

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

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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.



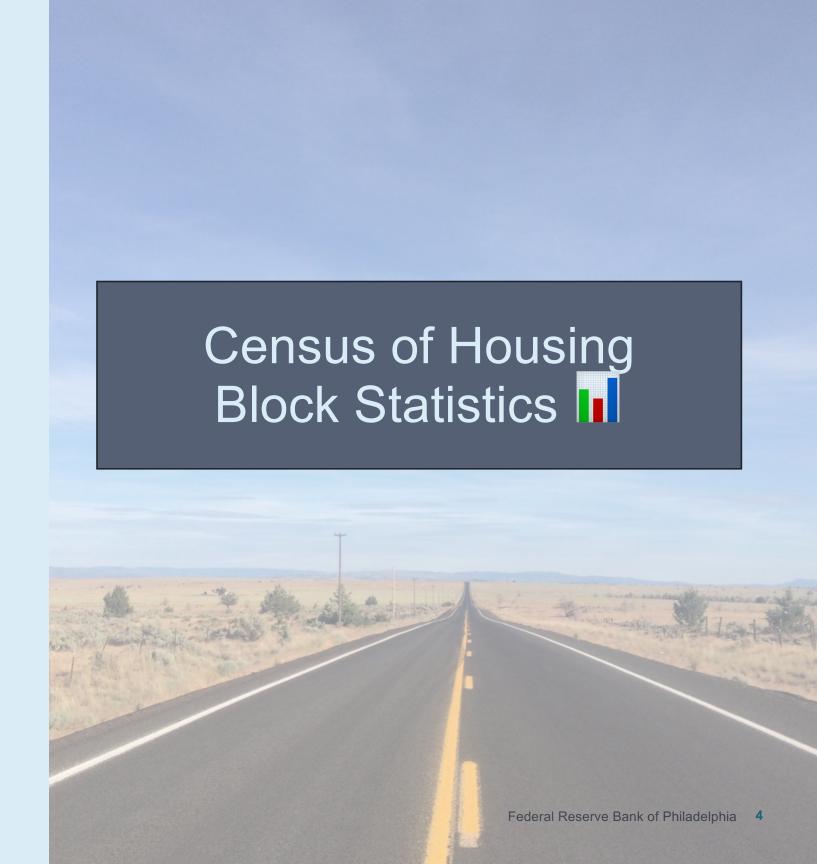
Digitizing Block Statistics

- What
- Why
- Goals
- Tasks and Challenges



Digitizing Block Statistics

- What
- Why
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- 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 34 35 37 38 39 40 41 42 43	21 41 334 268 111 248 282 283 241 437 88 282 283 283 283 283 283 283 283 283	3436894258334 9011364416 35334 901364416 3533	11 20 16 13 18 12 11 6 3 3 3 7 7 7 4 4 12 2 11 1 6	17 20 19 23 20 23 20 22 20 22 20 22 20 22 20 22 20 22 20 22 20 20	631111133211443133	1	342 367 494 225 332 332 331 436 441 736 493	2 1 1 2 2	4 1 3 2	4 4 2 3 9 2 2 2 2 2	3 4 4 4 6 6 2 3 4 3 8 4 1 4 4 3 3 8 2 9 7 8 3 3 9 6 4 1 1 7 3 4 4 9 3	28 400 35 37 48 24 24 35 31 33 39 39 30 66 68 35 47 3	1 9 1 1 5 1 2 1 1 2 2 7	28 39 37 48 42 44 328 33 99 039 44 468 35 463	3 5 2 1 4 3 4 1 4 6	37 48 24 238 32 34 30 27 43 26 44 71	17 15 24 8 15 33 29 6 19 18 21 7 14 25 23 23	35 23 29 6 31 10 15 10 14 125	1907 1154 11505 1930 1144 2023	68 1.44 1.11 1.82 1.22 1.00 5.33 1.00 3.55 4.77 1.50 1.00 4.4	11686255855	2223314 331334 331426441 3644164	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?

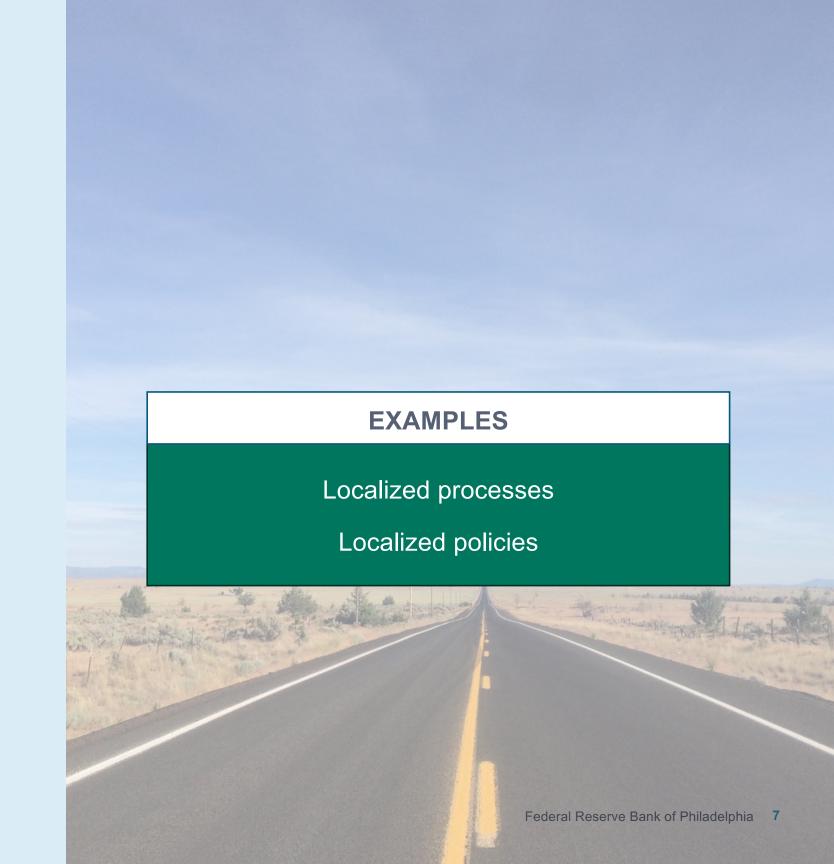
Studies of housing investment and maintenance and long-run urban dynamics.

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

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

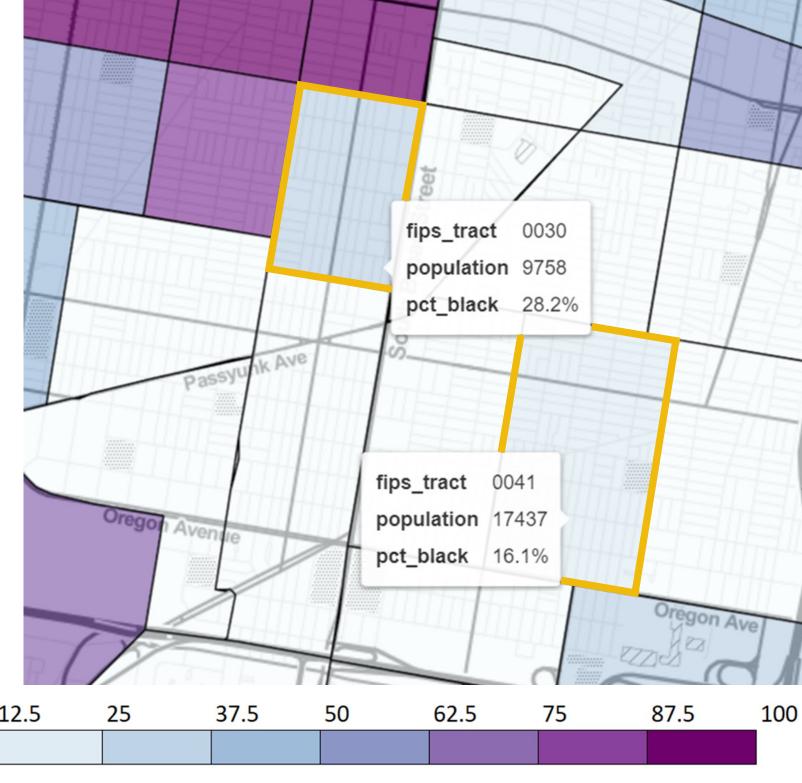
Digitizing Block Statistics

- What
- Why
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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



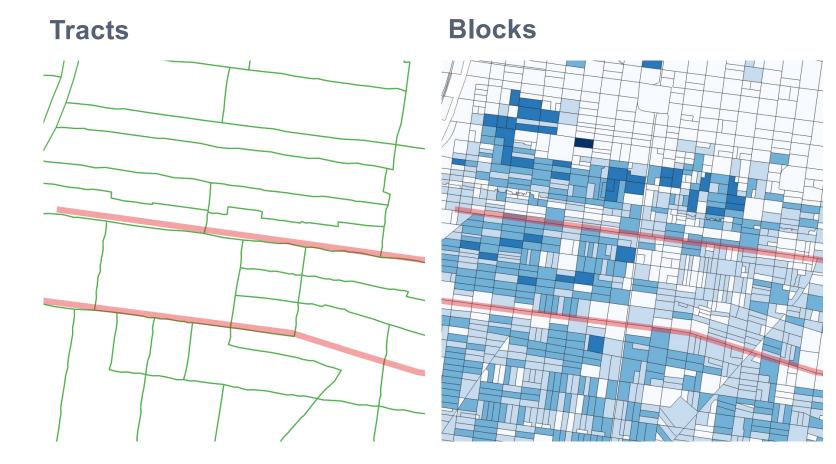
Percent of population Black

No Pop

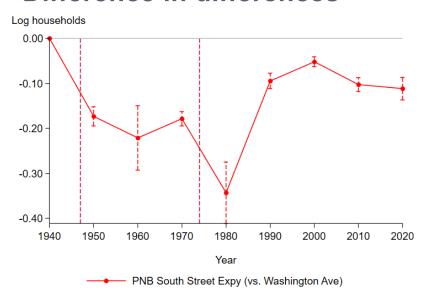
Localized Policies

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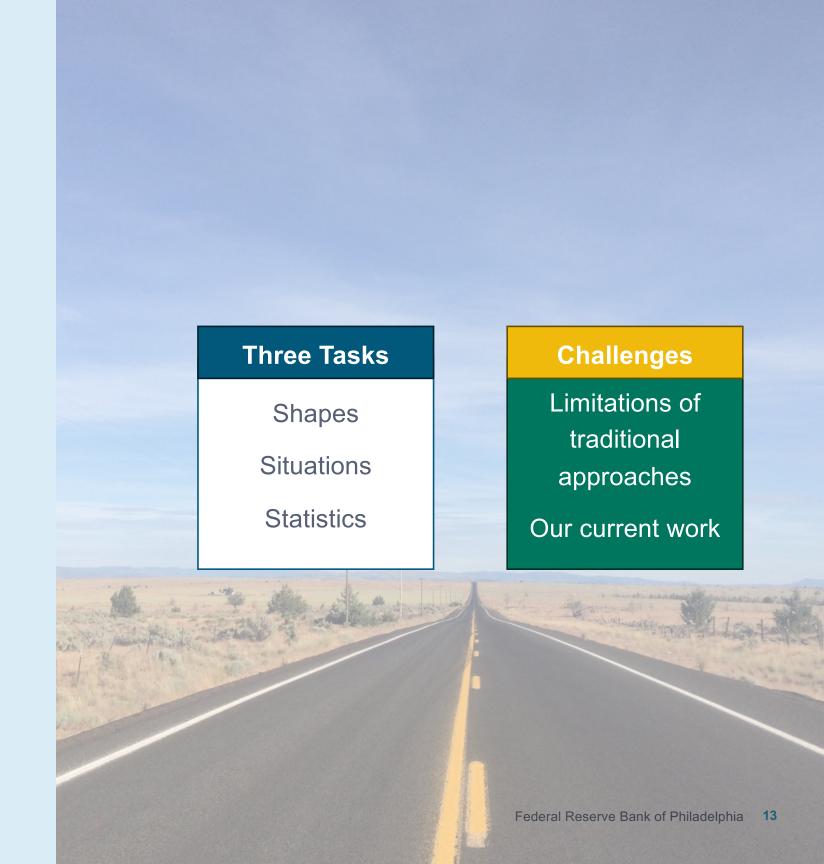
Boston Philadelphia Pittsburgh Columbus Washington Baltimore San Francisco Los Angeles Atlanta Houston

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



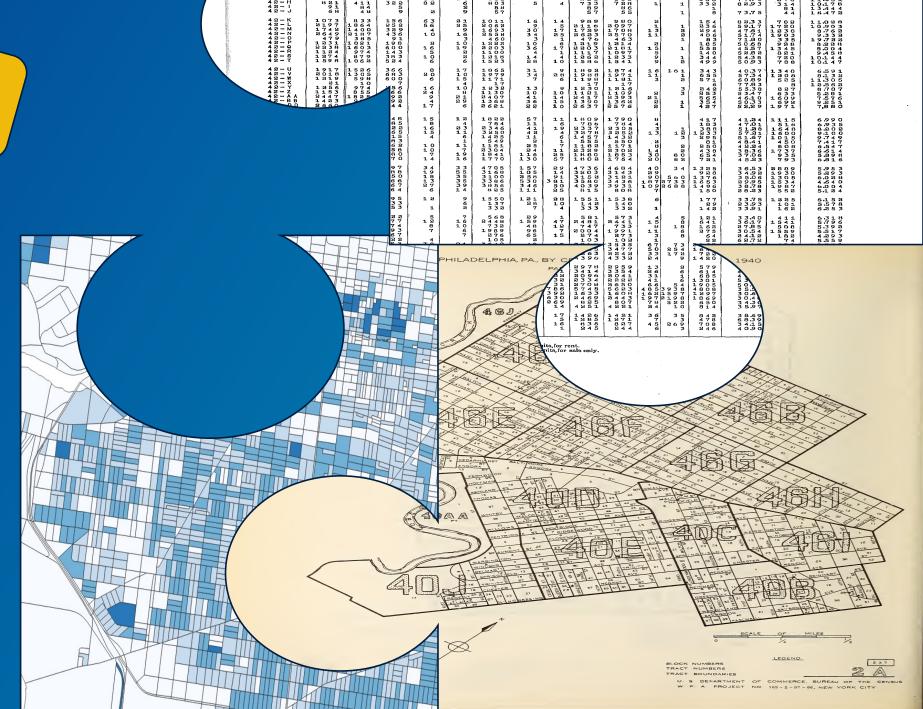


Shapes

Blocks need to know their:

2 Situations

3 Statistics



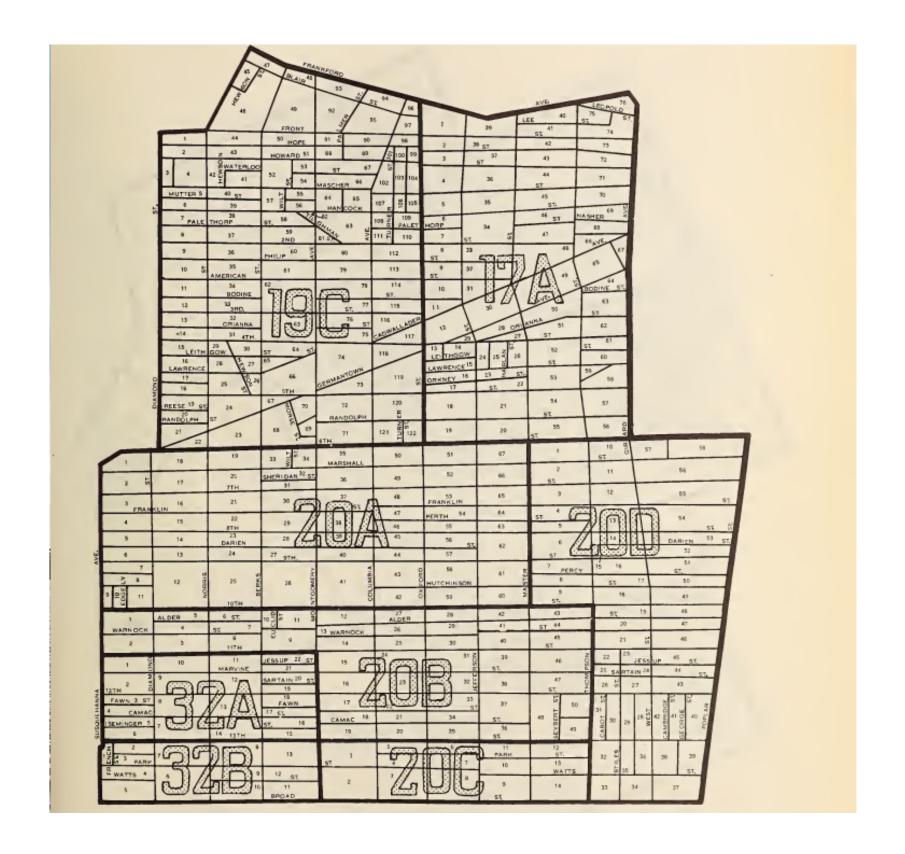
3 Tasks, 3 Pieces of Data

Shape 2 Situation 3 **Statistics**

Segmenting Block Shapes from Maps

Our Ideal Process Has Only Three Steps

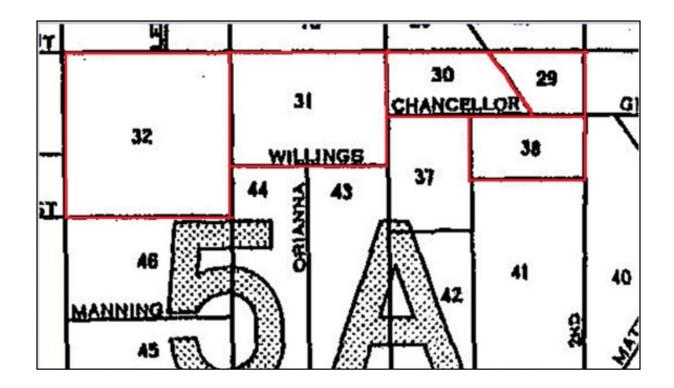
- 1. Identify closed loops of black ink.
- 2. Call them all blocks.
- 3. Declare victory.

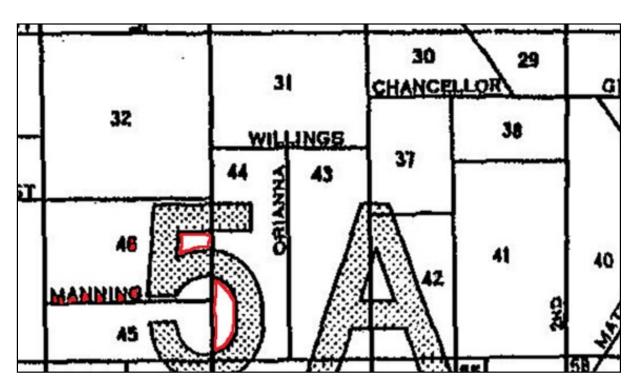


Unfortunately, This Process Fails Spectacularly

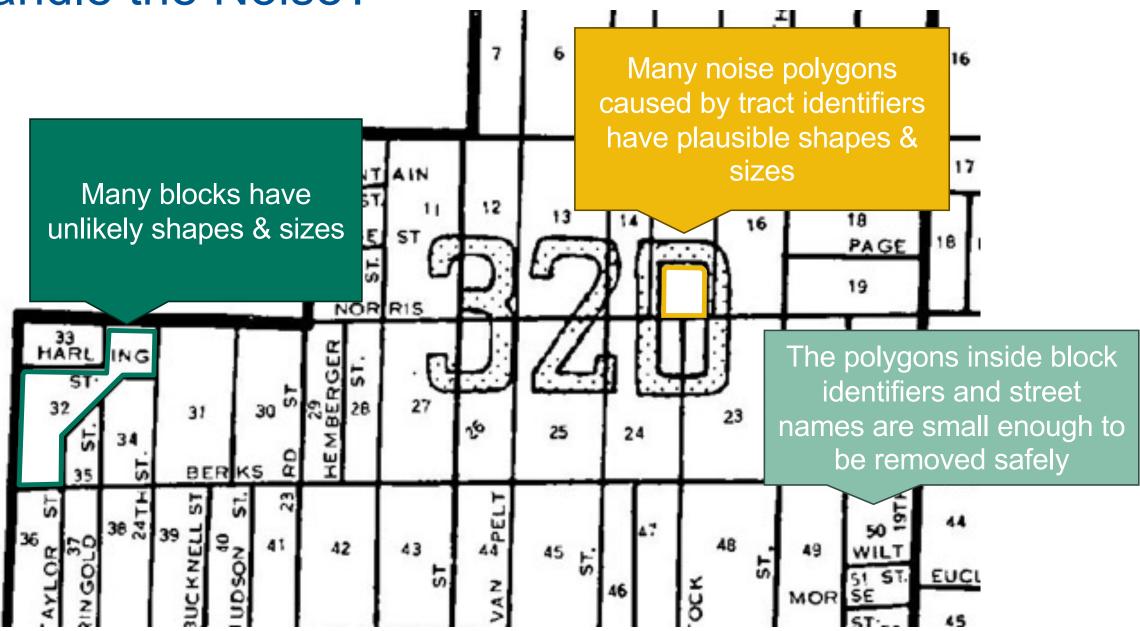
While many closed loops of black ink are blocks....

Many closed loops of black ink are not blocks 🕾



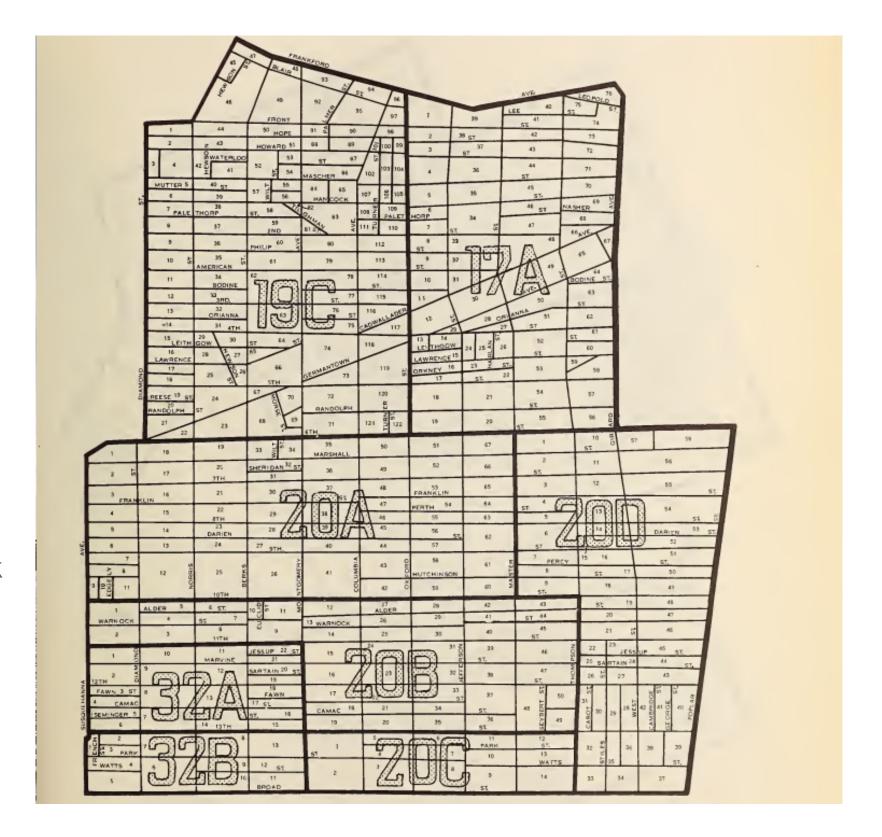


Can We Handle the Noise?



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

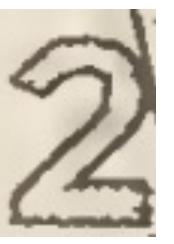
Arbitrary Rotations, Inconsistent Scale, Shape, and Font, Noise/Interference from other features.

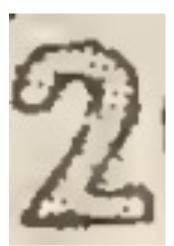
Low-confidence matches

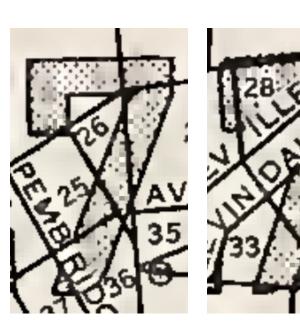
- If we accept low confidence matches of enough templates, everything starts to look like a tract identifier.
- Especially problematic with blocky characters like 1 and E.





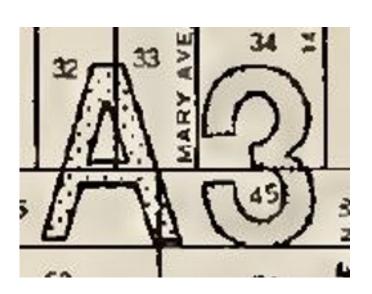


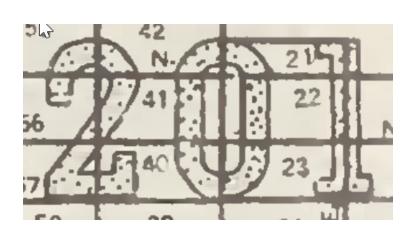




Traditional Method 2: Matching Patterns

- 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

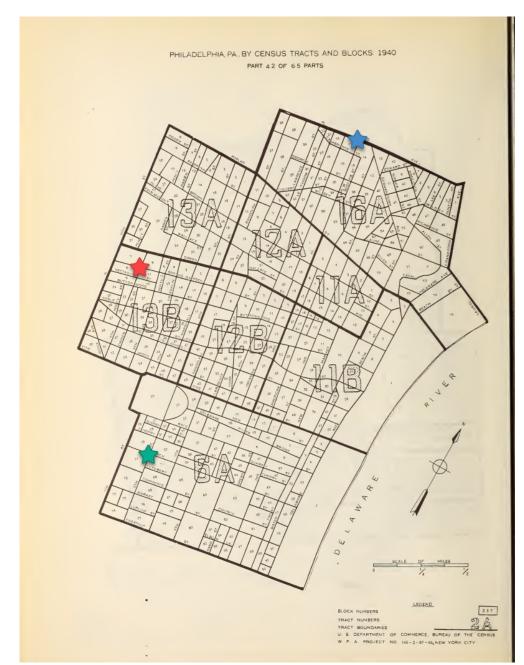
2 Situation

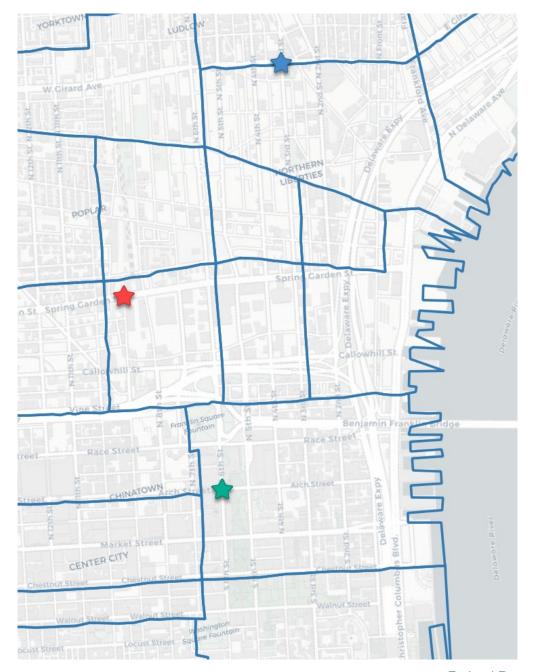
Geo-Referencing Maps

3 Statistics

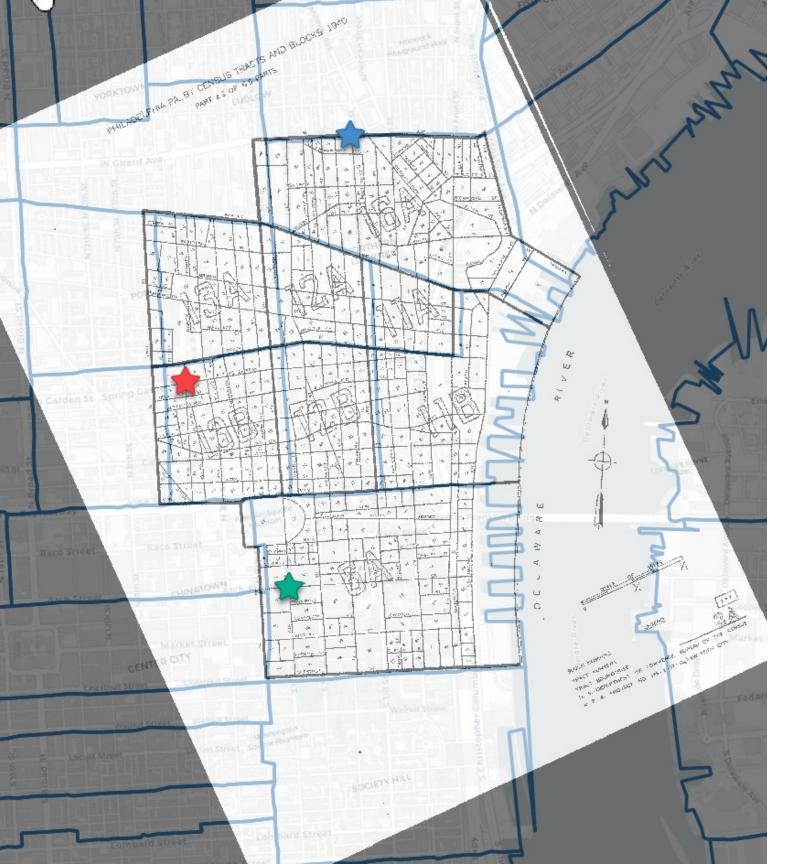
Just Keep Clicking











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?

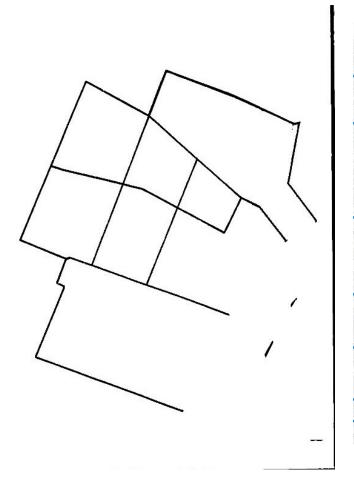
- Faster!
 - We want to process many maps.
- More accurate!
- How much of this can a computer do for us?

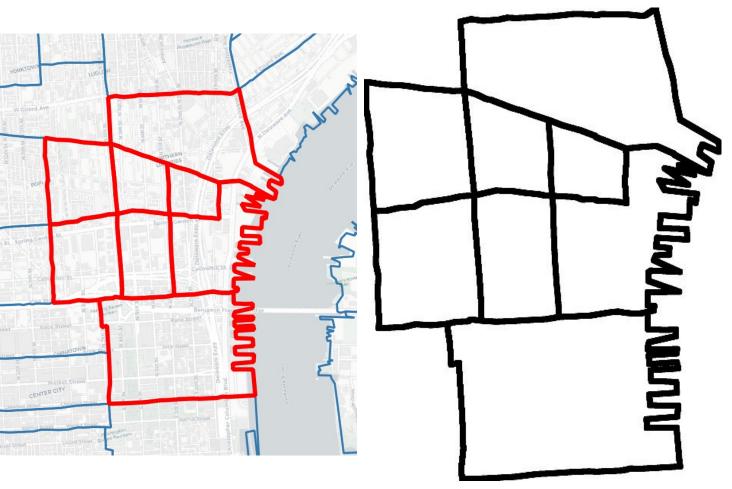
Simplifying the Problem





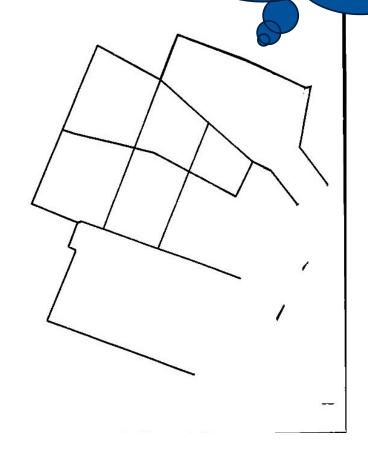


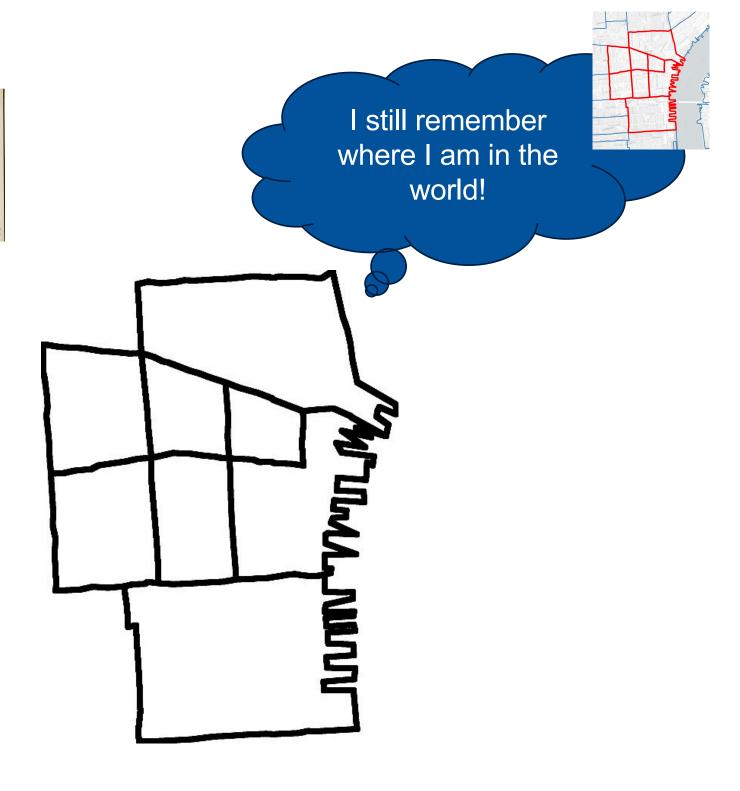




How Hard Could This Be?

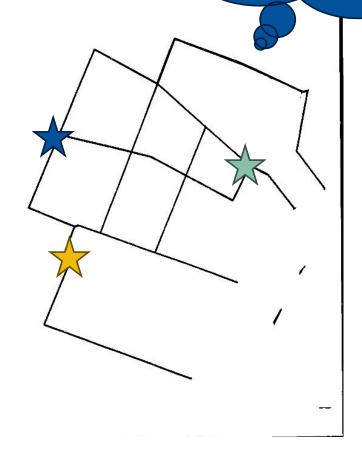
I still remember the block boundaries!



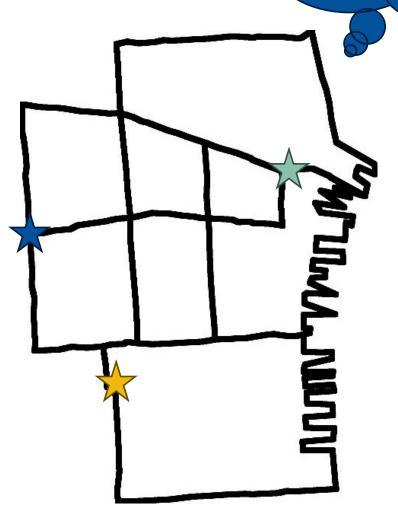


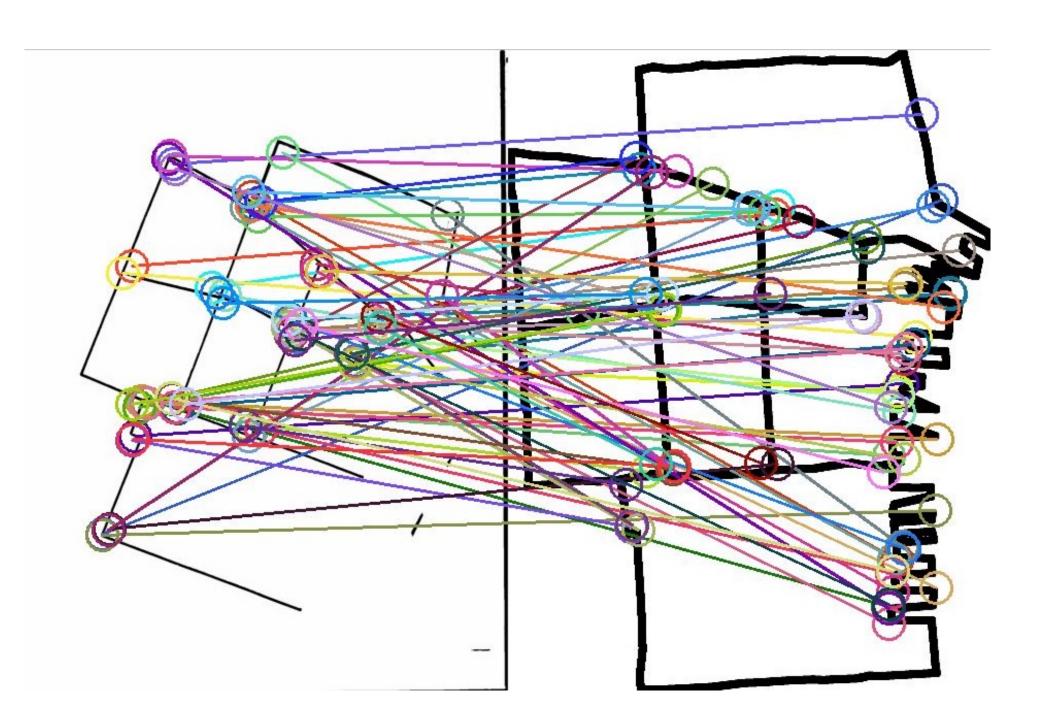
How Hard Could This Be?

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

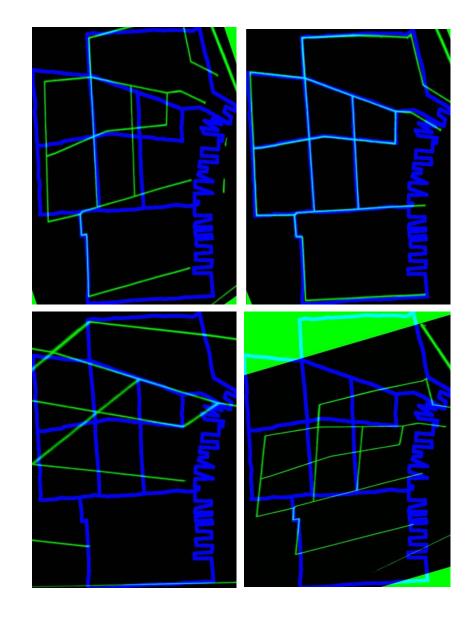
For selected links, how much do maps overlap?

3 Repeat

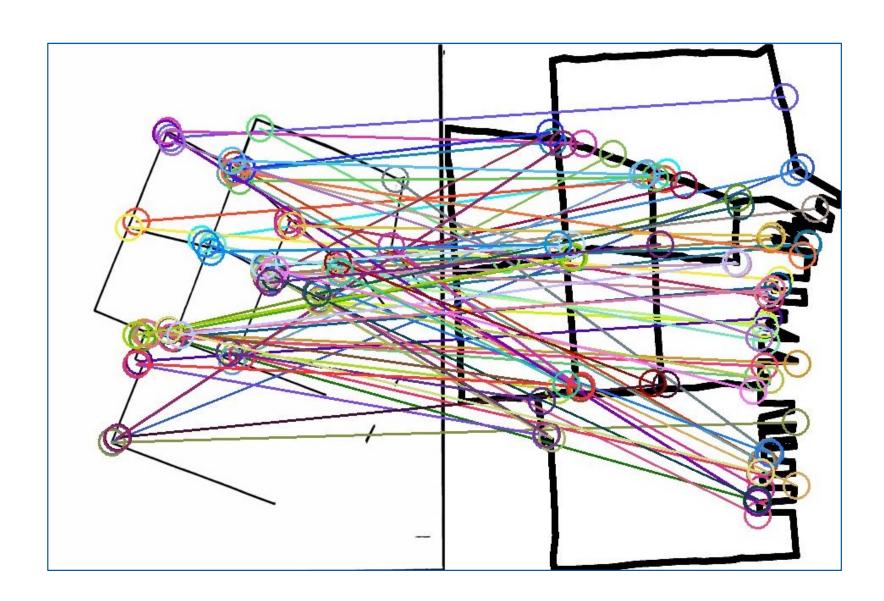
Keep best guess.

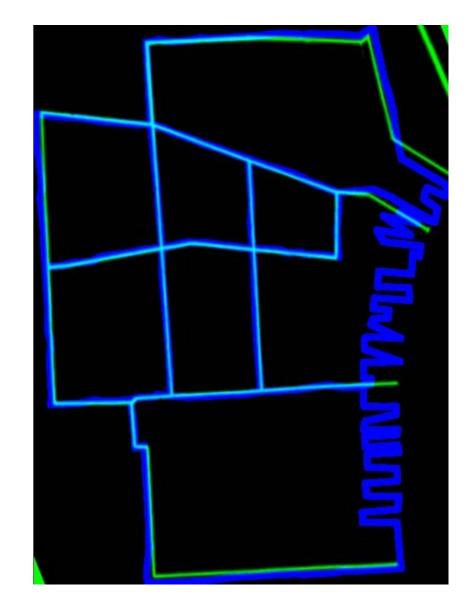
Selecting Links

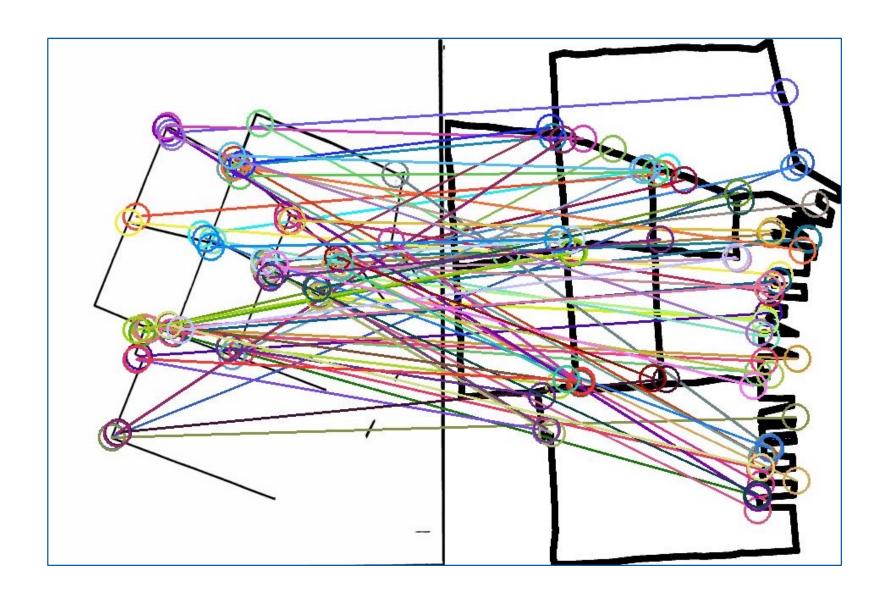




Selecting Links 2









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)
10-8	11	4.425
-	11 12 15 16 17 18 19 20 21 22	7.28 8 81 5 0
	23 24 25 26 29 30 31 32 34	6.366 3.900

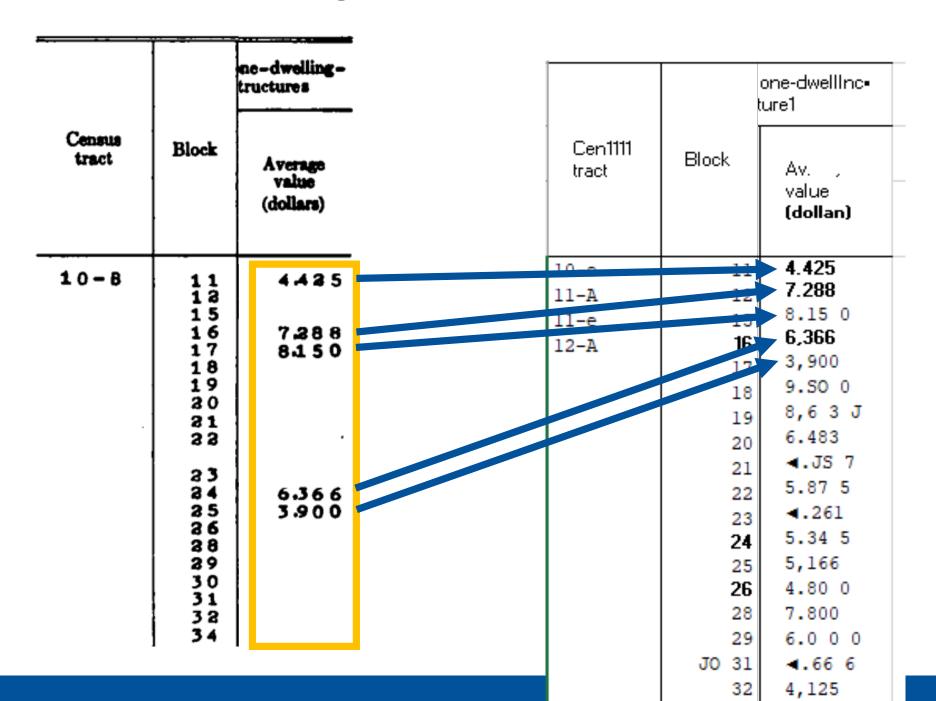
		one-dwellInc• ture1
Cen1111 tract	Block	Av. value (dollan)
10-e	11	4.425
11-A	12	7.288
11-e	15	8.15 0
12-A	16	6,366
	17	3,900
	18	9.SO 0
	19	8,63J
	20	6.483
	21	∢.JS 7
	22	5.87 5
	23	∢. 261
	24	5.34 5
	25	5,166
	26	4.80 0
	28	7.800
	29	6.0 0 0
	JO 31	∢. 66 6
	32	4,125

Adobe:

		ne-dwelling- tructures
Census tract	Block	Average value (dollars)
10-8	11 12 15 16 17 18 19 20 21	4.425 7.288 8150
	23 24 25 26 28 29 30 31 32	6.366 3.900

		one-dwellInc• ture1
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10-e	11	4.425
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	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:

Census tract Block Total Owner Renter Renter not	Occupied				
Total Wild Relief not and Number private water Total		mind	Average		Averes
occupied occupied dilap., for rent or sale or sale occupied dilap.	white	Number reporting	monthly rent (dollars)	Number reporting	Average value (dollars)
20 75 42 19 4 10 56 3 2 61 56 67 42 42 42 42 42 42 43 42 47 44	3 4 4 2 1 1 1 2 1 2 3	41 34 55 30 16 7 6 10 97 24 43 33	4 4.3 4 4 3.2 0 4 8.8 3 5 0.9 3 3 8.2 8 4 4.4 3 4 4.8 5 4 2.0 0 5 3.1 0 4 0.2 2 3 9.2 2 4 1.2 9 4 3.8 8 4 6.1 2 4 7.3 3	23 33 15 30 30 30 30 30 30 30 30 30 30 30 30 30	8.050 10.050 10.064 10.040 10.

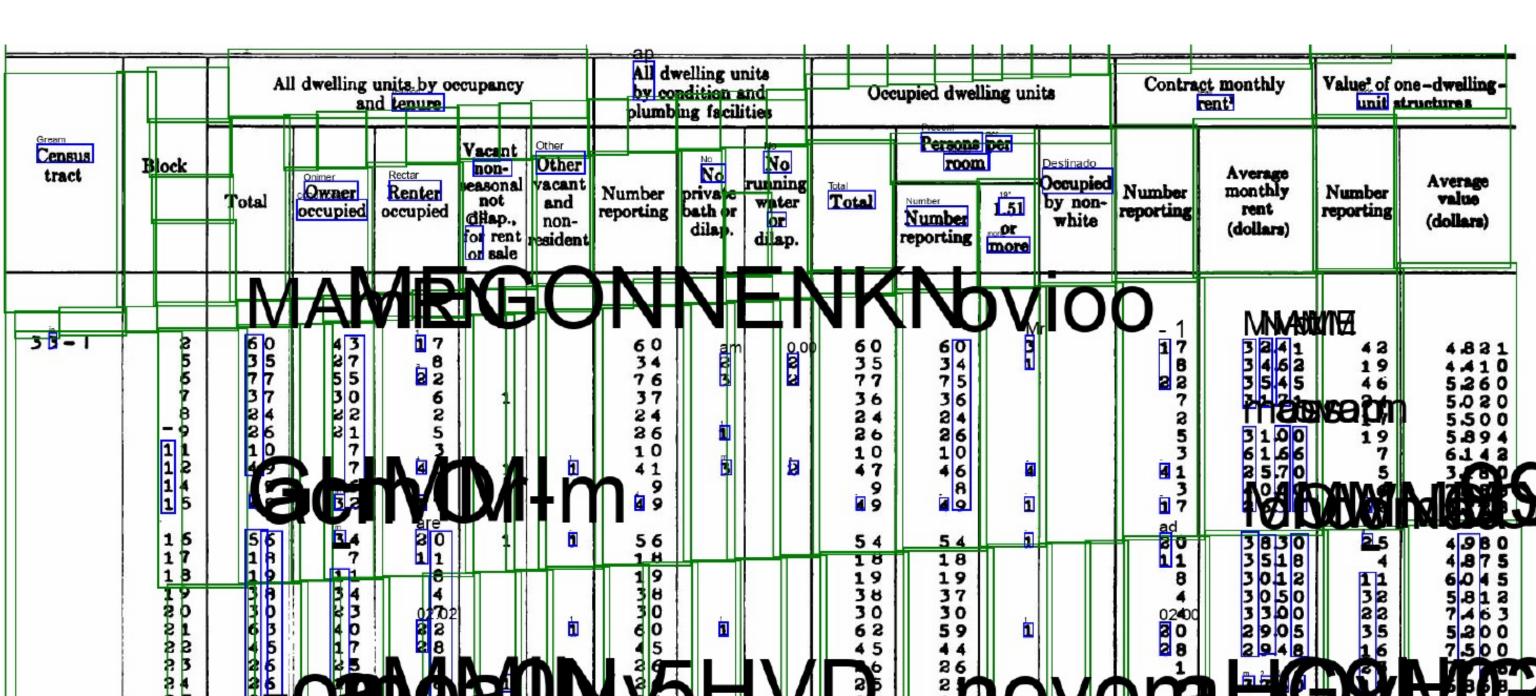
- Textract: when it works it works!
- 1.5% error rate, 0.22% ignoring cell alignment errors (stats for this page only)

1.7 1.90 36 5.7 5 2 94 1 1 93 93				,		- 5		AT DINE	Table Trains	<u> </u>						1 3	,	
Coupied Coup			Ä	dwelling an	units by oc d tenure	cupancy		by co	velling u ndition s ing facili	pite und ties	Occ	upied dwelling	ling uni	ita			Value of c	ne-dwelling- tructures
1.0	Census	Block	Total		Renter occupied	pot dilap.	vacant and non-	Number reporting	private	No running water or dilap.	Total	Number	1.51	Occupied	Number reporting	monthly	Number	Ayerage
28 77 32 64 1 76 1 1 76 77 76 77 77 76 77 77 78 76 77 77 78 78 77 78 <th>40-G</th> <th>1 7 1 8 1 9 2 0 2 1 2 2 2 3 2 4 2 5 2 6 2 7 2 8 2 9 3 0 3 1</th> <th>1.00 7.5 7.9 7.5 6.6 4.2 5.0 6.4 7.7 7.9 5.8 4.9</th> <th>36 43 66 42 59 53 47 47 47 39 39 42 42</th> <th>1 0 1 0 1 0 1 0 1 0 1 0 1 0 1 0 1 0 2 5 4 4 3 6 1 0</th> <th></th> <th></th> <th>94 66 72 56 62 42 62 54 98 54 98 54</th> <th></th> <th></th> <th>9 3 7 3 7 7 6 1 6 6 4 2 4 7 6 3 5 7 1 0 5 6 4 7 6 7 6 7 6 7 6 7 6 7 6 7 6</th> <th>9 3 7 3 7 7 5 6 6 6 4 2 4 4 6 3 5 5 9 7 6 4 7 6 9 7 4 7</th> <th></th> <th></th> <th>5 5 3 0 7 1 6 7 6 1 0 5 7 2 4 4 3 9</th> <th>4.8.8 3 5.0.9 3 3.8.2 6 4.4.4 3 4.4.8 5 4.2.0 0 5.3.1 0 4.0.2 2 3.9.2 2 4.1.2 9 4.3.8 6 4.6.8 3</th> <th>1 5 31 40 38 48 30 24 23 30 30 30 26 41</th> <th>8.630 1.6050 1.0866 1.0.064 7.040 6.921 7.479 8.300 8.102 7.434 9.575 9.600 6.543 2.570 7.869 8.323 8.402 8.128</th>	40-G	1 7 1 8 1 9 2 0 2 1 2 2 2 3 2 4 2 5 2 6 2 7 2 8 2 9 3 0 3 1	1.00 7.5 7.9 7.5 6.6 4.2 5.0 6.4 7.7 7.9 5.8 4.9	36 43 66 42 59 53 47 47 47 39 39 42 42	1 0 1 0 1 0 1 0 1 0 1 0 1 0 1 0 1 0 2 5 4 4 3 6 1 0			94 66 72 56 62 42 62 54 98 54 98 54			9 3 7 3 7 7 6 1 6 6 4 2 4 7 6 3 5 7 1 0 5 6 4 7 6 7 6 7 6 7 6 7 6 7 6 7 6	9 3 7 3 7 7 5 6 6 6 4 2 4 4 6 3 5 5 9 7 6 4 7 6 9 7 4 7			5 5 3 0 7 1 6 7 6 1 0 5 7 2 4 4 3 9	4.8.8 3 5.0.9 3 3.8.2 6 4.4.4 3 4.4.8 5 4.2.0 0 5.3.1 0 4.0.2 2 3.9.2 2 4.1.2 9 4.3.8 6 4.6.8 3	1 5 31 40 38 48 30 24 23 30 30 30 26 41	8.630 1.6050 1.0866 1.0.064 7.040 6.921 7.479 8.300 8.102 7.434 9.575 9.600 6.543 2.570 7.869 8.323 8.402 8.128

Textract: when it doesn't work...

		A	ll dwelling an	units by oc	cupancy		by co	welling undition a ing facil	and	Occ	supied dwe	lling un	its		ct monthly rent ¹		ne-dwelling- tructures
Census tract	Block	- · ·			Vacant non-	Other		No	No		Persona room	n	Occupied		Average		Average
		Total	Owner occupied	Renter occupied	seasonal not dilap., for rent or sale	and non-	Number reporting	private bath or dilap.	running water or dilap.	Total	Number reporting	1.51 or more	by non- white	Number reporting	monthly rent (dollars)	Number reporting	value (dollars)
33-1	256789112115 112115	60 35 77 37 26 10 49 49	437 550 221 77 62 34	17 8 22 6 2 5 3 4 0 17	1	1	60 34 76 37 24 26 10 41 9	1 3	20 20 20 20 20 20 20 20 20 20 20 20 20 2	60 35 77 36 26 10 49 49	60 34 75 36 26 10 48 49 54	3 1 4 1	,	17 82 7 25 3 41 37 20	32.41 34.62 35.45 31.71 31.00 61.66 25.70 40.00 26.35	429 469 464 179 197 563 25	4.821 4.410 5.260 5.020 5.500 5.894 6142 3180 7.483 3.728
	17 18 19 20 21 22 23	18 19 38 30 63 45 26	7 11 34 23 40 17 25 19	1 1 8 4 7 2 2 8 1 6	1	1	1 8 1 9 3 8 3 0 6 0 4 5 2 6	1		18 19 30 63 45 25	18 19 37 30 59 44 26 25	1		11 8 4 4 20 28 1	3 5 1 8 3 0 1 2 3 0 5 0 3 3 0 0 2 9 0 5 2 9 4 8	4 11 32 23 35 16 23	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	mer 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	I.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 44	30 87 124 **120 247 145 85 21 14 51 18 10 9 54 60 133	984455400743 44285221 15221	8 22 34 26 38 25 12 2 12 13 13 13 13 13 13 13 13 14 15 15 15 15 15 15 15 15 15 15 15 15 15	8 22 30 24 38 25 12 2 12	42	150775 10775 10775	1149975	1 1	5	2551	6 14 22 20 41 28 16 1 9	3500 6000 6000 4500 5000 5000 4000 4000 5500 55	5.3 7.0 7.5 7.5 7.0 7.5 7.0 7.0 8.4 7.4	3 12 16 12 11 8 4 4 2	43 28 40 47 39 	4.2 5.3 4.4 5.6 5.9		1 2 5 5 2 2 2 2 3 2 3
45	6 3	1			:::	***	:::	***	:::	***	***	:::	***	***	:::	:::	•••	***
49	141	43	17	16	1	26	20	3	3	•••	15	5500	8.3	21	31	4.9	1	4
50 51	84 22	35 9	14	13 7		21 2	1	10	7	:::	12 5	3500	5•0 7•6	15 1	22	4.4	:::	٠.,
18-B	*8802 111	2781 53		2287 33	71 5	338 14	248 12	57 2	33	85 1	1823	5000	6.5	727	40	4.7	2	171

Pass through Textract

20

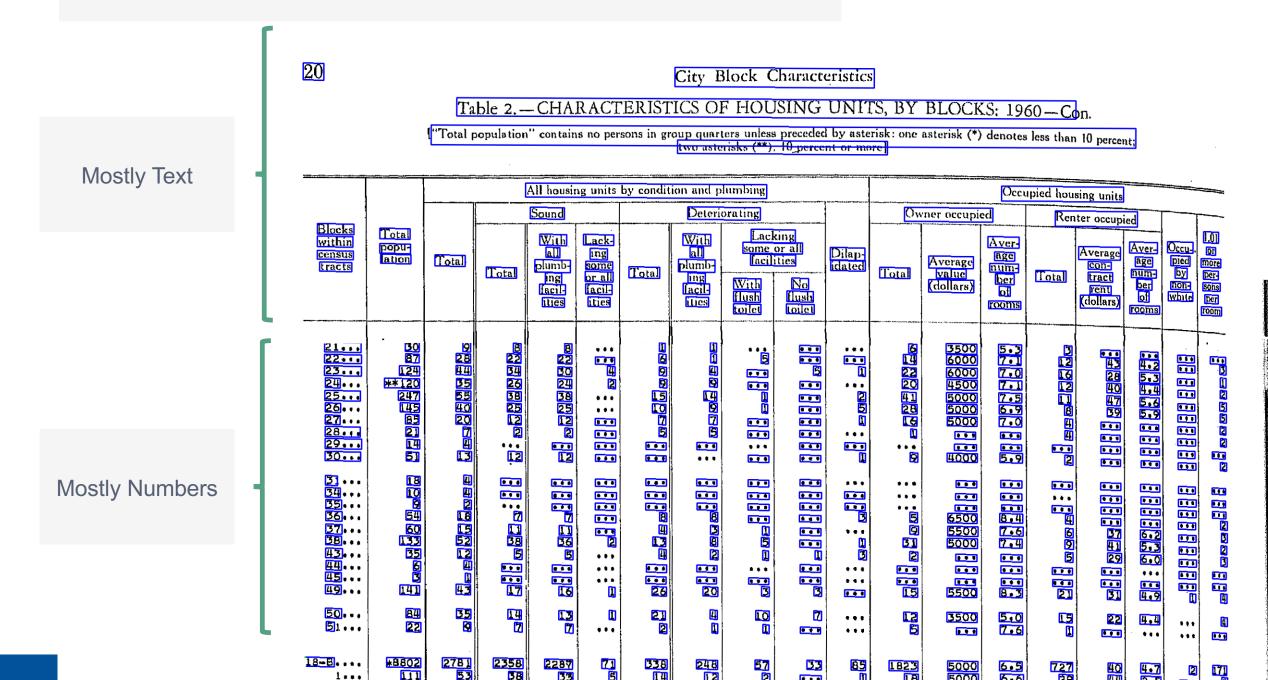
City Block Characteristics

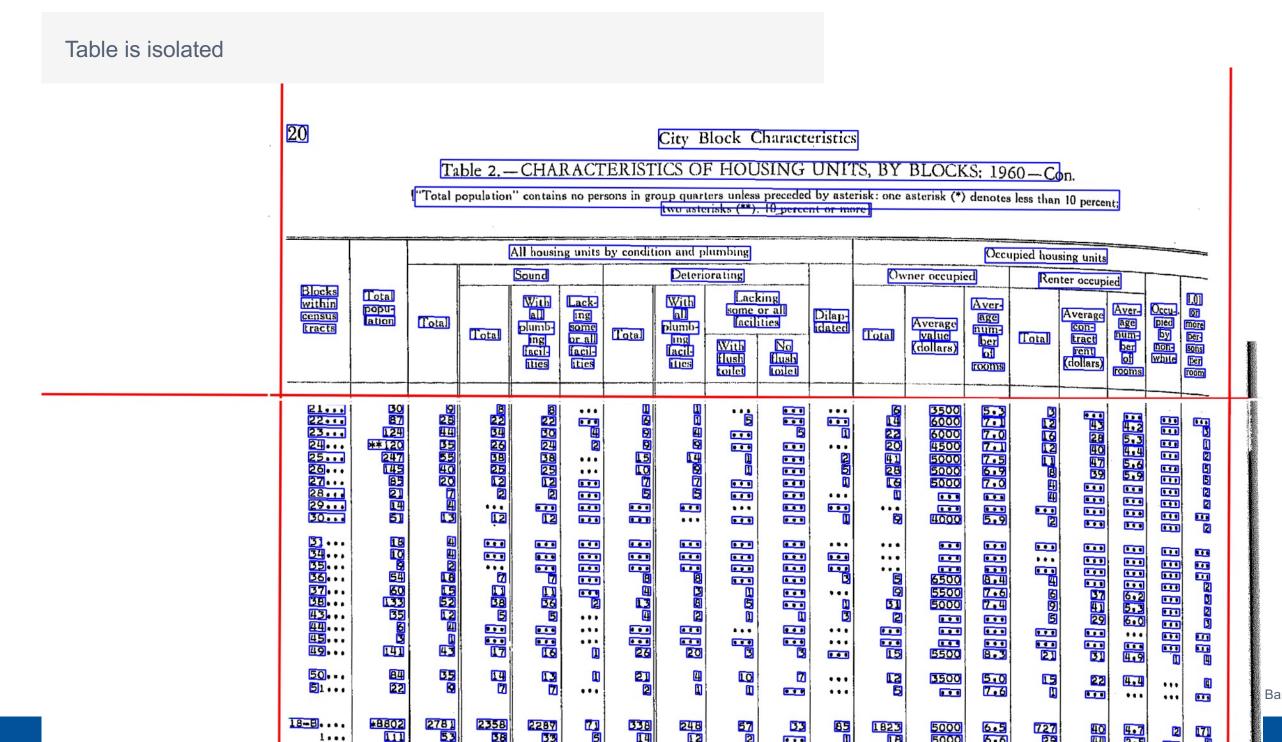
Table 2.—CHARACTERISTICS OF HOUSING UNITS, BY BLOCKS: 1960—Con.

"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

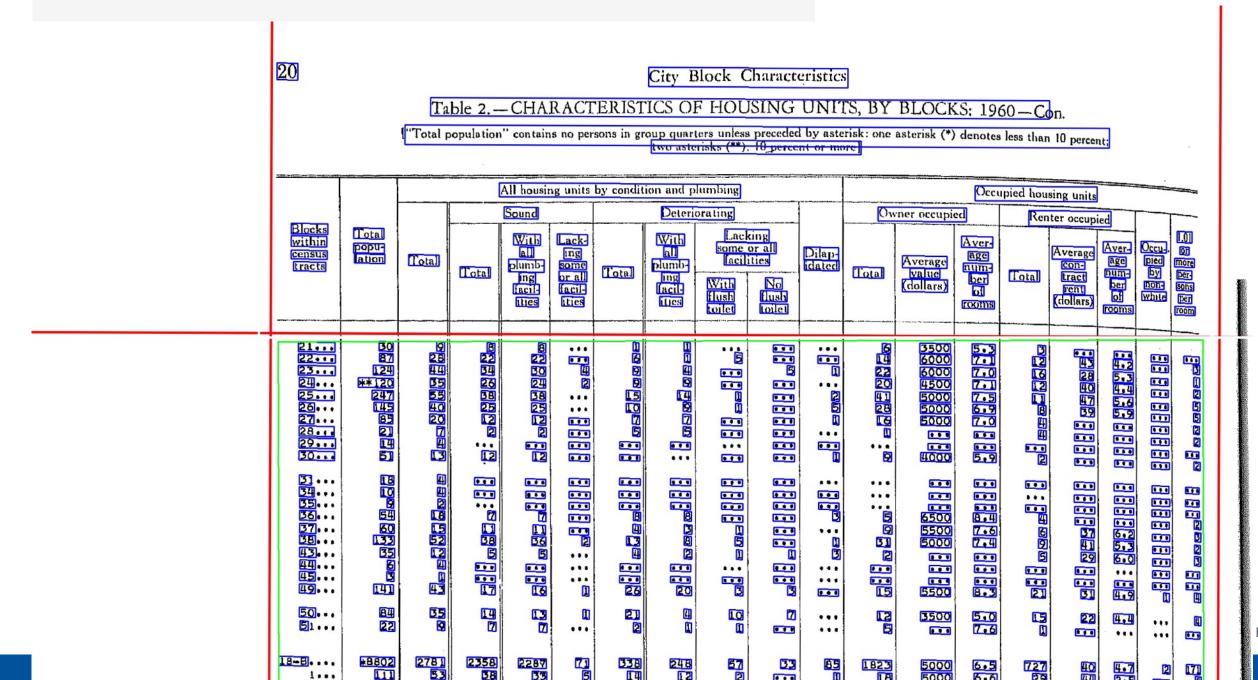
ļ			[All housir	g units	by condit	ion and p	lumbing					Occi	ipied hou	sing units			
				Sound			Deterio	orating			Ow	ner occupi	ed	Ren	ter occupi	ed		
Blocks within census tracts	Tota popu- ation	Total	Total	With all plumb-ing facil-ities	Lack- ing some or all facil- ities	Total	With all plumb- ing acil- ities	Lack some facil With flush toilet	or all	Dilap- idated	Total	Average value (dollars)	Aver- age num- ber of rooms	Total	Average Contract rent (dollars)	Aver- age num- ber of	Occu pied by non- white	OT More Per- Sons Per Toom
21 22 23 24 25 26 27 28 29	20 20 20 20 20 20 20 20 20 20 20 20 20 2		22 54 26 55 25 25 21 21									3500 6000 6000 4500 5000 5000 5000	5.3 7.1 7.0 7.1 7.5 6.9 7.0		49 49 49 49 49	5.6 5.9 5.9		
51 55 55 57 58 43 44 45		E ENGREE	7 7 28 5 17			26						6500 5500 5000 5000 5000	9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9 9			5.3 5.3 6.0		
50••• 51•••	<u>84</u> 22	55	[4]	13	•••	21 2	1	10	7	• • •	[<u>3</u>	3500	5 • 0 7 • 6	15	22	4.4	•••	4
18-B	*8802 111	278 <u>1</u> 53	23 <u>58</u> 58	2287 33	7 1	338 174	248	57	<u> </u>	8 <u>5</u>	18 <mark>23</mark>	50 <u>00</u>	6.5	727	40	4.7	[2]	171

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





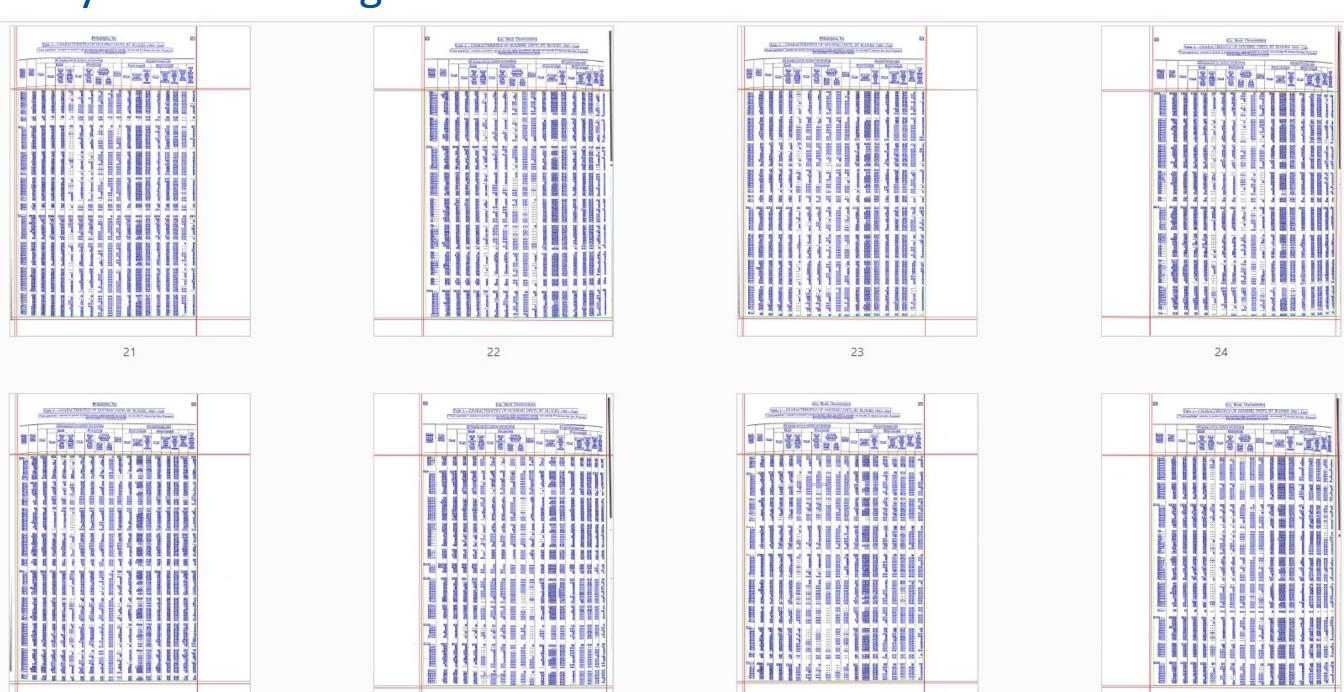
Find the rotated bounding box that contains all the body bounding boxes



Rotate image around center of table – image is straightened

20								ICS OI		SING preceded	UNIT	S, BY	BLOCK				ıt;		
					All housir Sound	ng units	by condit	ion and p Deterio	lumbing prating			Ow	ner occupie		ipied hour	sing units	ha		
wi CC	ocks ithin nsus acts	Total popu- ation	Total	Total	With all plumb- ing facil- ities	Lack- ing some or all facil- ities	Total	With all plumb- ling facil- ties	Some of facilians with flush toilet	or all	Dilap- idated	Total	Average value (dollars)	Aver- age num- ber of	Tota	Average Con- tract rent (dollars)		non- white	OI More Der- Sons Der Toom
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	23 24 25 25 26 27 28 13 14 15			7 28 5									6500 5500 5000 5000 5000		6 5 2	11 11 11 129 111 129	6.0 6.0 6.0	0 0 0 0 0 0	
	51	84 22 +8802	2781 53	14 [7] 2358 58	13 7 2287		21 2 338	248 248	10 11 57	[7] •••	<u>65</u>	1823 [8	5000 5000	5.0 7.6 6.5	727	22	4.7		

Always Be Checking

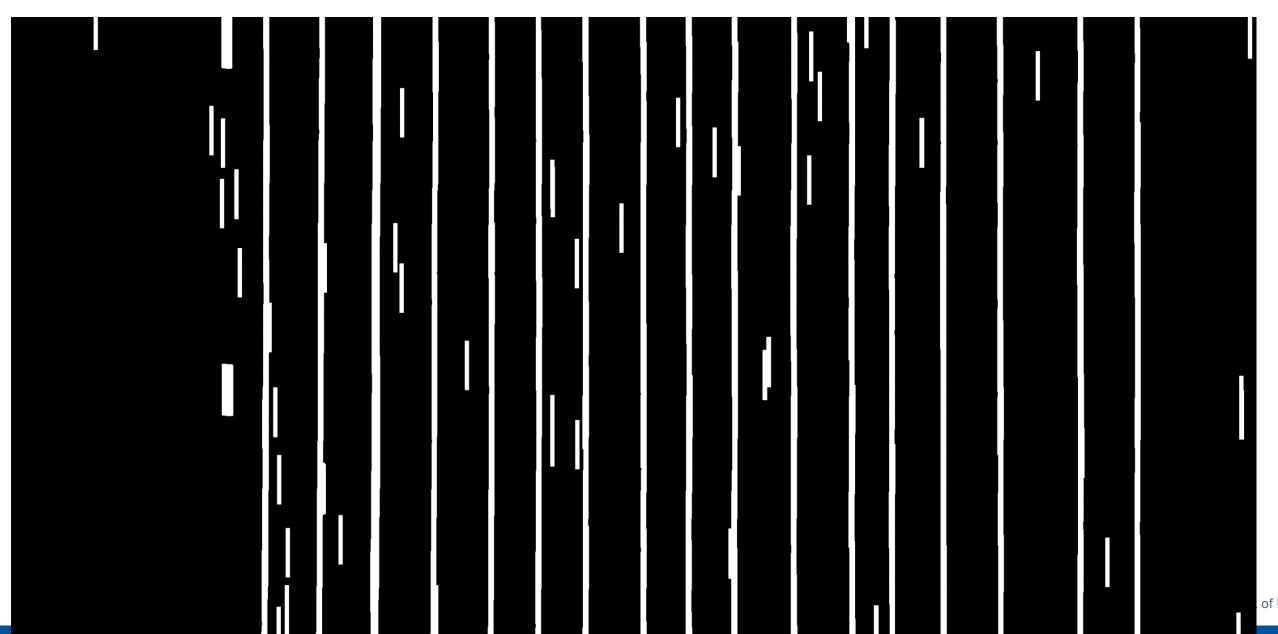


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21-U	23456789118	8 6 3 3 5 4 4 2 8 0 5 5 3 8	591755922 64332111	1512897121	3 1	1	303448053 86345881	S		003542043 003542043	80935 454 3420 143			1 6 8 8 8 8 7 2 1	4 2.6 2 5 3.2 5 4 5.8 7 5 0.3 7 4 9.7 1	6 1 5 4 5 1 3 2 2 3 9 6 2 2	8.69.6 9.41.1 10.84.5 10.04.6 10.27.2 12.83.3 3.37.7 7.150
	13 15 16 17 18	2 16 36 9 1	3 31 6	9 5 3	3	1	1.5 3.5 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 3 1 4	4 3 3 3 3 6 3 3 2 9 0 0 0
	20 22 23	9 43 34 6	28 21 3	3 3 15 13 3		. !	5 42 34 6	6		5 9 4 3 3 4 6	5 9 4 3 3 4 6			3 1 3 1 2 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 13000
	24 25 26 27	14 28 29 19	13	1 3 4		8	14 28 19			1 4 2 6 1 9	14 26 19	1.		1 3	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	1834567890 1	59 5505574 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1	9026730228 16 4232214	345 3339 1899 1699 1146	1 1 1	2	1352809574 59 5502574	10 35 8	3	59 55000505 1	59 55959965 5598885	1	1	34 34 10 12 61 12 14 5	47.55 48.08 70.00 32.50 43.41 44.90 52.75 57.78 34.60	15 43 40 25 18 15 10 47	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 222	1440244975530	126 138 198 758 411	14687622798	1 1 1	1.	14404497429	431 x x 51	2 1 1	1490 1497 1497 151	1 3 1 3 1 3 1 3 1 4 9 7 5 3 1 5 1 5 1 5 1 5 1 5 1 5 1 5 1 5 1 5	2		13 8 23 6 12 79 6	51.76 31.50 31.00 40.50 30.41 22.57 36.66 35.33	1 2 2 2 7 1 8 3 5 3 0 7 3 6 7	7.9329 8.6273 9.073001 7.6572 4.0220 6.020 5.50

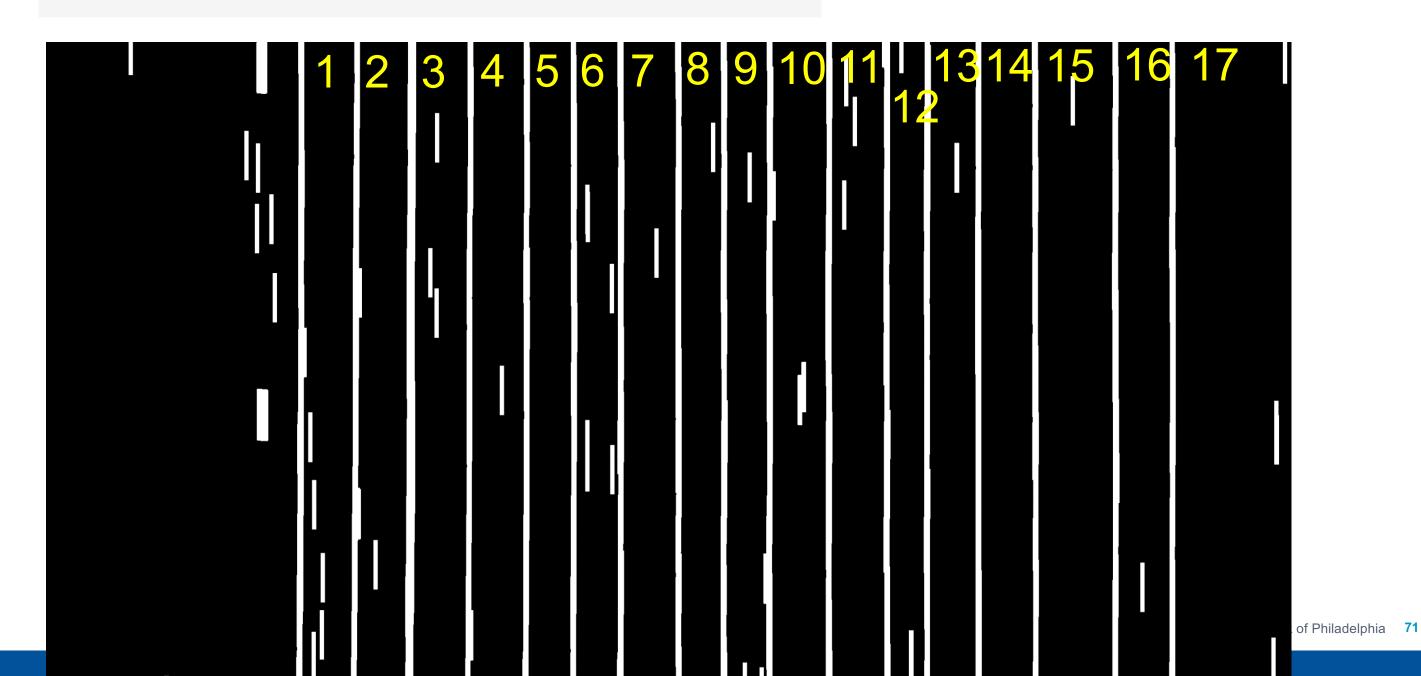
Find everything that *could* be a character. Be aggressive, recall is important

20-0		HWWANAOR MGGGWBGGRWG	PHENGGEDO		3 1	1		2			그 아무리 하시요 되었다.			4 2 5 2 5 2 2 5 4 5 8 7 5 0 3 7 4 9 7 1		8 9 6 8 4 4 4 5 8 4 4 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6
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					1 1 1	22	면 되었다. 전체도 없다면 보고 보고면 연극되었다.		3	ප අත්පත්තය අප පරිකම්කුම්ස්තියේ	 	60 60	1			728 6 500 0 500 0 500 5 500 5 500 5 500 5 500 5 600 5
			6년 교 이번 중요 장면 된 이번 중요 요		8	1			28	다 Separatration Temperatra	OCINENCENC POLICIENCENCE POLICIENCENCE	2				

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



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



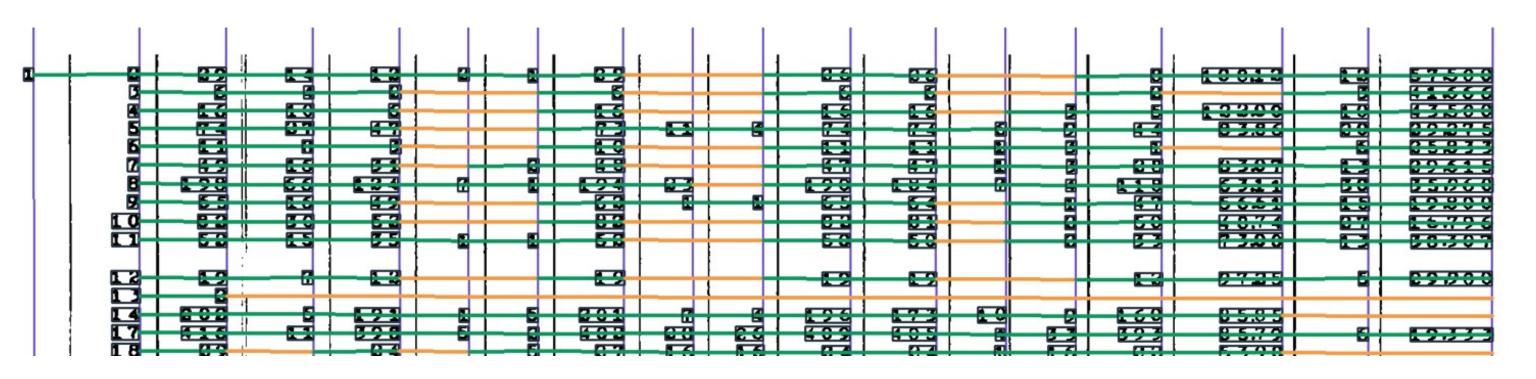
Columns are isolated

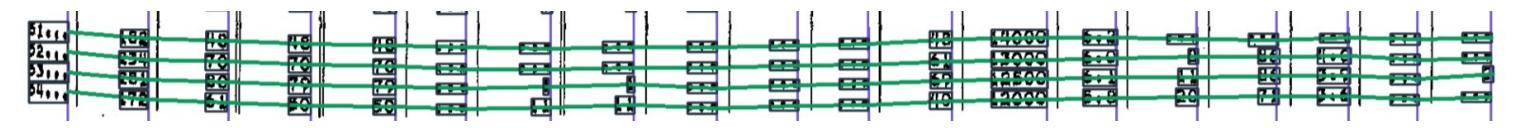
21-U	2345678912	1215420578 86747221	591755922 64332111	1512897121	3 1	1	303442053 63442053	8		86348043 8634881	093542045 55345221		·	1 682887 21	4 2.6 2 5 3.2 5 4 5.8 7 5 0.3 7 4 9.7 1	151883968 64338	8.696 9.411 10.845 10.272 12.833 3,377 7.150
	13 15 16 17 18	26 36 9 1	3 3 1 6	9 5 3	3	; 1	1 5 3 5 9	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 4 3 6	28 28 21 3	3 15 13 3			5 9 4 2 3 4 6	6		5 9 4 3 3 4 6	5 9 4 3 3 4 6			3 13 12 3	4 2.3 3 5 6.00 4 5.8 4 3 9.00 6 8.3 3	1 2 23 15	9,07 B 9,833 13,000
	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	59. 53005574 1 1 .	9026730226 16 4232214	3453 9899 1699 1146	1 1 1	2	1352809574 59 5802885	10 33 55 ខ	3	<i></i>	75555959965 59 5598885	1	1	33 1460245	47.55 48.08 70.00 32.50 43.41 44.89 52.75 537.78	14 42585907	7286 8081 5965 8480 10233 123315 7361
	1346789012 2112	1440244975530	123 1 32 41 1 32 41	14683688798	1 1 1	1.	1440 440 1849 7429 151	43 1 2 3 5	2 1 1	0908497589 4318148151	99902497529 3312142151	8		13823622796	51.76 31.50 31.00 40.50 30.41 22.57 36.66 35.33	1 2 7 7 8 5 5 0 7 6 7 3 7 6 7	7,929 8,673 9,073 7,607 6,573 4,973 8,073 6,53 6,53 6,53

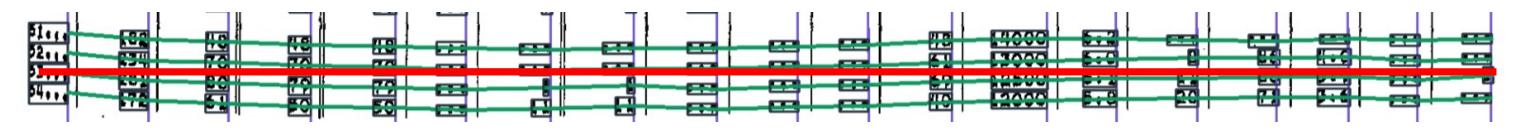
Tesseract Each Row in Each Column

- 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
TOTAL	TOTAL	
1-1	1-A	
1-8	1-B	
1 - C	1-C	
2 - Y	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!

