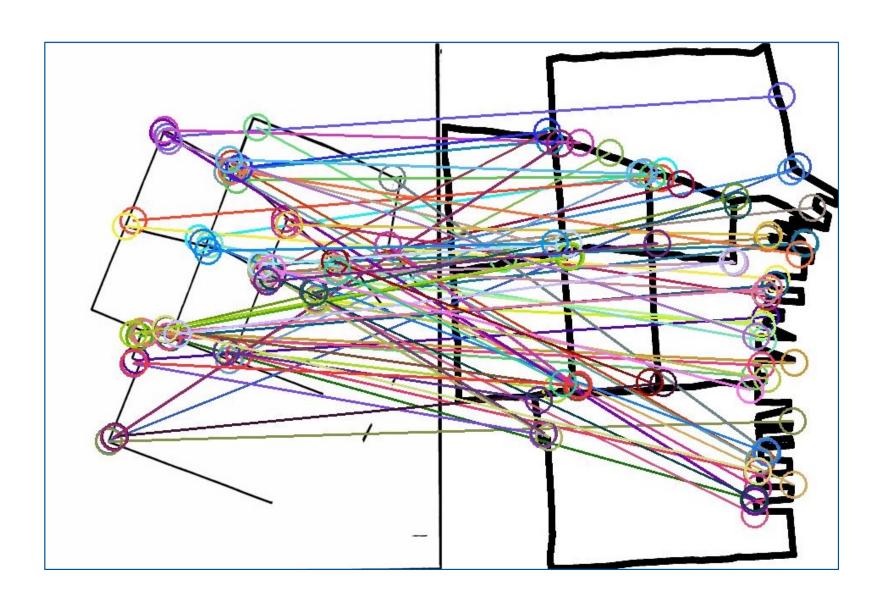
Keep best guess.

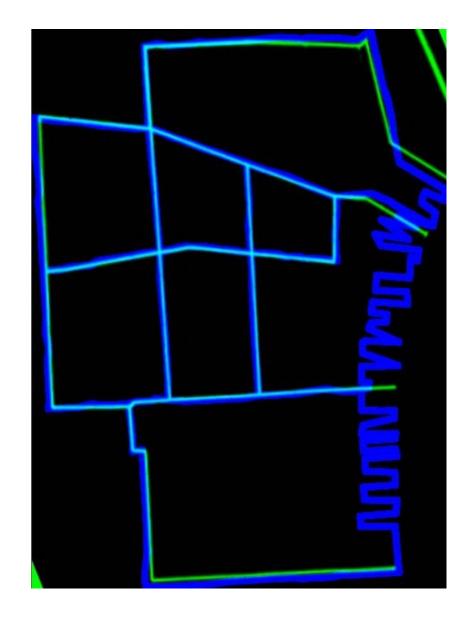
Selecting Links 2 2 2



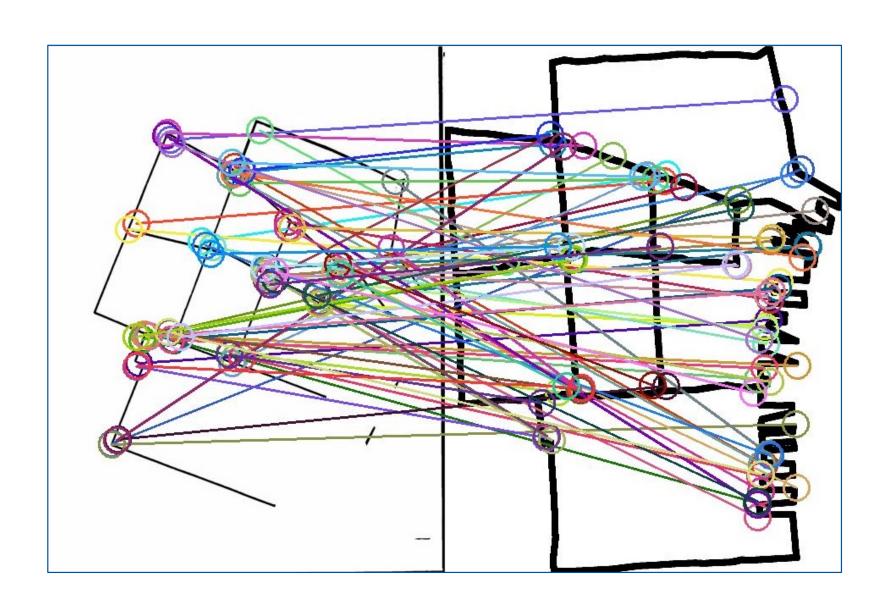


Selecting Links 2





Selecting Links 2 2 2





3 Tasks, 3 Pieces of Data

1 Shape
2 Situation

3 Statistics

Digitizing Tables

Scale and Scope of Problem

- 1940, 1950, and 1960 Census of Housing, Block Statistics (1970 is digital)
- Sixteen Target Cities
 - New York City
 - Chicago
 - Philadelphia
 - Los Angeles

- Detroit
- Baltimore
- Cleveland
- St. Louis

- Washington, DC
- Boston
- San Francisco
- Pittsburgh

- Houston
- Cincinnati
- Columbus, OH
- Atlanta

Scale and Scope of Problem

- 1940, 1950, and 1960 Census of Housing, Block Statistics
- Sixteen Target Cities

New York City

Detroit

Washington, DC

Houston

Chicago

Baltimore

Boston

Cincinnati

Philadelphia

Cleveland

San Francisco

Columbus, OH

Los Angeles

• St. Louis

Pittsburgh

Atlanta

- ~2,000 pages of tabular data, ~170,000 blocks, ~2.5 million cells per decade
- Structured, tabular form, with rows and columns properly associated and with accuracy better than 99%

Bottom Line Up Front

- Four stage process
 - Isolate table and each column
 - First pass with Tesseract
 - Algorithm to structure table
 - ML model to correct errors in OCR
- Great results
- Approach only makes sense if dataset is large and consistent

Bottom Line Up Front (1950)

- Custom Solution
 - 0.07% Observations with Error
 - 0.03% Character Error Rate

Bottom Line Up Front (1950)

Custom Solution

- 0.07% Observations with Error
- 0.03% Character Error Rate

Data Entry

- 0.12% Observations with Error
- 0.13% Character Error Rate

Tesseract (with assist with table structure)

- 12.94% Observations with Error
- 7.24% Character Error Rate

Adobe:

		ne-dwelling- tructures
Census tract	Block	Average value (dollars)
10-B	11	4.425
	15 16 17 18 19 20 21 22	7.28 8 8.1 5 0
	23 24 25 26 29 30 31 32 34	6.366 3.900

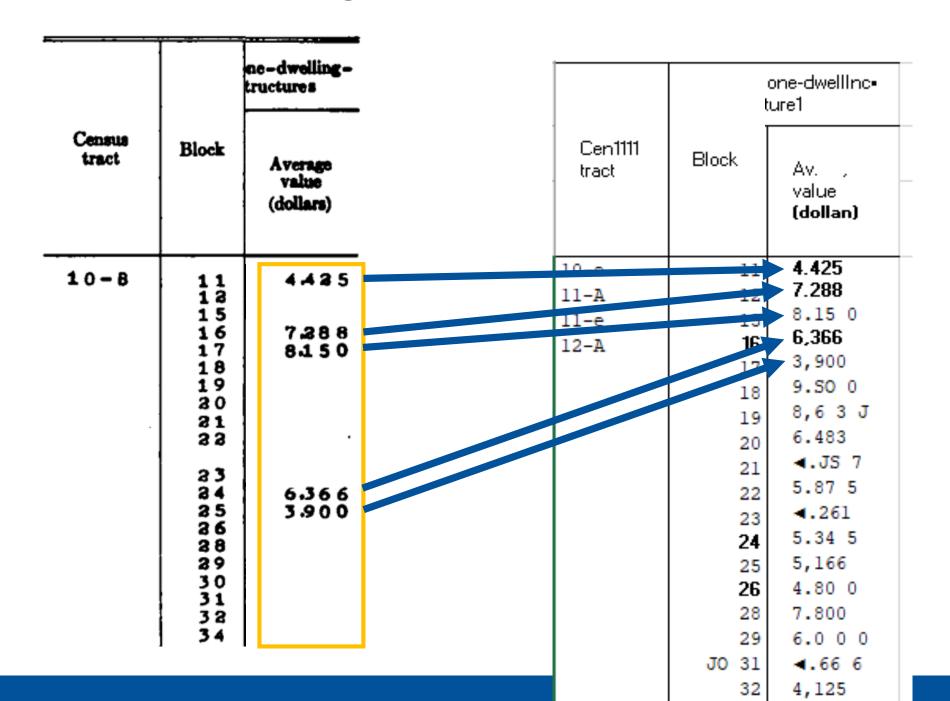
	l	one-dwellinc• :ure1
Cen1111 tract	Block	Av. value (dollan)
10-e	11	4.425
11-A	12	7.288
11-e	15	8.15 0
12-A	16	6,366
	17	3,900
	18	9.SO 0
	19	8,63J
	20	6.483
	21	∢.JS 7
	22	5.87 5
	23	∢.261
	24	5.34 5
	25	5,166
	26	4.80 0
	28	7.800
	29	6.0 0 0
	JO 31	∢. 66 6
	32	4,125

Adobe:

		ne-dwelling- tructures
Census tract	Block	Average value (dollars)
10-8	11 12 15 16 17 18	4.425 7.288 8.150
	19 20 21 22	
	23 24 25 26 28 29 30 31 32	6.366 3.900

	l	one-dwellinc• ture1
Cen1111 tract	Block	Av. value (dollan)
10-e	11	4.425
11-A	12	7.288
11-e	15	8.15 0
12-A	16	6,366
	17	3,900
	18	9.50 0
	19	8,63J
	20	6.483
	21	∢.JS 7
	22	5.87 5
	23	∢. 261
	24	5.34 5
	25	5,166
	26	4.80 0
	28	7.800
	29	6.0 0 0
	JO 31	∢. 66 6
	32	4,125

Adobe: Bad character recognition, relation of rows lost



• Textract:

		Al	l dwelling an	units by or	cupancy		by co	relling undition a	und	Occ	supied dwe	ling un	its		ct monthly rent		ne-dwelling- tructures
Census	Block				Vacant non-	Other	d	No	No running		Persona room	n	Occupied		Average		Average
		Total	Owner occupied	Renter occupied	not dilap., for rent or sale	vacant and non- resident	Number reporting	private bath or dilap.	water or dilap.	Total	Number reporting	1.51 or more	by non- white	Number reporting	monthly rent (dollars)	Number reporting	Average value (dollars)
46- G	1678901234 2222234 222233333	91005 98005 7756204 504 7798 92	24463629563 44334629563 4792922425	46 34 57 30 11 19 7 10 10 10 10 10 10 17 17	2 5 2 4 1	1 2 2 10 3	751462626222 67626222 6756222 6756222 6756	2 1 1 3 1 6 2 3 1 2	21 1 21 6 2312	88 98 97 97 97 97 97 97 97 97 97 97	877377662443 57662443 57662443	3 4 4 2 1 1 2 1 2 3 1 1 1 2 3 1 1 1 2 3 1 1 1 2 3 1 1 1 2 3 1 1 1 1		4145537676 1097433966 16	44.34 43.20 48.83 50.93 38.28 44.43 44.85 42.00 53.10 40.22 41.29 43.88 46.12 47.33 46.83	3251028063 00003619	8.630 16.866 10.

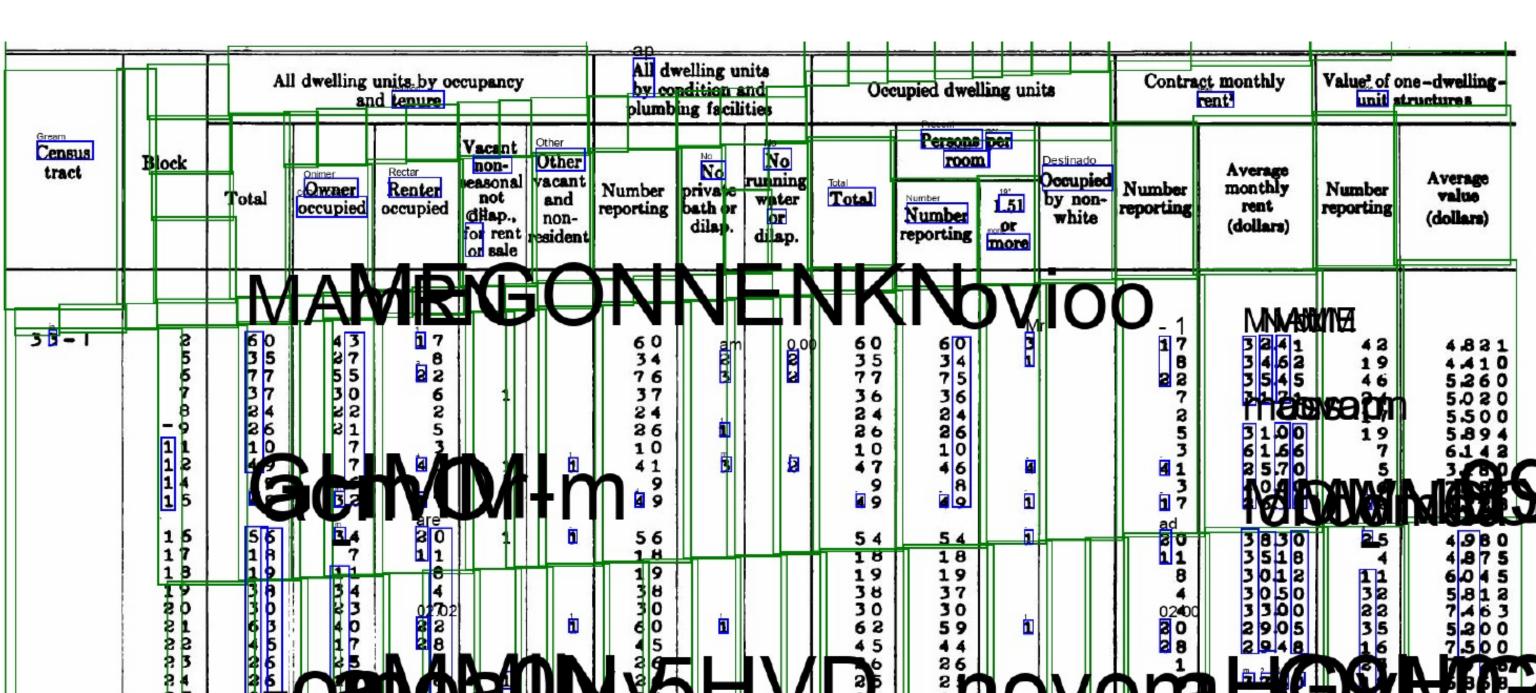
- Textract: when it works it works!
- 1.5% error rate, 0.22% ignoring cell alignment errors (stats for this page only)

									<u> </u>						1 0		
		ÎĀI	dwelling dwelling	units by oc d tenure	cupancy		All dy by co plumb	relling undition sing facili	nite und ties	Occ	dweling dweling dwel	ling uni	ta		cs monthly rent	Value of o	e-dwelling - tructures
Census	Block		Owner	Renter	Vacant non-	Other	Number	Na	No running		Persons		Occupied Occupied	Number	Average	Number	Average Average
8000		Total	Owner occupied	Renter	dilap. for rent	and non- resident	Number reporting	private bath or dilap	water ii.or dilap.	Tota	Number reporting	IL.51	by non- white	Number reporting	monthly rent (dollars)	Number	Ayerage value (dollars)
46-G	1.5	9 1 100 8 0	42	4 6 3 4	2	1	7 5 7 1	į.	ĵ.	8 8 7 8	B 6	3		41 4 1 5 4	4.4.3.4	23 32	8.630 B .6 3 Q
	17	1.00	36	5 7	ie.		94		- 4	9 3	7.1	4		3.4	4.3.20	32	1.5050
	18	7.5	43	30	5	2	66	1	1	7.3	93	4		5 5 3 0	4.8.8 3	1.5	10866
	19	79	5.6	11	а	-	72	1	1	77	77	2		. 7	5.0.9 3 3.8.2 6	31	7.040
	2.0	7.5	4 2	19	4	10	5 6	2		6 1	5 6	1		16	4.4.4 3	3.2	6.921
	2.1	5.6	5.9	7			52	3	2	6 6	66	1		7	4.4.8 5	4.8	7.479
	2.2	\$ 2	3.5	7		3_	4.2	1	1	4 2	4 2			6	4200	3.0	8.300
2	3.4	5 0 6 4	4.6	10 1				į.	į a	47	4 4			120	23.10	2.4	9.102
	24	7 11 11 17 1	5.3	1.0	1		6 2	6	6	6.3	6.3			10	5 3 1 0	2.3	7.434
	2 5	57 57	47	10			5 4			5 7	5 5	İ		, j g	4.0.22	<u>4 0</u>	9.575
	2.6	1.06	47	58		1	98	2	2	1.05	97	2		57	3.9.22	20	9,600
	2.7	5 4	39	25		1	58	3	3	6 4	6 4	1	9	24	4.1.29	30	6.543
	2.8	77	32	4 4	-	1	7.6	1	1	7.6	7 6 6 9	(Z)		7.7	4.3.8 8	20	2.570
	2 9 3 0	7 g 5 B	3 q 4 2	3 6 1 .0	<u>(3</u>	5	5 9	<u> </u>	12	7 5 5 2	6.0	3		3.3	4.6.1.2	2.1	7.869
	31	49	4.2	., 7			4.9	1		49	47	1		6	4.7.33	2.6	8.323
	3.2	6.2	4.5	1 7			6.0	5		62	61	2		16	4.6.8 3	39	8.402 8.128
	200							<u> </u>			- L			10	9 0.4 3	KE	01120

Textract: when it doesn't work...

		A	ll dwelling an	units by o	ecupancy		by ∞	welling undition a	and	Oc	cupied dwe	lling un	its		ct monthly rent ¹		ne-dwelling- tructures
Census tract	Block	1			Vacant non- seasonal	Other		No	No running		Person	'n	Occupied		Average		Average
		Total	Owner occupied	Renter occupied	not dilap.,	and non- resident	Number reporting	private bath or dilap.		Total	Number reporting	1.51	by non- white	Number reporting	monthly rent (dollars)	Number reporting	value (dollars)
33-1	25 67 89 112 14 15	60 35 77 37 24 26 10 49 49	43 27 55 30 22 21 7 6 32	17 8 22 6 2 5 3 40 3	1	1	60 34 76 37 24 26 10 41 9	2 3	2 2	60 35 77 36 24 26 10 47 9	60 34 75 36 24 26 10 46 8	3 1 4 1		17 8 22 7 25 3 41 3	32.41 34.62 35.45 31.71 31.00 61.66 25.70 40.00 26.35	42 19 46 24 17 19 5 6	4.821 4.410 5.260 5.020 5.500 5.894 6142 3180 7.483 3.728
	16 17 18 19 20 21 22 23	56 18 19 38 30 63 45 26	34 7 11 34 23 40 17 25	20 11 8 4 7 22 28 1	1	1	56 18 19 38 30 60 45 26	1		54 18 19 30 62 45 25	54 18 19 37 30 59 44 26	1		20 11 8 4 20 28 1	3830 3518 3012 3050 3300 2905 2948	25 11 32 25 16 27	4.980 4.875 6.045 5.812 7.463 5.200 7.500 7.026 6.868

• Textract: when it doesn't work... it doesn't work! Small input tweaks do not fix error.



- Textract:
 - Sample Size: 169 Pages
 - Catastrophic Failures: 45
 - Moderate Failures: 6
 - Unacceptable page level error: 30%
 - Small errors in table layout can be algorithmically corrected, catastrophic failures cannot

Method

- Isolate table
- Isolate columns
- Tesseract columns
- Structure into table
- Match to labeled data
- Train model to correct Tesseract errors
- Visualize and correct issues throughout
- Final check for internal consistency and vs tract

20

City Block Characteristics

Table 2.—CHARACTERISTICS OF HOUSING UNITS, BY BLOCKS: 1960—Con.

("Total population" contains no persons in group quarters unless preceded by asterisk: one asterisk (*) denotes less than 10 percent; two asterisks (**), 10 percent or more]

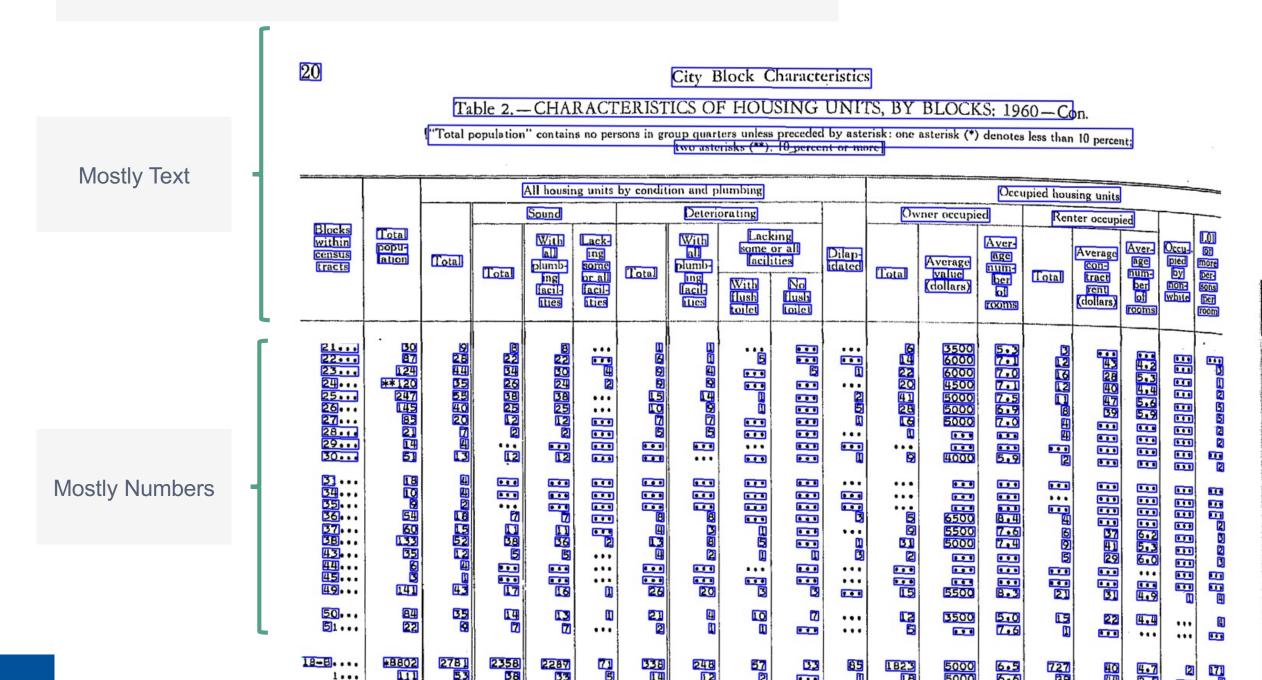
				All housir	ng units	by condit	ion and p	lumbing					Occi	ipied hou	sing units			
				Sound			Deteri	orating			Ov	vner occupi	ed	Ren	iter occup	ied		
Blocks within census tracts	Total popu- lation	Total	Total	With all plumb	Lack- ing some	Total	With all plumb-	Laci some facil		Dilap- idated	Total	Average value	Aver- age num-		Average con-	Aver- age	Occu pied by	1.01 or more
			Total	ing facil- ities	or all facil- ities	Idiai	ing facil- ities	With flush toilet	No flush toilet		Total	(dollars)	ber of rooms	Total	tract rent (dollars)	ber of rooms	non- white	per- sons per room
21 22 24 25 26 27 28 29 30 31 35 36 37 38	30 87 124 **120 247 145 85 21 14 51 18 10 9 54 60 133 35	984550074B 4428522	8 22 34 26 38 25 12 2 12 13 13 13 13 13 13 13 13 13 13 14 15 15 15 15 15 15 15 15 15 15 15 15 15	822 3024 3825 12 22 12 1365	42	16 9 15 10 7 5 8 4 13	1149914997558382	1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1	5	2551	6 14 220 41 28 16 1 1 9	3500 6000 4500 5000 5000 5000 4000 6500 5500	5.1015.9 7.1015.9 7.1015.9 7.1015.9	3 12 16 12 11 18 4 4 4 2	43 28 40 47 39	4.2 5.3 4.4 5.6 5.9		3 1 2 5 5 2 2 2 2 3 2
44 45	6 3	4	•••	• • • •	•••	•••	•••	• • • •	•••	•••	***	•••		,,,		•••		3
49	141	43	17	16	1	26	20	''3	***	***	15	5500	8.3	21	31	4.9	***	4
50	84	35	14	13	1	21	4	10	7	•••	12	3500	5.0	15	22	4.4		a
51	22	9	7	7	•••	2	1	1	•••	••••	5	•••	7.6	1	••••	• •••	:::	
18-B	*8802 111	2781 53		2287 33	71 5	338 14	248 12	57 2	33	85	1823	5000	6.5	727	40	4.7	2	171

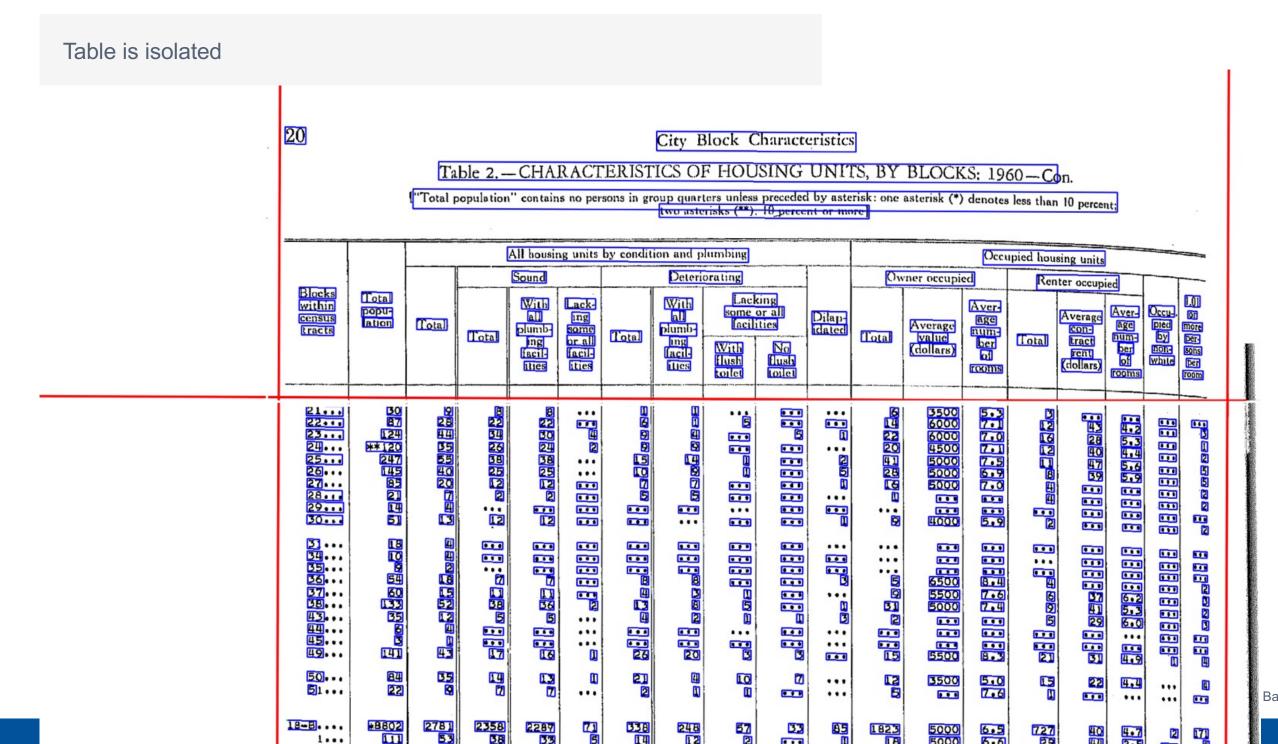
Pass through Textract

20City Block Characteristics Table 2.—CHARACTERISTICS OF HOUSING UNITS, BY BLOCKS: 1960—Con. "Total population" contains no persons in group quarters unless preceded by asterisk: one asterisk (*) denotes less than 10 percent; two asterisks (**). 10 percent or more

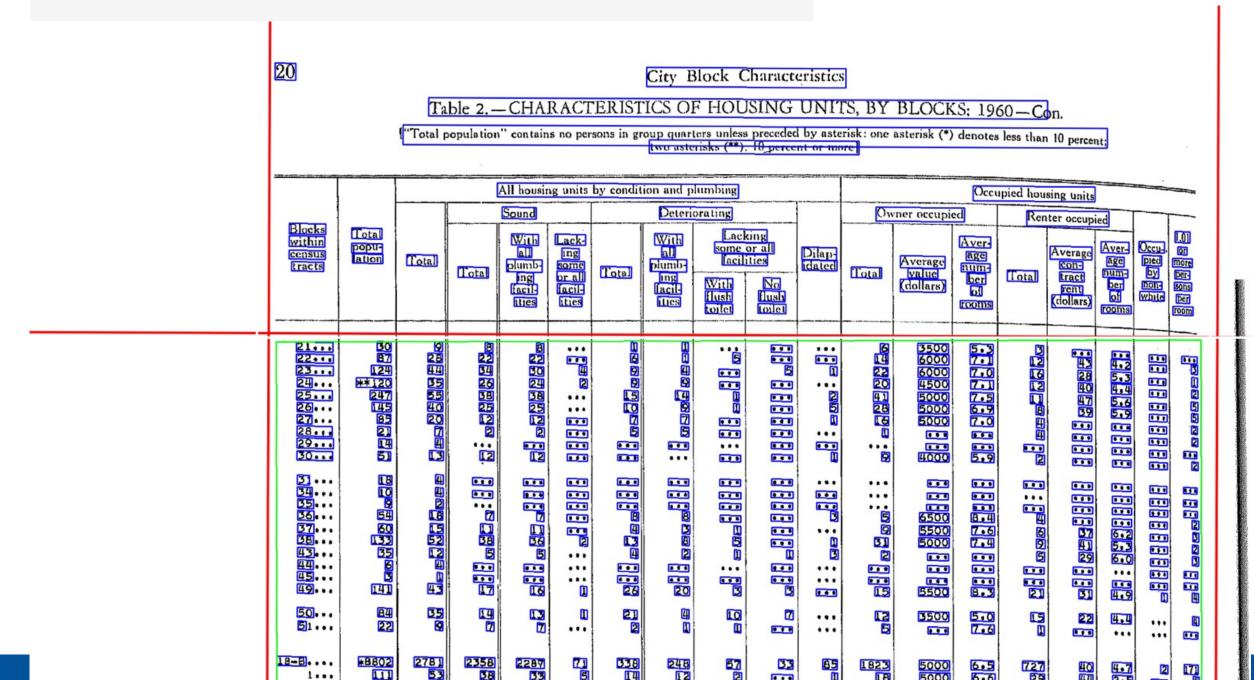
			[All housir	ng units	by condit	ion and p	lumbing					Occi	ipied hou	sing units			
				Sound			Deteri	orating			Ow	vner occupi			iter occup			_
Blocks within census tracts	Tota popu- ation	[Cotal]	Total	With All plumb- ing facil- ities	Lack- ing some or all facil- ities	Total	With all plumb- ing [acil- ities	Lacksome facil		Dilap- idated	Total	Average value (dollars)	Aver- age num- ber of rooms	Total	Average con- tract rent (dollars)	Aver- age num- per of	Occu pied by non- white	DET Sons Teom
21 22 23 24 25 26 27 28 29	** 120 ** 120 147 149 149 149 149 15		22 54 26 58 25 25 27 27									3500 6000 6000 4500 5000 5000	5.3 7.1 7.0 7.5 6.9 7.0			5.6 5.9 •••		
	13 54 60 133 55 60	Leenshare				26							0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0	6 5 5 2	51	6.0 5.3 6.0 6.0 6.0		
50	84 22	35 9	1.4 7	[13] [7]		21 2			7	:::	12	3500	5•0 7•6	E3 (1)	22	4.4	:::	
18-B	#8802 111	278 J	2358	2287	71 5	338 174	248	57	33	65	1823	5000	6.5	727	40	4.7	[2]	17]

Find where, vertically, we go from mostly alpha to mostly numeric

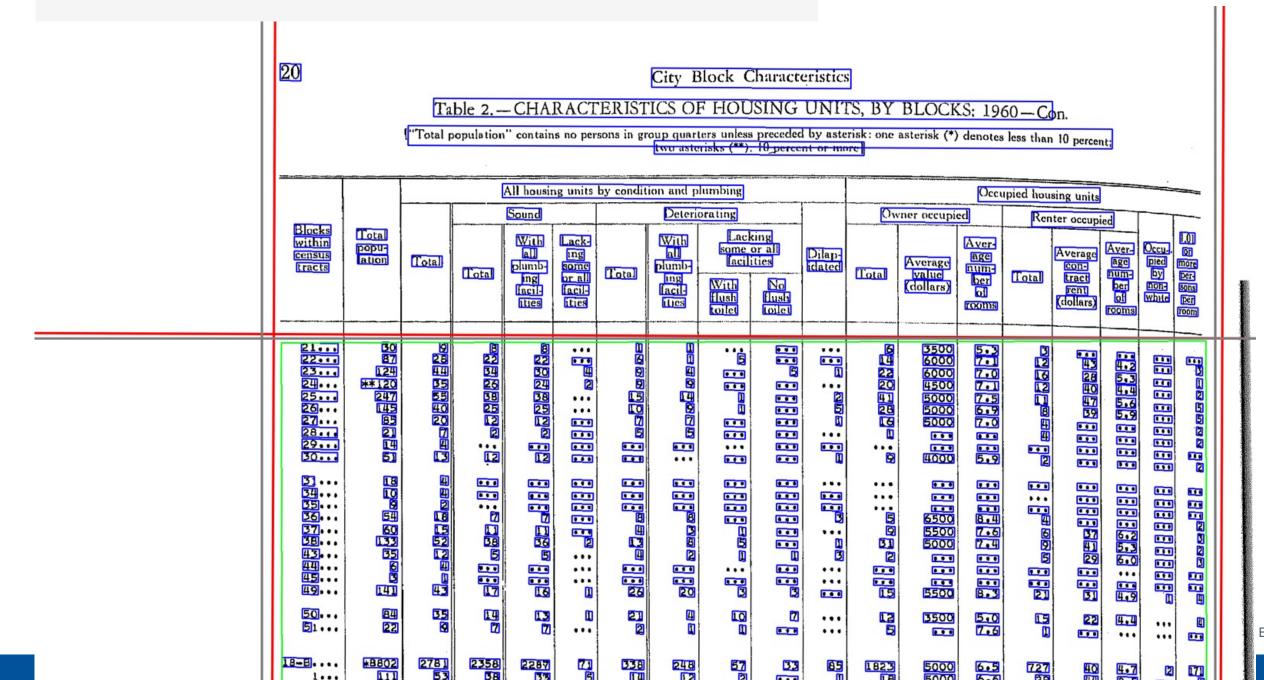




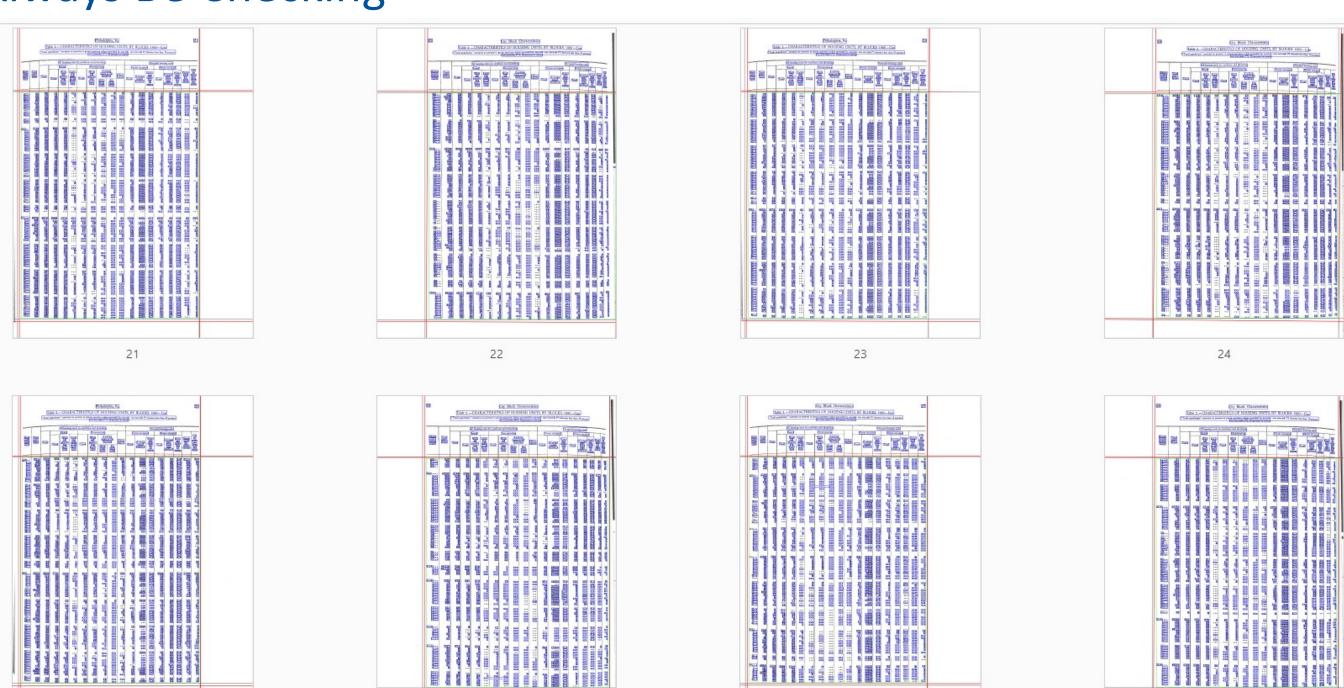
Find the rotated bounding box that contains all the body bounding boxes



Rotate image around center of table – image is straightened



Always Be Checking

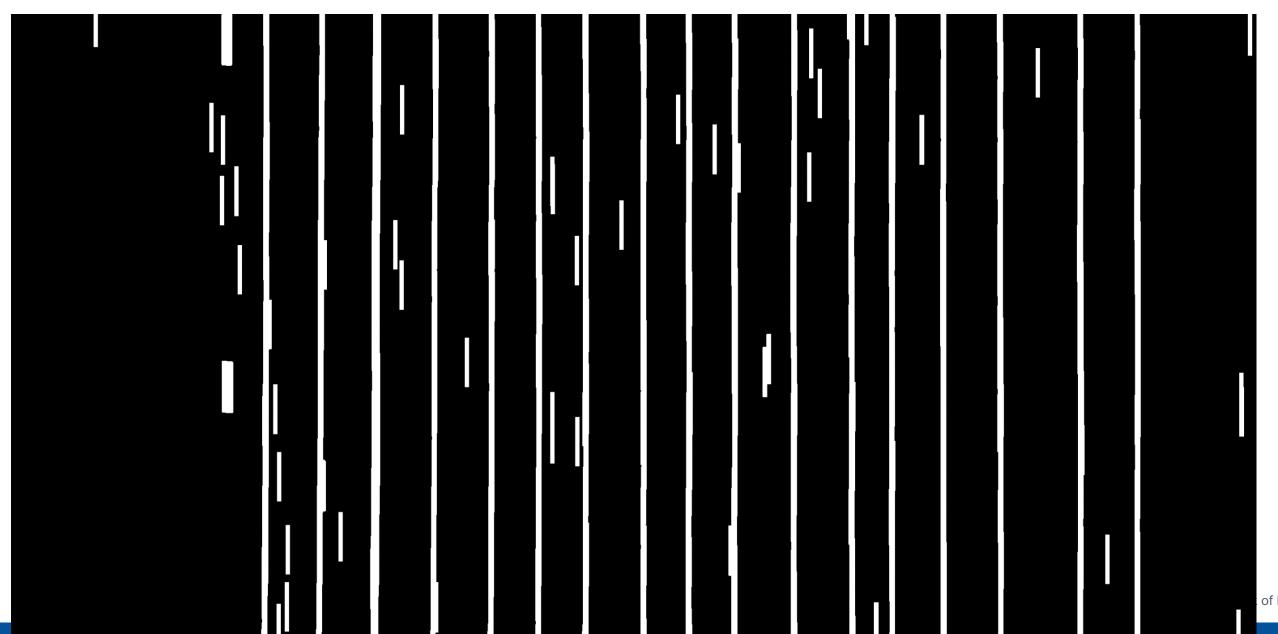


21-U	23456789 112	8335 434 230 132 132	591755922 64332111	15128 97121	3 1	1	863448053 4438853	8		863542043 432243	85335 4320 4320 43		·	1 6 8 8 8 7 2 1	42.62 53.25 45.87 50.37 49.71	6 1 5 1 2 2 3 2 3 9 6 2	8.696 9.411 10.845 10.046 10.272 12.833 3377 7150
	13 15 16 17 18 19	26 36 9 15 9	3 3 1 6 2	9 5 3 3	3	1	15 35 9	3 1	3	1 2 3 6 9	12 35 9	3		11 5 3 3	5 3.1 8 2 8.4 0 4 9.6 6 4 2.3 3	3 31 4	4 3 3 3 3 6 3 3 2 9 0 0 0
	2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2	43 34 6	28 21 3	15 13 3			5 9 42 34 6	6		9 43 34 6	9 43 34 6			1 3 1 2 3	5 6.0 0 4 5.8 4 3 9.0 0 6 8.3 3	1 2 3 1 5 3	9.07 B 9.833 13.000
	a 6 a 7	28 29 19	23 15	4		B	19			19	19	1		1 3 4	3 2.3 3 4 8.7 5	13 13	1 2 5 0 0 9 0 0 0 8 7 3 0
21-V	1234567890 10	59 576920574 105525	9026730228 4232226	33 33 16 98 146 146	1 1 1	2	1052809574 59 5302574	10 35 8	3	59 5M0RMR5	59 55989965 5598885	1	1	3 4 3 4 1 0 1 2 6 6 1 2 1 4 5	47.55 48.08 70.00 32.50 43.41 44.90 52.75 37.78 34.60	15 43 40 25 18 15 19 147	7.28 6 8.08 1 5.96 5 8.18 0 1 0.23 3 8.33 1 5 7.30 0 7.36 1
	1346789012	144024497530	126 31 19 37 28 37 28 431	14827622798	14	1.	14401497489	431 2 2 51	2 1 1	1490 18497 18497 1891 1891	99908497589 18148151	8		13 8 2 3 6 12 7 9	51.76 31.50 31.00 40.50 30.41 22.57 36.66 35.33	1 2 2 2 7 1 8 5 5 2 7 3 6 7	7.9329 8.6273 9.0736 7.43001 4.9775 6.0220 5.500

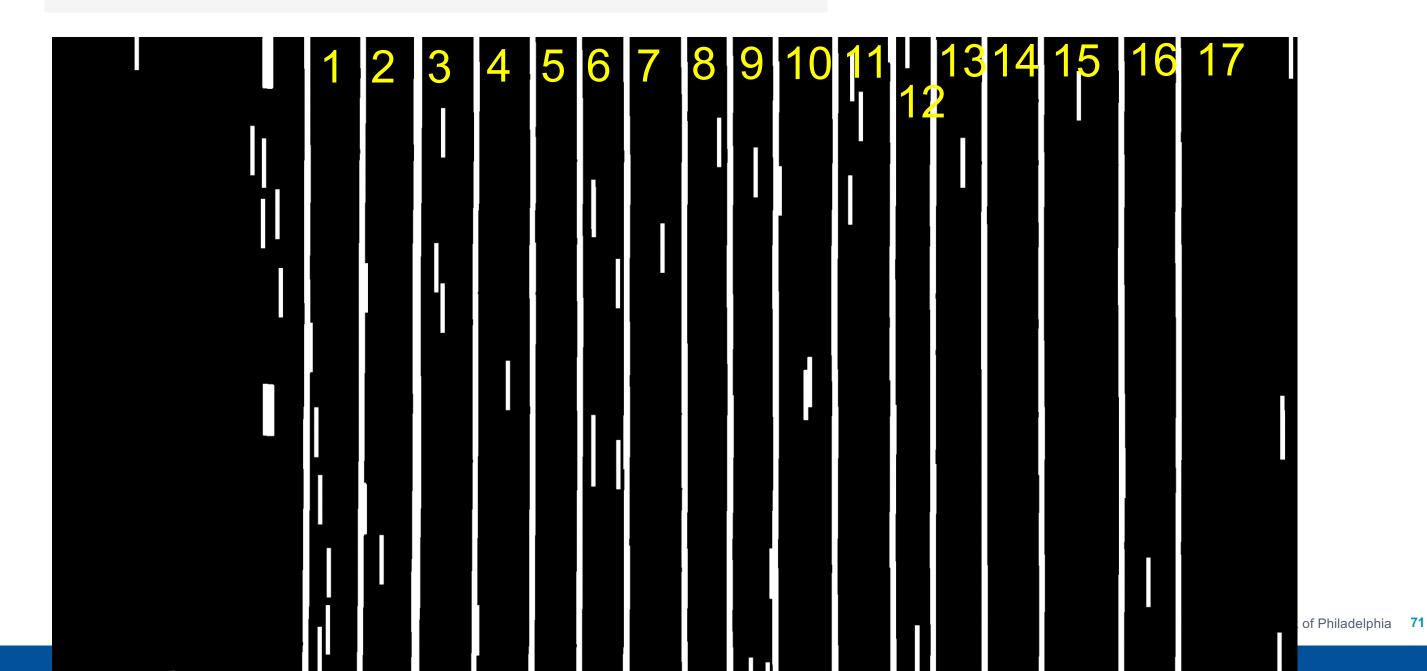
Find everything that *could* be a character. Be aggressive, recall is important

	_	_		_			_					_		_	_			
20-0					3 3 0	1		2							4 2 5 2 5 5 2 5 6 5 5 7 5 5 7 6 5 7 7			
		연명 연료의 연료 연료	2 9 5 8		3	; 6		3	3	7 7 7 7 8	13 13 13 13 13 13 13 13 13 13 13 13 13 1	3			55.0 86.0 87.0 87.0 88.0	3 D	5 0 0 0 0 0 0 0 0 0 0 0	
·	ì	_				, 	55 45 45 45 45 45 45 45 45 45 45 45 45 4	5			5 2 3 3 5 5				4 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2	1 6 2 0 2 0 1 0	9.07 B 9.85 S 0.00 0 0	
	20 20 20 20 20 20 20 20 20 20 20 20 20 2	11 9 11 4 11 4	11 5 11 5	11 [3]		8	11 4 21 8 11 9				0 4 2 5 0 9				3 2.5 3 4 8.7 5	<u>6</u> 11	11 2 5 0 0 2 0 0 0 8 7 5 0	
		DINCORNOLUE ENGRANCIONO ENGRAPERO ENGRANCIONO ENGRAPERO ENGRANCIONO			田 田 田 田田		HONMOCOMEN HECHEDENIO OR CHOOKEND HECHEDENIO H		3	는 기업을 하는 지원 기업을 보고 기업을 보고 있는데 보고 기업을 보고 있는데 보고 기업을 보고 있는데 되었다.	POURDED DOCKERLEDE		1					
	SS 30	2 C	4 5 0 1	<u> </u>	1		5 2 1 9	5	<u>11</u>	1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1	5 N	. [<u>7</u>	28.57 26.66	7	<u> </u>	

Isolate and smear (slightly horizontally, aggressively vertically) what is left



Find (N-1) longest lines that are nearly vertical, N=# of columns



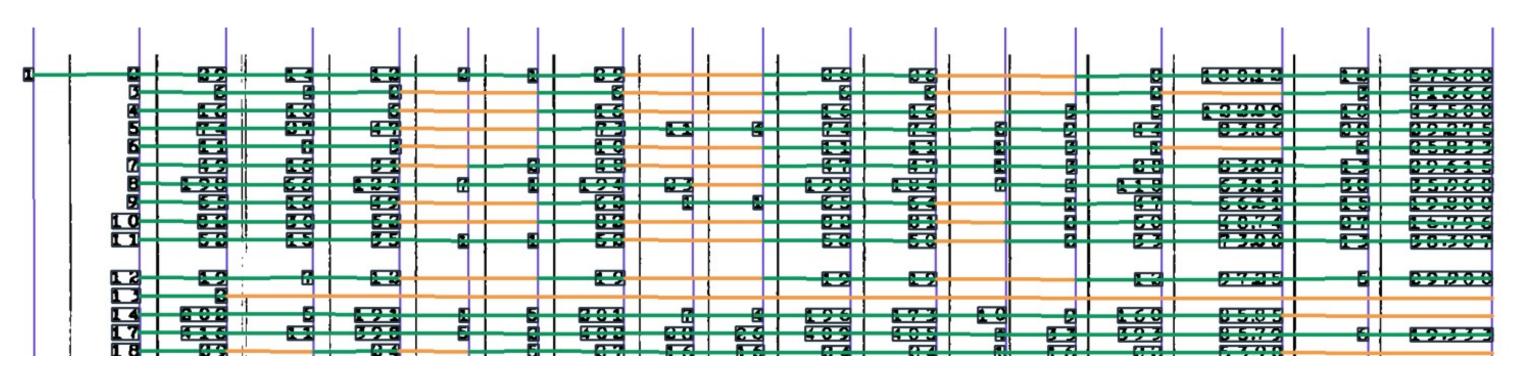
Columns are isolated

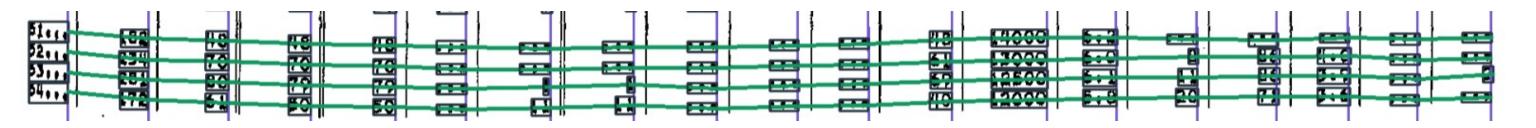
	_			_	_	_	_	_	_	_	_	_	_	_		_	
21-U	2345678912	3235420553 86343221	591755922 64332111	151 289 71 21	3 1	1	303448053 43881	8		86348043 86348881	093542043 2043 213		·	1.682887 21	4 2.6 2 5 3.2 5 4 5.8 7 5 0.3 7 4 9.7 1	151883968 64338	8.696 9.411 10.845 10.272 12.833 7.150
	13 15 16 17 18	2 16 36 9	3 3 1 6	9 5 3	3	· 1	15 35 9	3 1.	3	1 2 3 6 9	12 35 9	3		11 5 3	5 3.1 8 2 8.4 0 4 9.6 6	3 31 4	4 3 3 3 3 6 3 3 2 9 0 0 0
	20 21 22 23	5 9 43 34 6	2 28 21 3	3 3 15 13 3	:		5 9 4 2 3 4 6	6		5 9 3 3 6	5 9 43 34 6			3 3 13 12 3	4 2.3 3 5 6 0 0 4 5 8 4 3 9 0 0 6 8 3 3	1 2 23 15 3	9,07 B 9,83 3 1 3,0 0
	24 25 26 27	14 28 2 19	13 23 15	1 3 4		8	14 28 19			1 4 2 6 1 9	14 26 19	1.		1 3 4	3 2.3 3 4 8.7 5	6 13 13	12500 9000 8730
21~V	1234567890 1	567 5367 5392 10357 10357 10357	9026750228 16 4252214	345 333 989 1699 116	1 1 1	2	1052809574 59 58028055	10 3 5 ន	3	55 55000000000000000000000000000000000	59 55959965 59 5598885	1		344 302 102 102 112 113 15	47.55 48.08 70.00 32.50 43.41 44.87 42.75 52.78 34.60	14 40 159 1190 147	7286 8081 5965 8180 102335 123315 7361
	1346789012 111111222	1 4 1 4 4 0 2 2 4 4 9 7 5 5 2 0	126189875831 1875831	148835622798	1 4 1 1	1.	1440 1440 1849 1497 1489 151	431 2 3 51	2 1 1	0900497509 4310140151	1399 3310 22497 1497 1529	8		13823622796	51.76 31.50 31.00 40.50 30.41 23.57 36.66 35.33	1 22 7 8 5 5 0 7 6 7 3 7 6 7	7,922 1 9,023 1 9,073 0 7,657 2 4,97 2 8,020 5,50

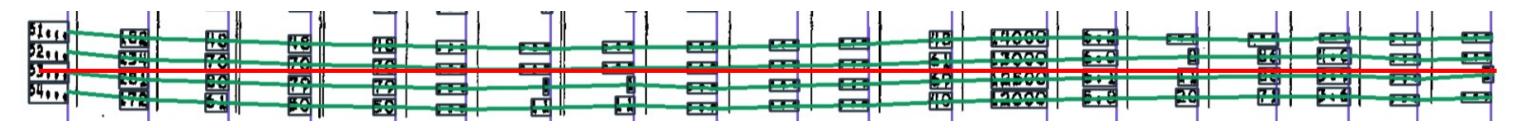
Tesseract Each Row in Each Column

- Tesseract highly sensitive to input parameters, but flexible and governable
- Use restricted character set and character level confidence
- Collect character level text, bounding boxes and confidence

- Start with block column (always populated)
- Look right to find the two-way unique nearest neighbor for each row, requiring the angle to the nearest neighbor be similar for all rows
- Create a synthetic cell for cells that do not have a nearest neighbor conforming to angle and distances of other cells in column
- Repeat moving out left and right to cover all columns
- Create PDF of all pages to scan for errors







Train and Apply Custom Model

- Match cells to training data Washington DC, Mapping Segregation
- Train random forest model at the character level
 - Pixel value by position in bounding box
 - Tesseract predicted text
 - Tesseract confidence
- Grid search with cross validation to tune hyperparameters
- Apply model (out of sample) to remaining cities

Identify Internal Inconsistencies, Compare to Tract Totals

- Internal consistency, e.g. Owner Occupied + Renter Occupied = Occupied
- Check for outliers at column level
- Compare stats to tract totals, accounting for suppression
- Make corrections easy with Excel tool

tract	ocr	human
	ocr	Hulliali
TOTAL	TOTAL	
1-4	1-A	
1-8	1-B	
1 - C	1-C	
2 - A	2-A	
2 - B	2-B	
2 - C	2-C	
3-1	3-A	
- -		

Caveats

- Approach requires some customization per dataset
- Manual steps remain (and probably always will)
 - Identifying unusable scans
 - Identification of page ranges in source documents (missing pages)
 - Always be checking
 - Tract transcription is still manual

Current State

- Scaling work to all 16 cities for 1950
- Refining issues with 1960 model
- Starting 1940 work
- Textract for assist with tract identifiers?
- Claude or other LLM based service for first cut?

Summary Federal Reserve Bank of Philadelphia 84

Summary

- We are working on digitizing the historical Censuses of Housing Block Statistics, 1940 to 1970.
- Our goal: Develop & release data for 16 cities, training & validation data, and methods & code.
- The three major tasks are digitizing block **shapes**, the block **situations**, and the block **statistics**.
- This is a work in progress; Questions and comments welcome!

Thanks!

