

# Differential Privacy for the 2020 Census

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Twin Cities Research Group Webinar  
September 16, 2020

# Protecting the Confidentiality of America's Statistics: Adopting Modern Disclosure Avoidance Methods at the Census Bureau

*Fri Aug 17 2018*

WRITTEN BY: DR. JOHN M. ABOWD, CHIEF SCIENTIST AND ASSOCIATE  
DIRECTOR FOR RESEARCH AND METHODOLOGY

# Protecting the Confidentiality of America's Statistics: Ensuring Confidentiality and Fitness-for-Use

*Tue Sep 04 2018*

WRITTEN BY: DR. JOHN M. ABOWD, CHIEF SCIENTIST AND ASSOCIATE  
DIRECTOR FOR RESEARCH AND METHODOLOGY

# Census Bureau Adopts Cutting Edge Privacy Protections for 2020 Census

*Fri Feb 15 2019*

WRITTEN BY: DR. RON JARMIN, DEPUTY DIRECTOR AND COO



# Census Bureau Continues to Boost Data Safeguards

*Tue Jul 30 2019*

WRITTEN BY: RON JARMIN, DEPUTY DIRECTOR, US CENSUS BUREAU

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# 2010 Demonstration Data Products

To help data users understand how differential privacy may or may not impact data products they are used to receiving, the Census Bureau created demonstration data products for review. This set of data products demonstrate the current computational capabilities of the 2020 Disclosure Avoidance System (DAS). The products include the 2010 Demonstration Public Law 94-171 (P.L. 94-171) Redistricting Data Summary File and the Demonstration Demographic and Housing Characteristics Summary File.

# DAS Development Update 2020-05-27



Detailed Summary Metrics [ $<1.0$  MB]



Data Metrics Overview [1.0 MB]



IPUMS NHGIS Privacy-Protected Census Demonstration Data



Download: 2010 Demonstration Privacy-Protected Microdata Files (PPMF)



# Outline

- How is differential privacy implemented?
- What was the impact on 2010 data?
- What are the next steps?

# HOW IS DIFFERENTIAL PRIVACY IMPLEMENTED?

# “True” microdata

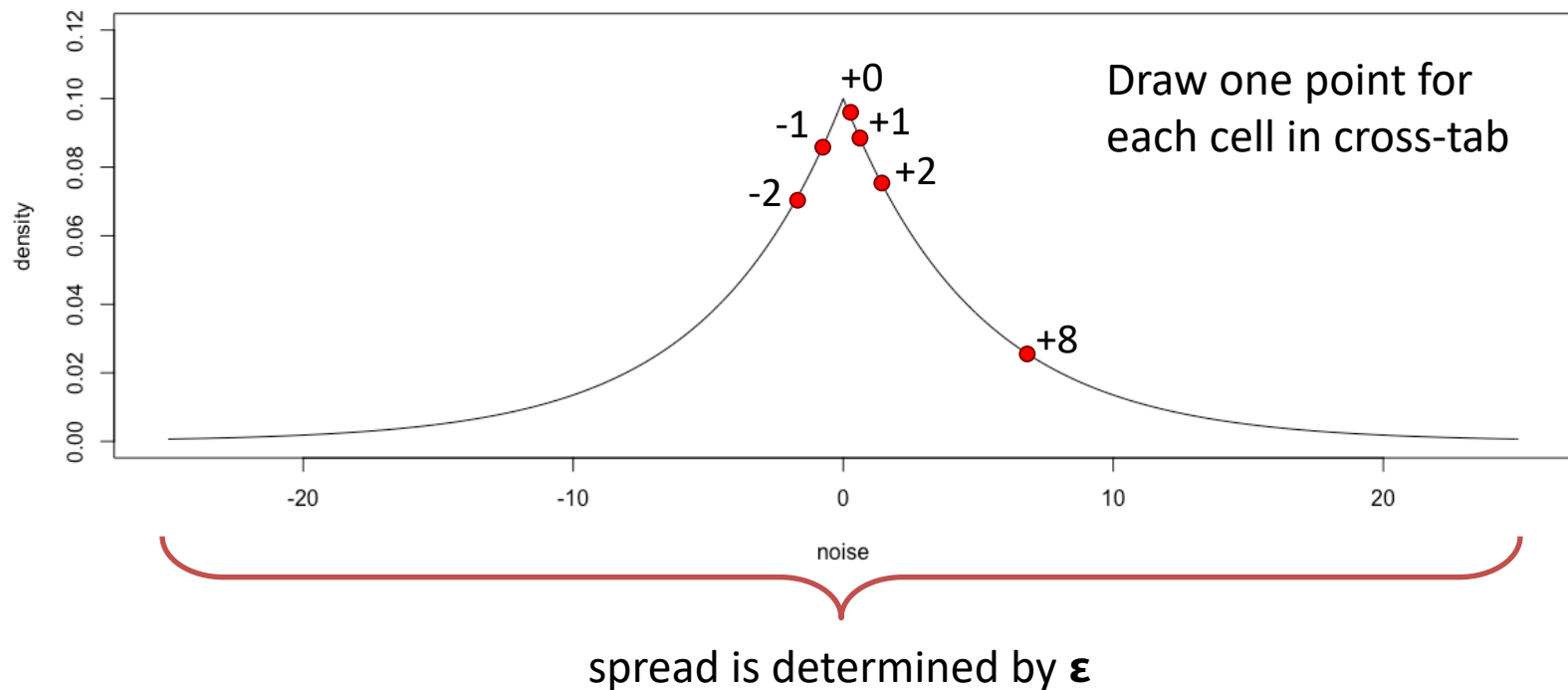
	<u>Sex</u>	<u>School</u>		<u>Sex</u>	<u>School</u>
	Male	Never		Female	Never
	Male	Never	x4 {	⋮	
	Male	Never		Female	Never
x12 {	Male	Attending		Female	Attending
	Male	Attending	x17 {	⋮	
	⋮			Female	Attending
	Male	Attending		Female	Past
x33 {	Male	Past	x31 {	⋮	
	⋮			Female	Past
	Male	Past			

# Construct cross-tabs from “true” data

	School Attendance		
	Never	Attending	Past
Male	3	12	33
Female	4	17	31

Population = 100

# Draw noise from Laplace distribution

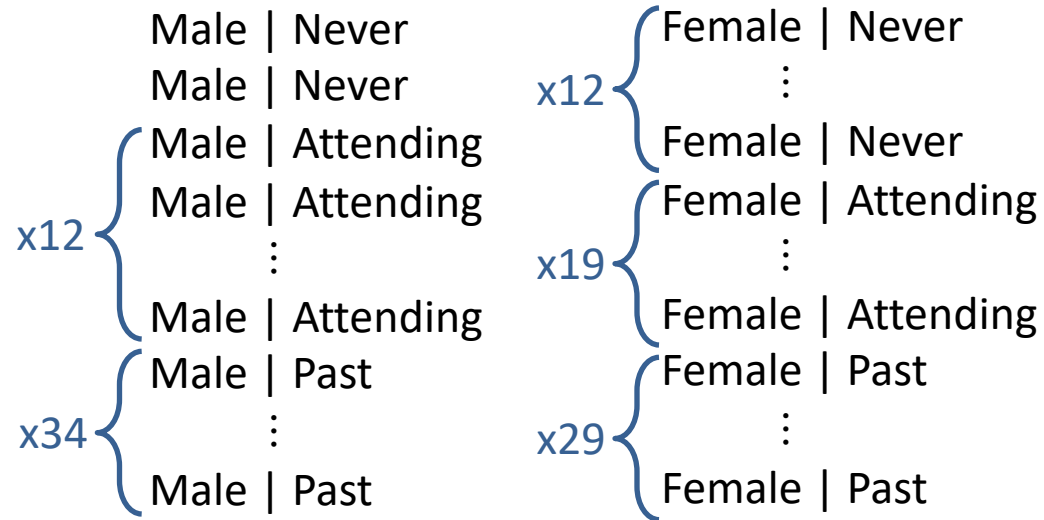


# Add noise to cross-tab

	School Attendance		
	Never	Attending	Past
Male	$3 - 1 = 2$	$12 + 0 = 12$	$33 + 1 = 34$
Female	$4 + 8 = 12$	$17 + 2 = 19$	$31 - 2 = 29$

Sum = 108

# Construct diff. private microdata



# POLICY DECISIONS



# Policy decisions

- Global privacy loss budget ( $\epsilon$ )
- Fractional allocations
- Invariants and constraints

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- Global privacy loss budget ( $\epsilon$ )
- Fractional allocations
- Invariants and constraints
- **Post-processing**

# Global privacy loss budget

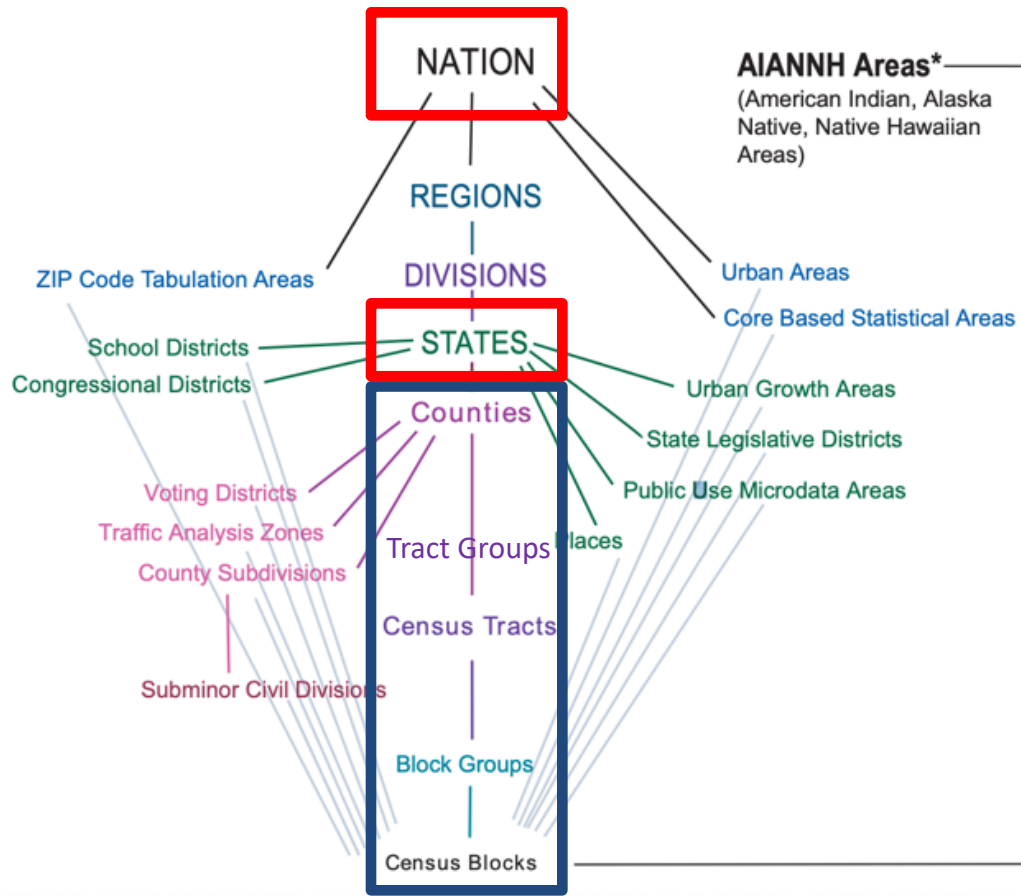
- Global privacy loss budget
  - $\epsilon = 6.0$
- Person tables
  - $\epsilon = 4.0$
- Housing tables
  - $\epsilon = 2.0$

# Fractional allocations

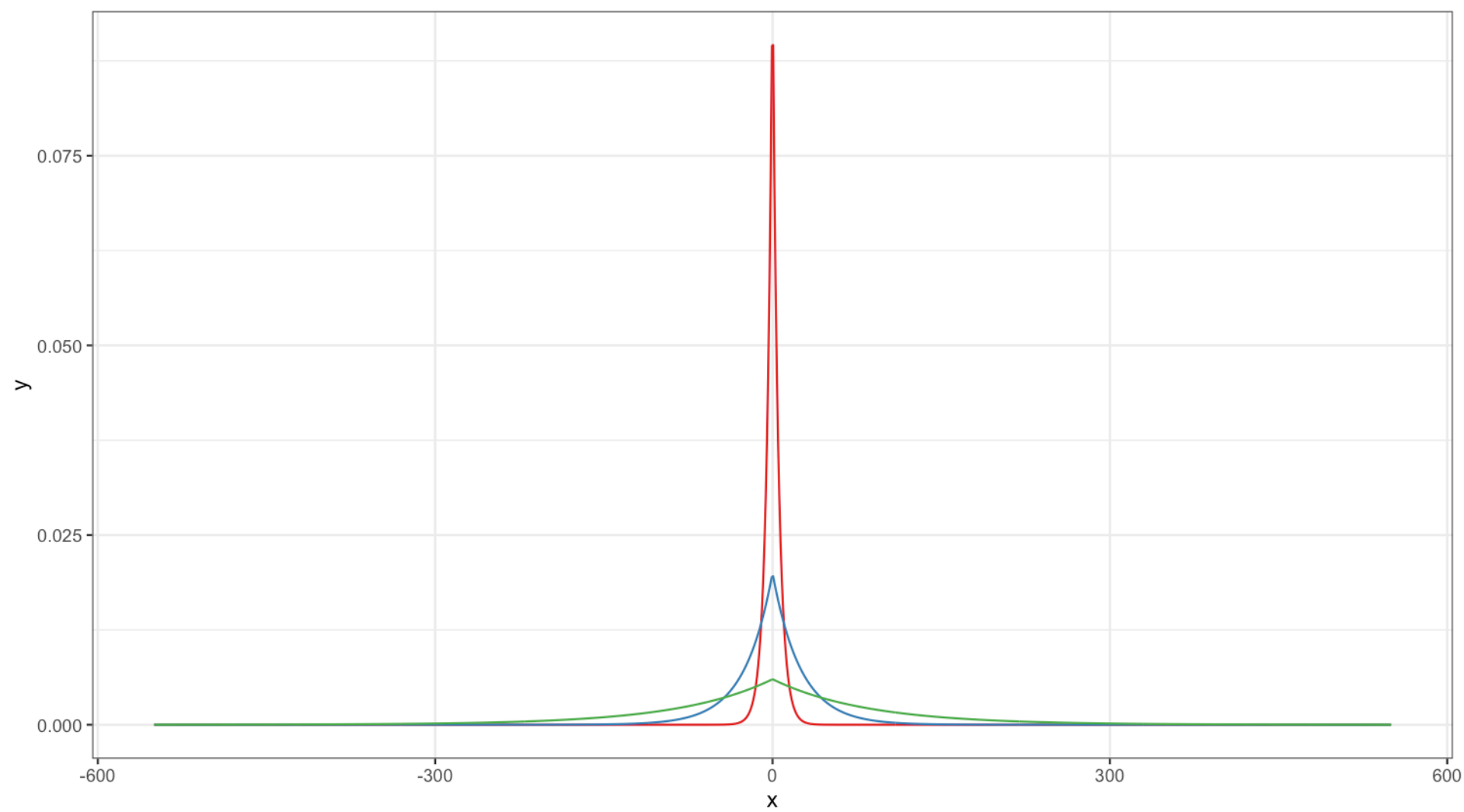
- Geographic levels
- Queries

20% each

12% each



Query	Allocation (%)
Voting age * Hispanic * Race * Citizen	50
Household – Group quarters	20
Detailed	10
Sex * Age (single year of age)	5
Sex * Age (4-year age bins)	5
Sex * Age (16-year age bins)	5
Sex * Age (64-year age bins)	5



# Invariants and Constraints

- Invariants are counts not subject to noise injection



---

## 2010 Decennial Invariants

---

Total population (block)

---

Total housing units (block)

---

Group quarters count (block)

---

Group quarters type count (block)

---

Occupancy status (block)

---

Voting age population (block)

---

---

## 2010 Demonstration Data Invariants

---

Total population (state)

---

Total housing units (block)

---

Group quarters count (block)

---

Group quarters type count (block)

---

---

## 2010 Decennial Invariants

---

Total population (**block**)

---

Total housing units (block)

---

Group quarters count (block)

---

Group quarters type count (block)

---

Occupancy status (block)

---

Voting age population (block)

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---

## 2010 Demonstration Data Invariants

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Total population (**state**)

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Total housing units (block)

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Group quarters count (block)

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Group quarters type count (block)

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# Invariants and Constraints

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- Invariants are counts not subject to noise injection
- Constraints
  - Non-negativity
  - Consistency

# Post-processing

- Non-negative least squares + constraints = positive bias for small counts and negative bias for large counts

# **ANALYZING DIFFERENTIALLY PRIVATE 2010 CENSUS DATA**

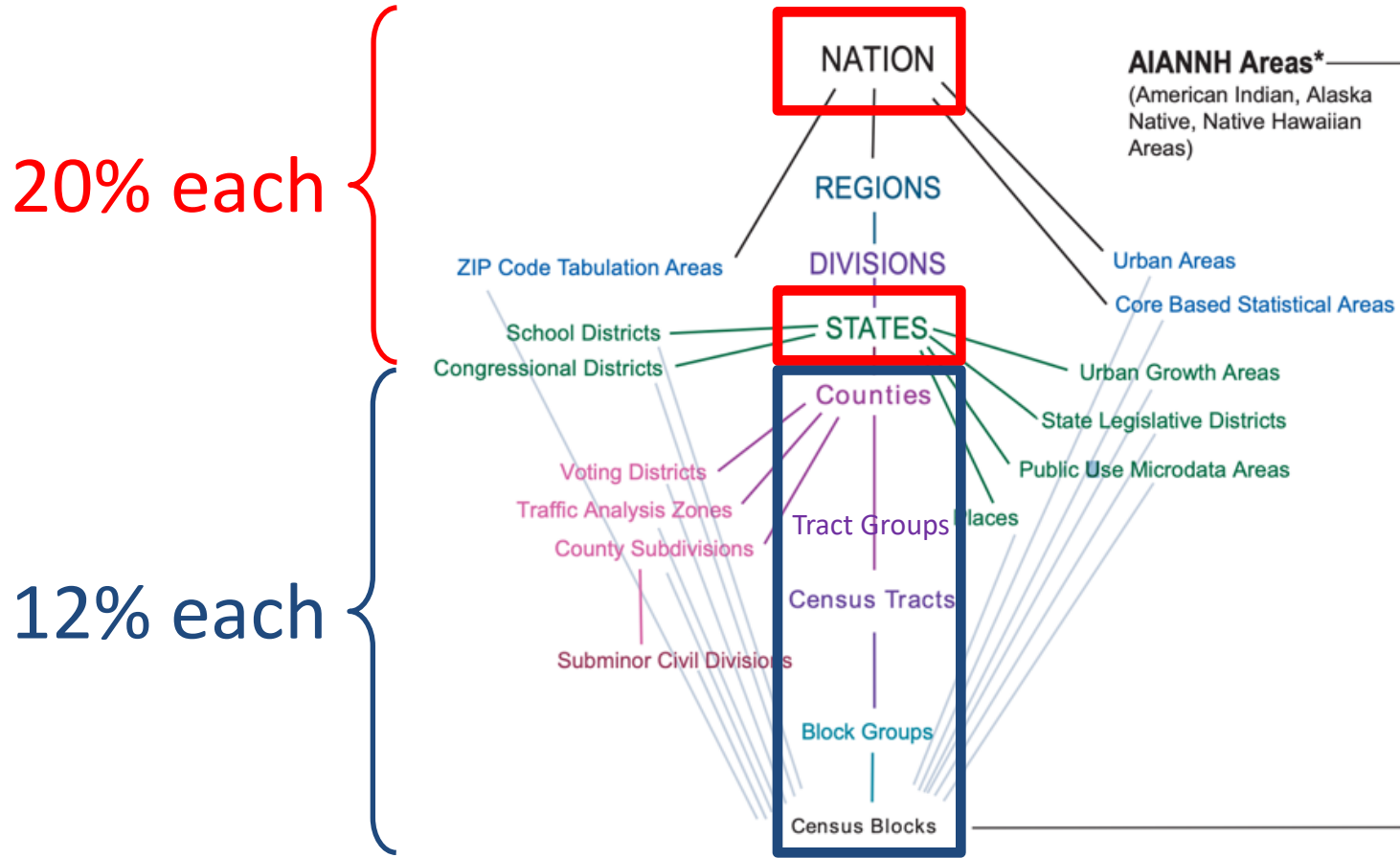
# Data

- 2010 Summary File 1
- Vintage 1 (October 2019)
- Vintage 2 (June 2020)

# Comparisons

- Comparing data from vintage 1 and 2 with data from Summary File 1
- Summary File 1 essentially serves as our “ground truth”
  - Acknowledging that prior disclosure avoidance techniques introduced error into SF1





## Vintage 1

Query	Allocation (%)
Voting age * Hispanic * Race * Citizen	50
Relation to HH/Group quarters	20
Detailed	10
Sex * Age (single year of age)	5
Sex * Age (4-year age bins)	5
Sex * Age (16-year age bins)	5
Sex * Age (64-year age bins)	5

## Vintage 2

Query	Allocation (%)
Total population	30
Voting age * Hispanic * Race	29
Age * Sex * Hispanic * Race	25
Relation to HH/Group quarters	15
Detailed	1

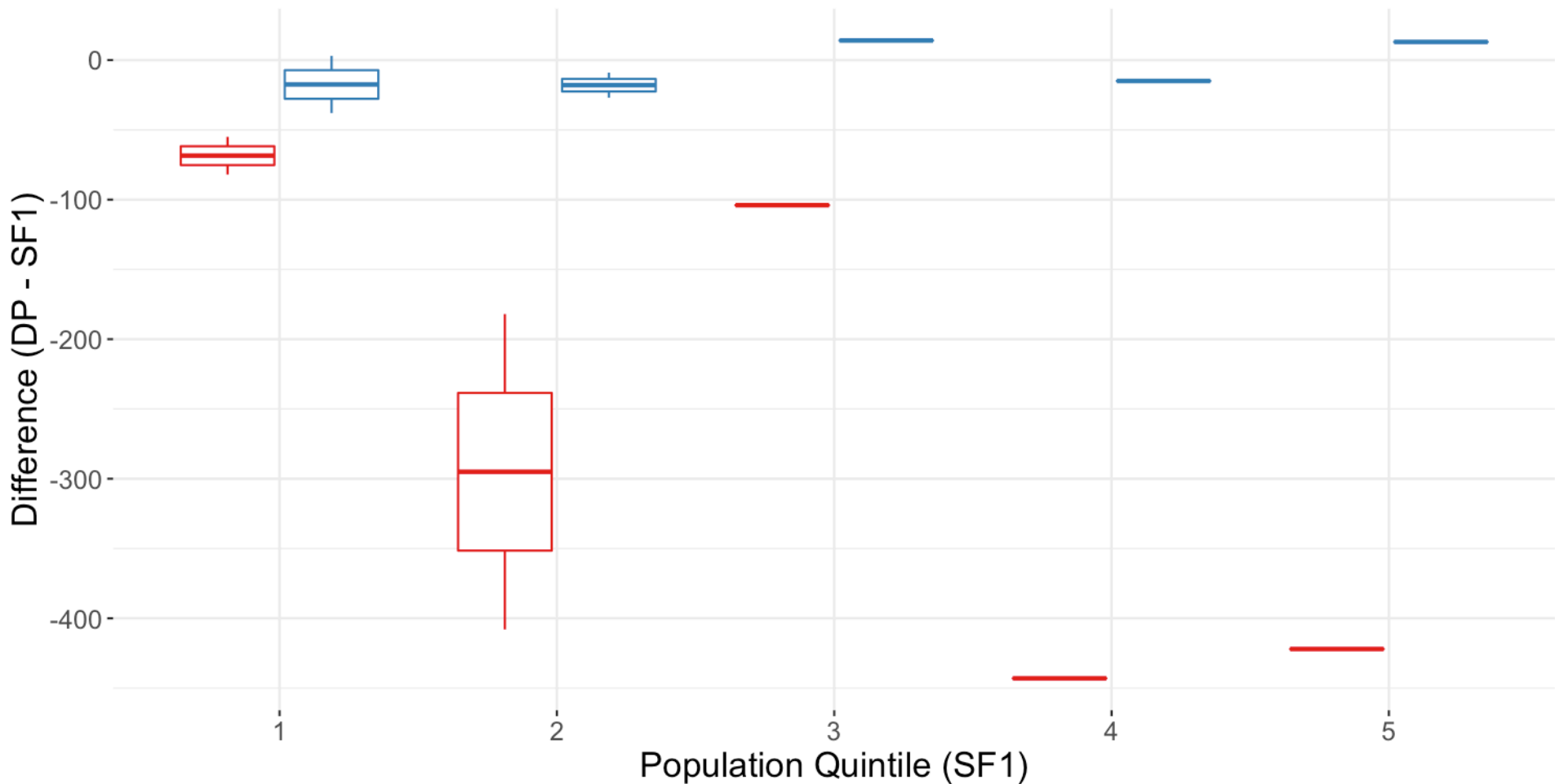
## Vintage 1

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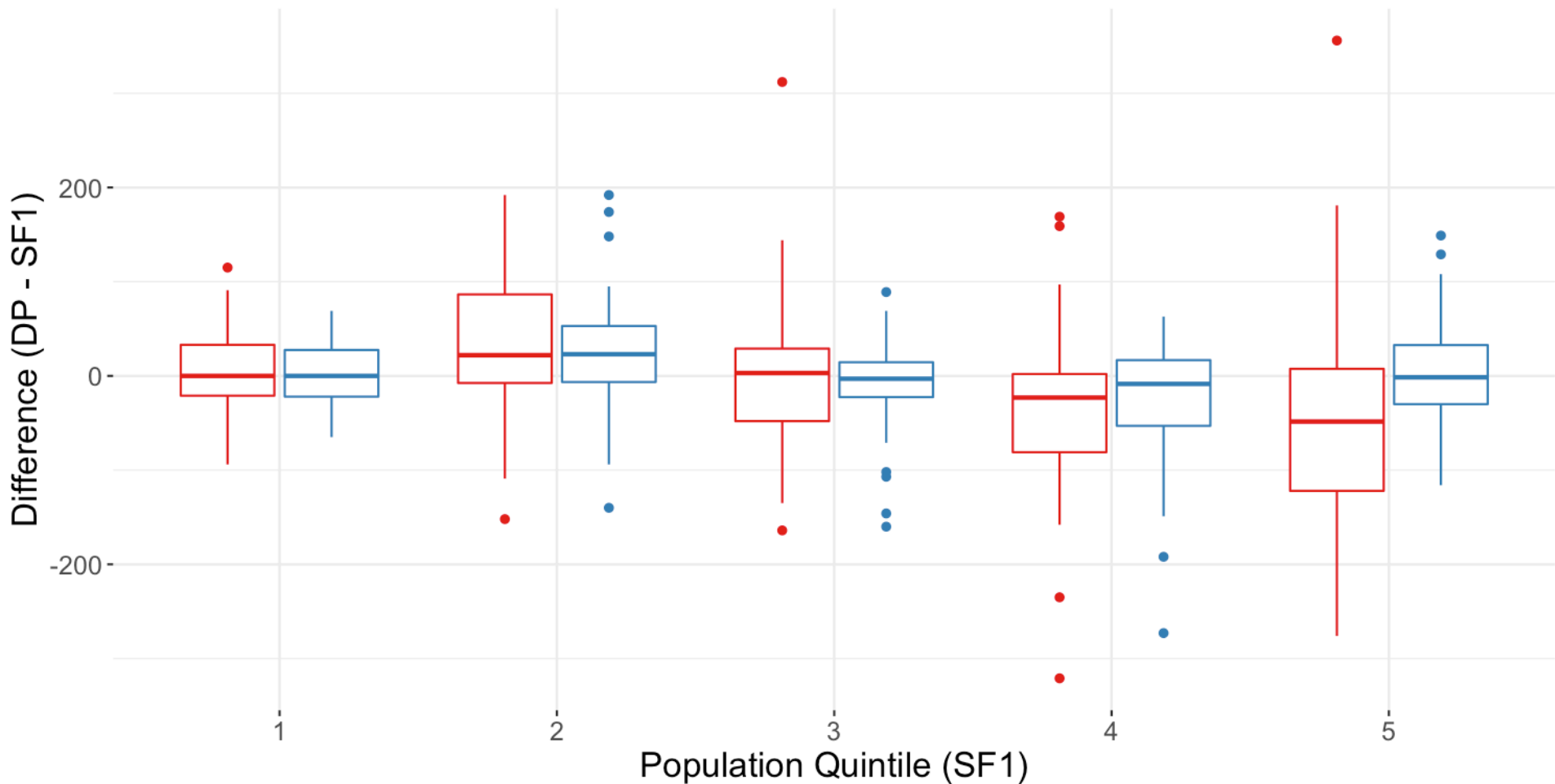
## Vintage 2

Query	Allocation (%)
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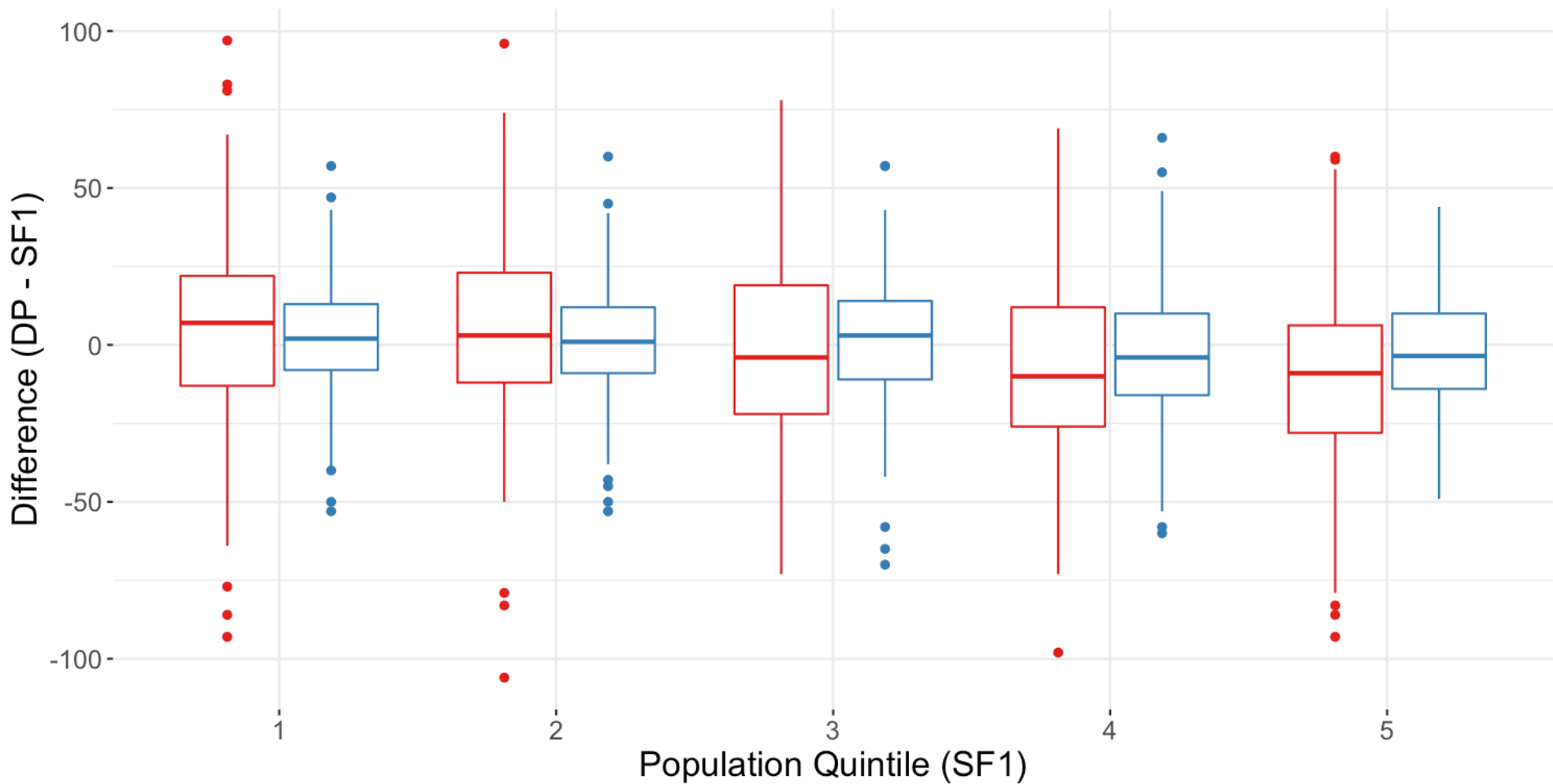
# 2010 SF1 vs. Diff. Private: Total Population for Counties



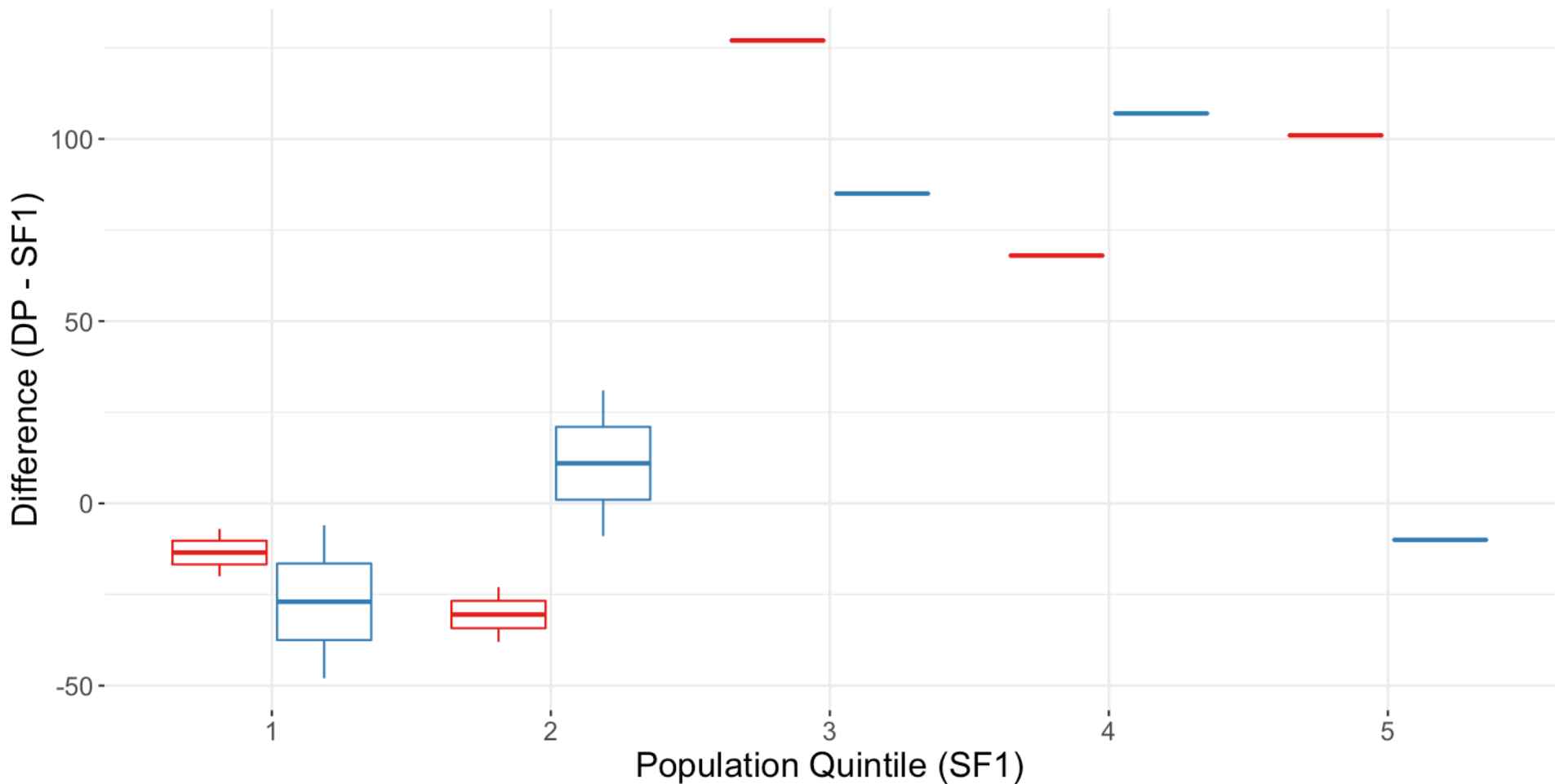
# 2010 SF1 vs. Diff. Private: Total Population for County Subdivisions



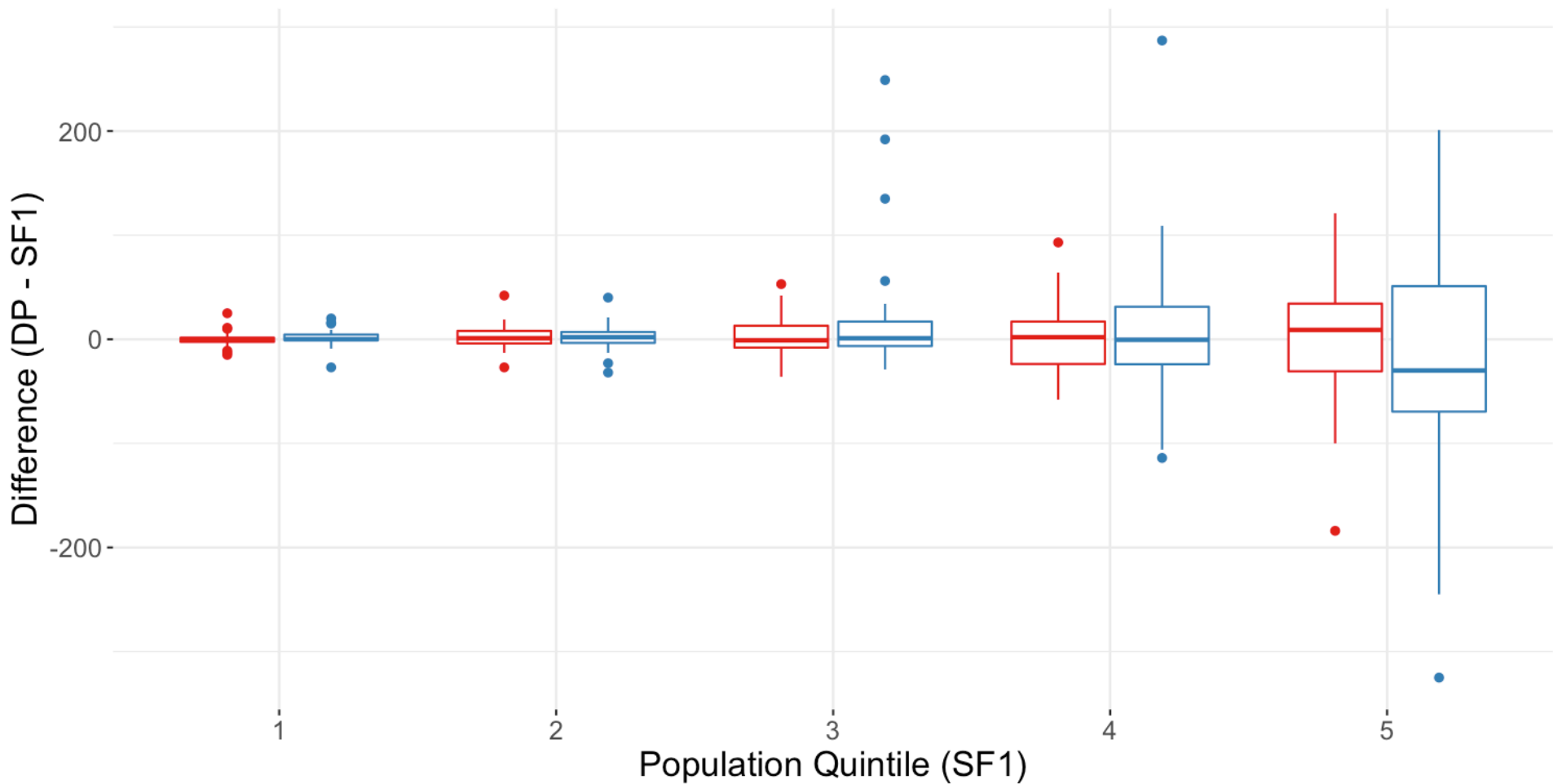
# 2010 SF1 vs. Diff. Private: Total Population for Tracts



# 2010 SF1 vs. Diff. Private: Black Alone Population for Counties

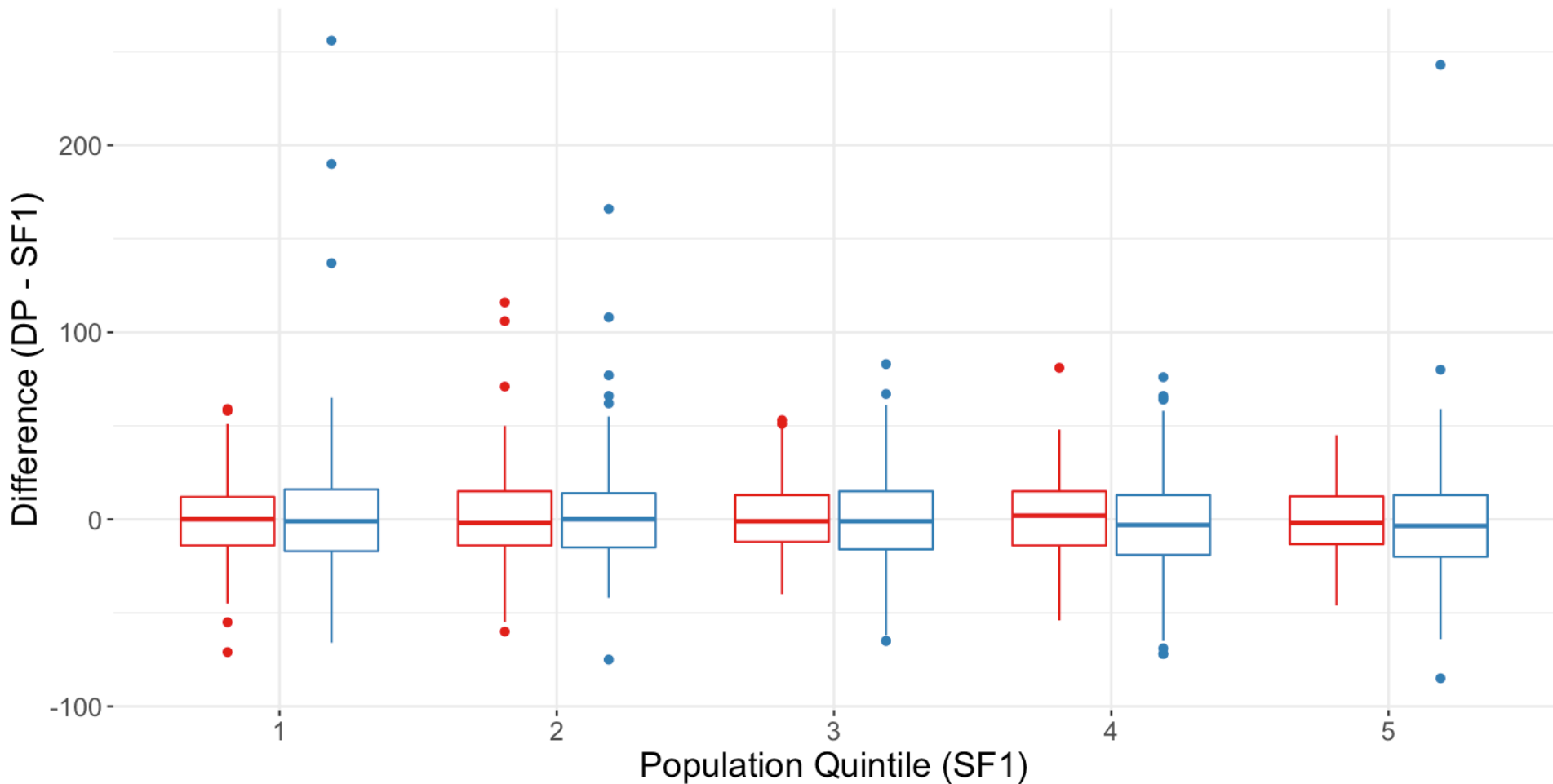


# 2010 SF1 vs. Diff. Private: Black Alone Population for County Subdivisions

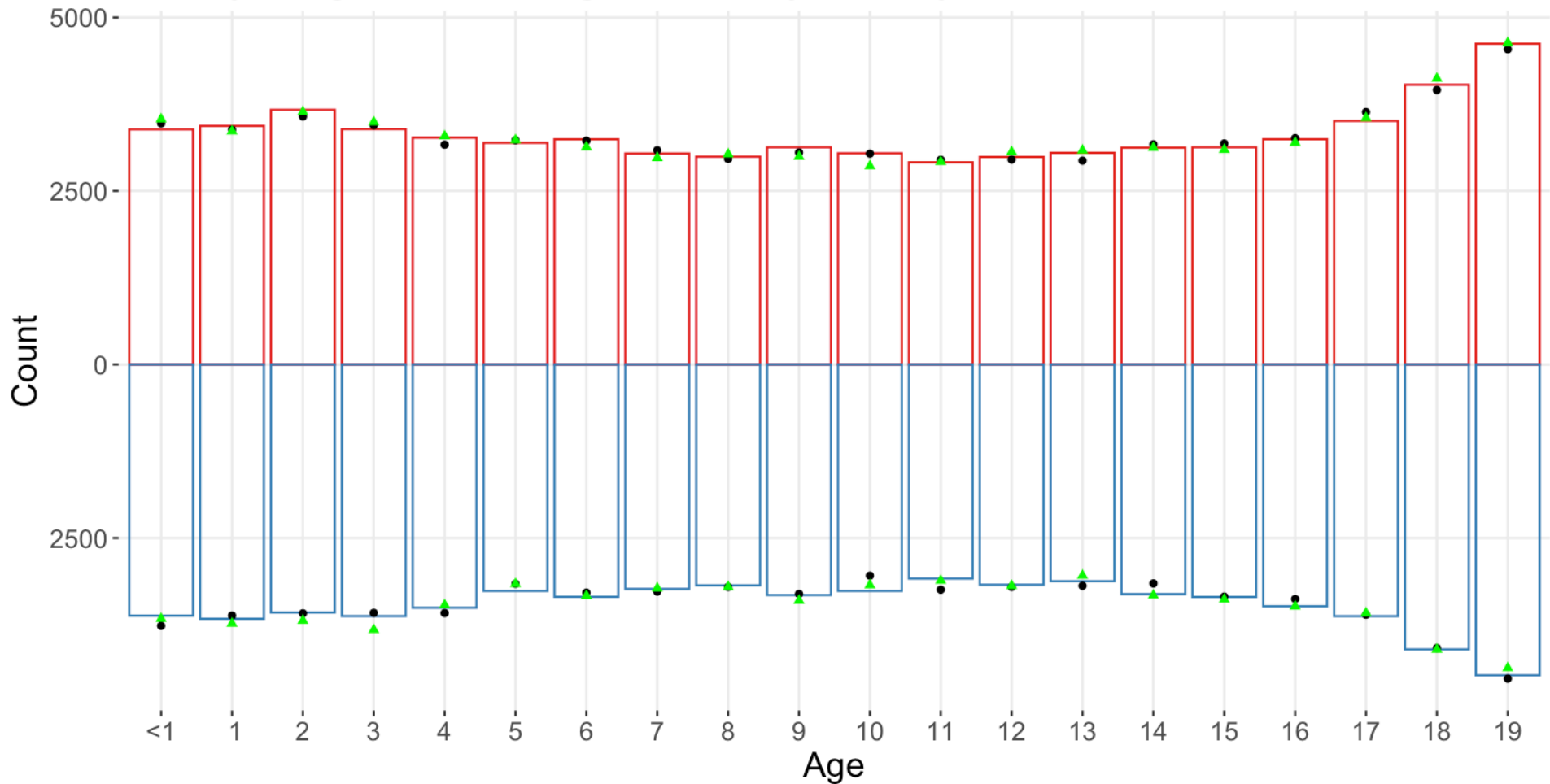




# 2010 SF1 vs. Diff. Private: Black Alone Population for Tracts

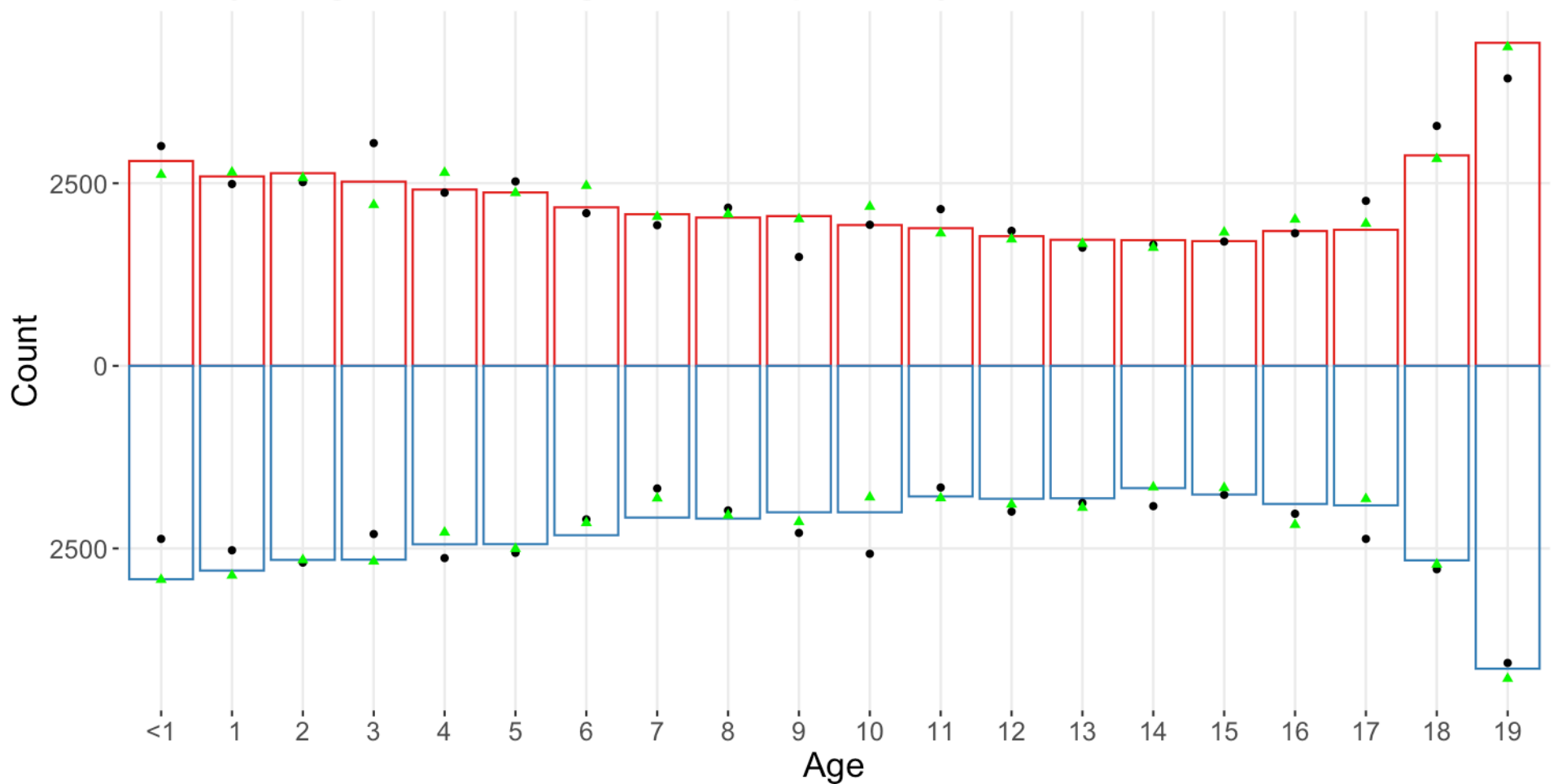


# Sex by Single Year of Age: Ramsey County



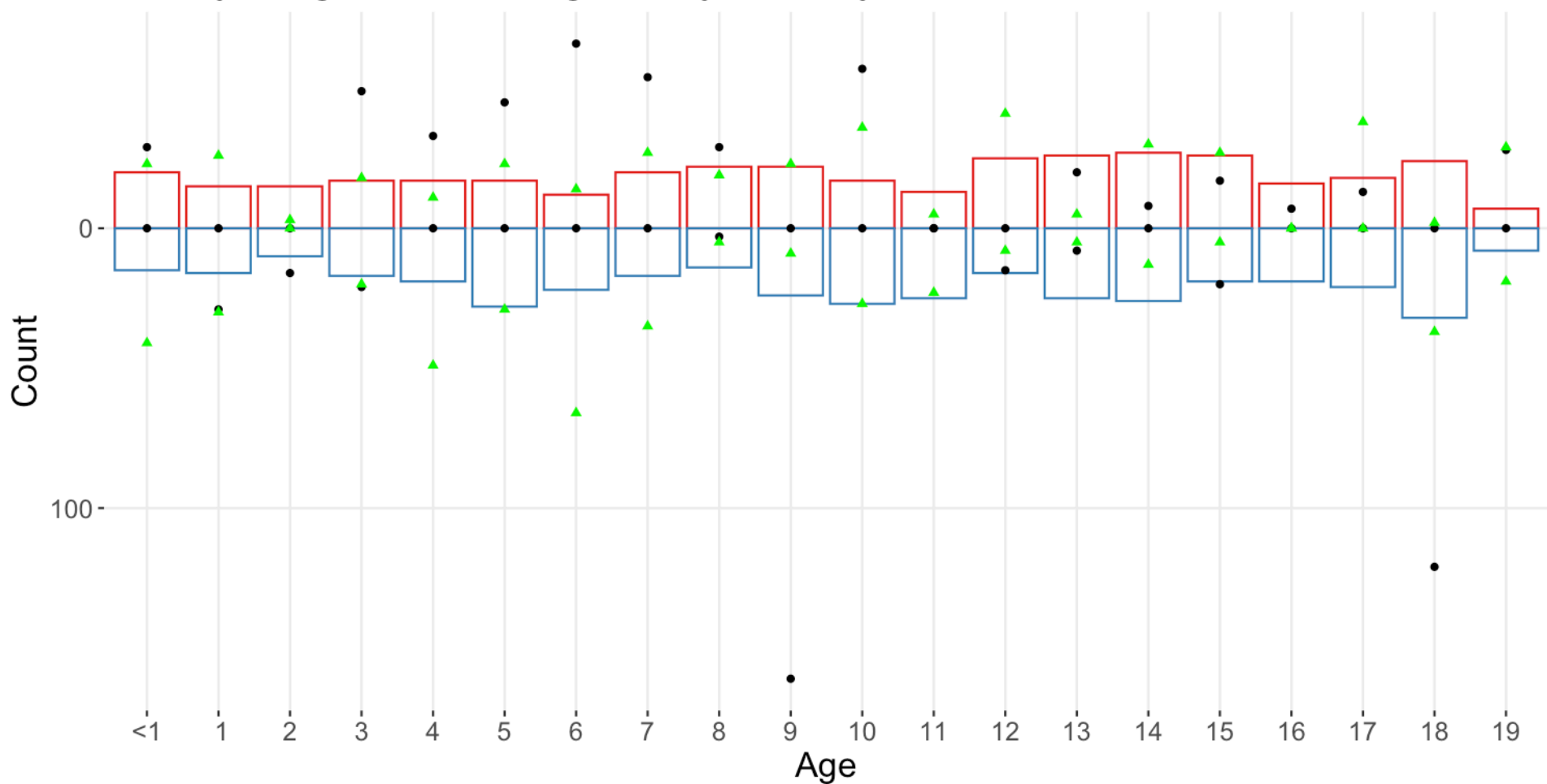
Source: US Census Bureau 2011; US Census Bureau 2019; Van Riper et al. 2020

# Sex by Single Year of Age: Minneapolis city

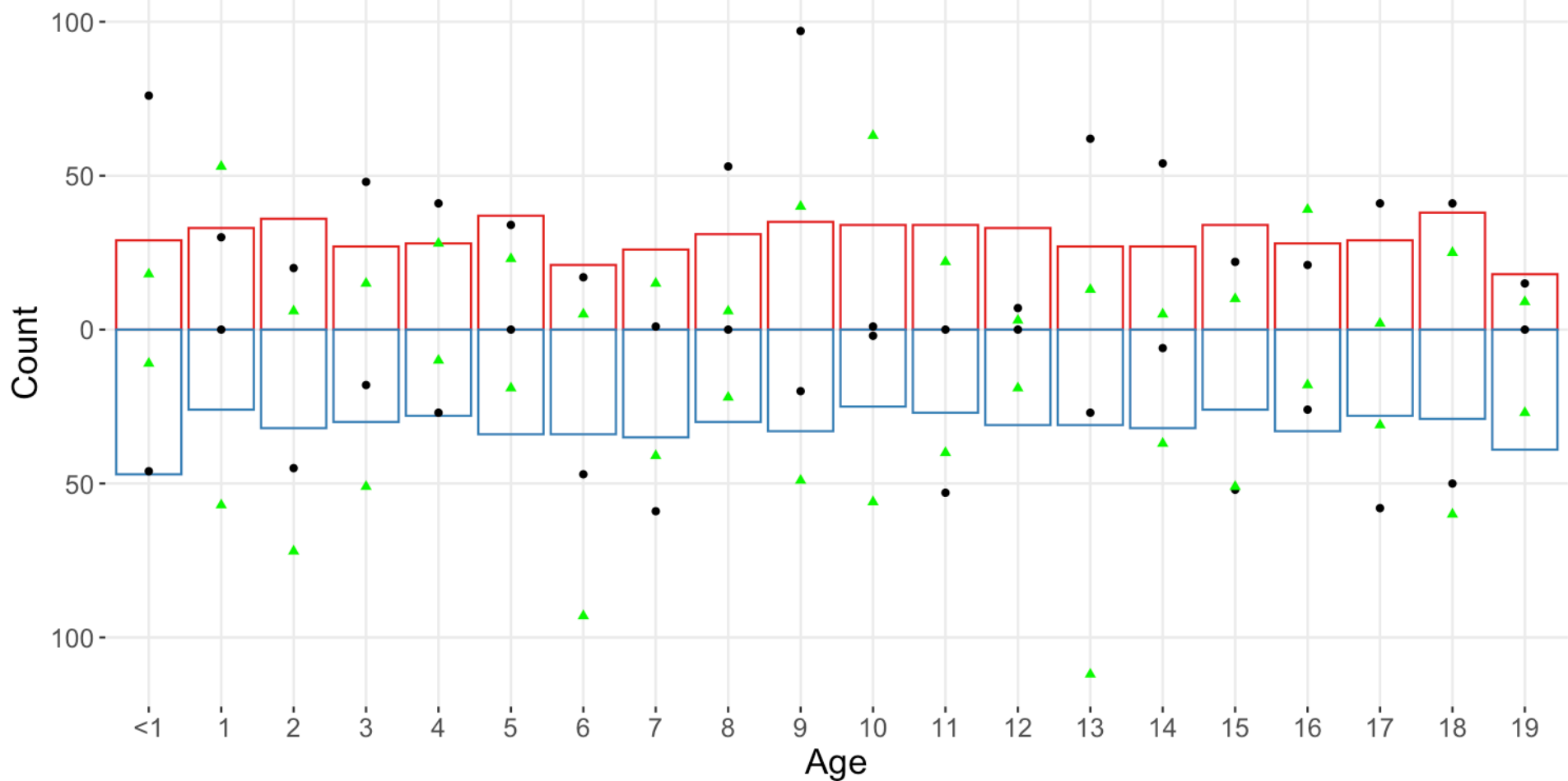


Source: US Census Bureau 2011; US Census Bureau 2019; Van Riper et al. 2020

# Sex by Single Year of Age: Wayzata city

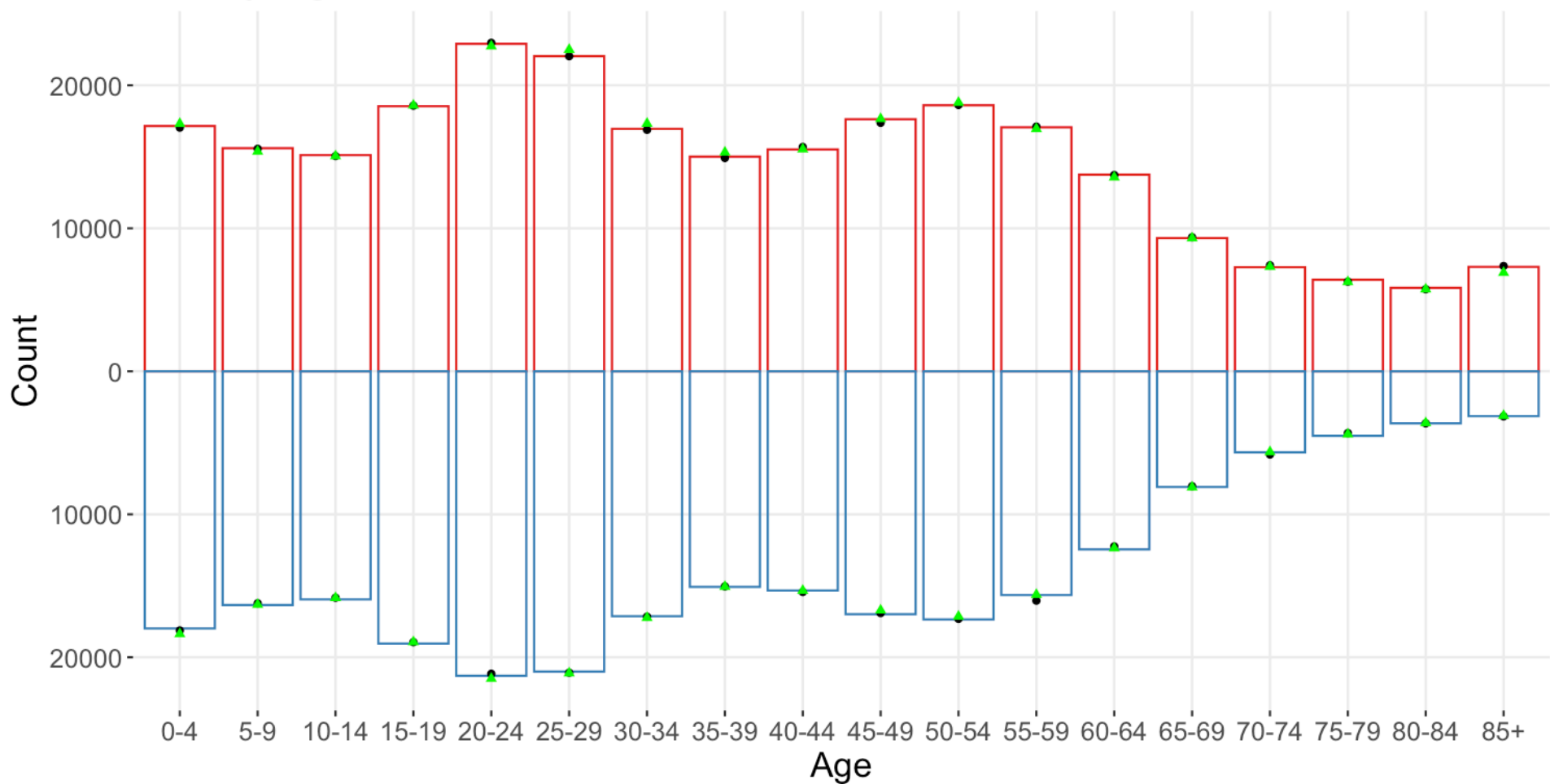


# Sex by Single Year of Age: Census Tract 307.03



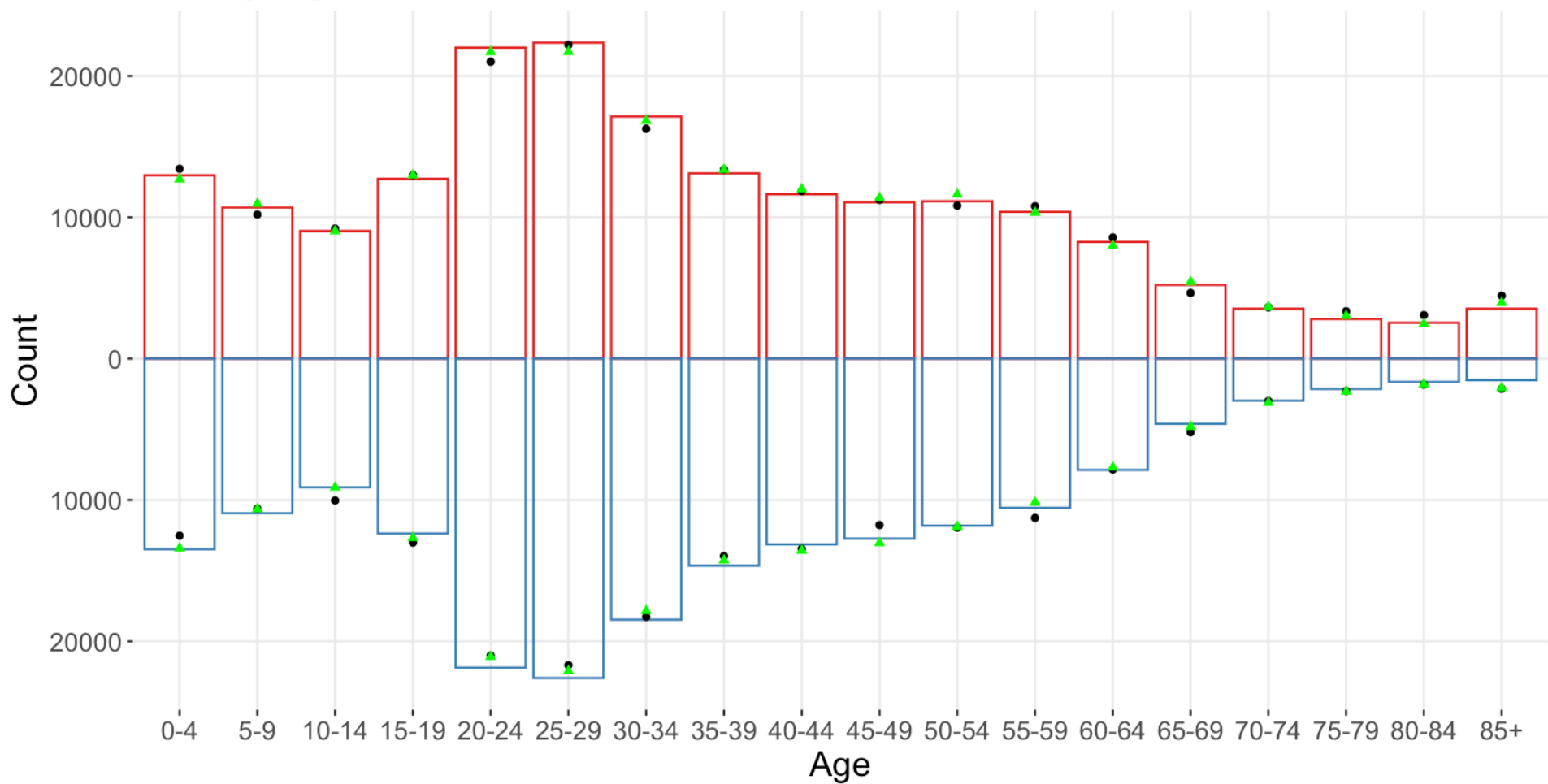
Source: US Census Bureau 2011; US Census Bureau 2019; Van Riper et al. 2020

# Sex by Age: G2701230



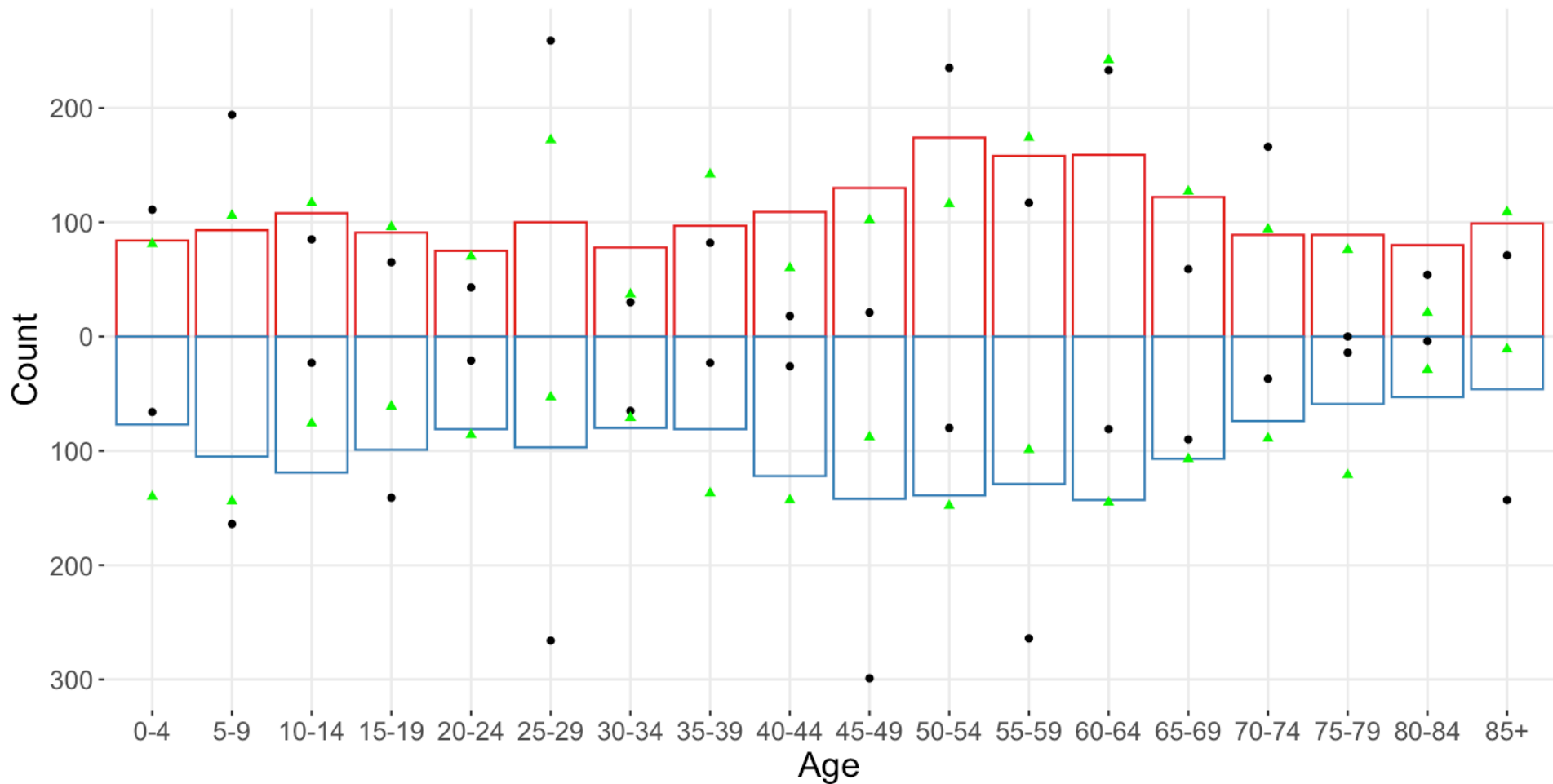
Source: US Census Bureau 2011; US Census Bureau 2019; Van Riper et al. 2020

# Sex by Age: G270053043000



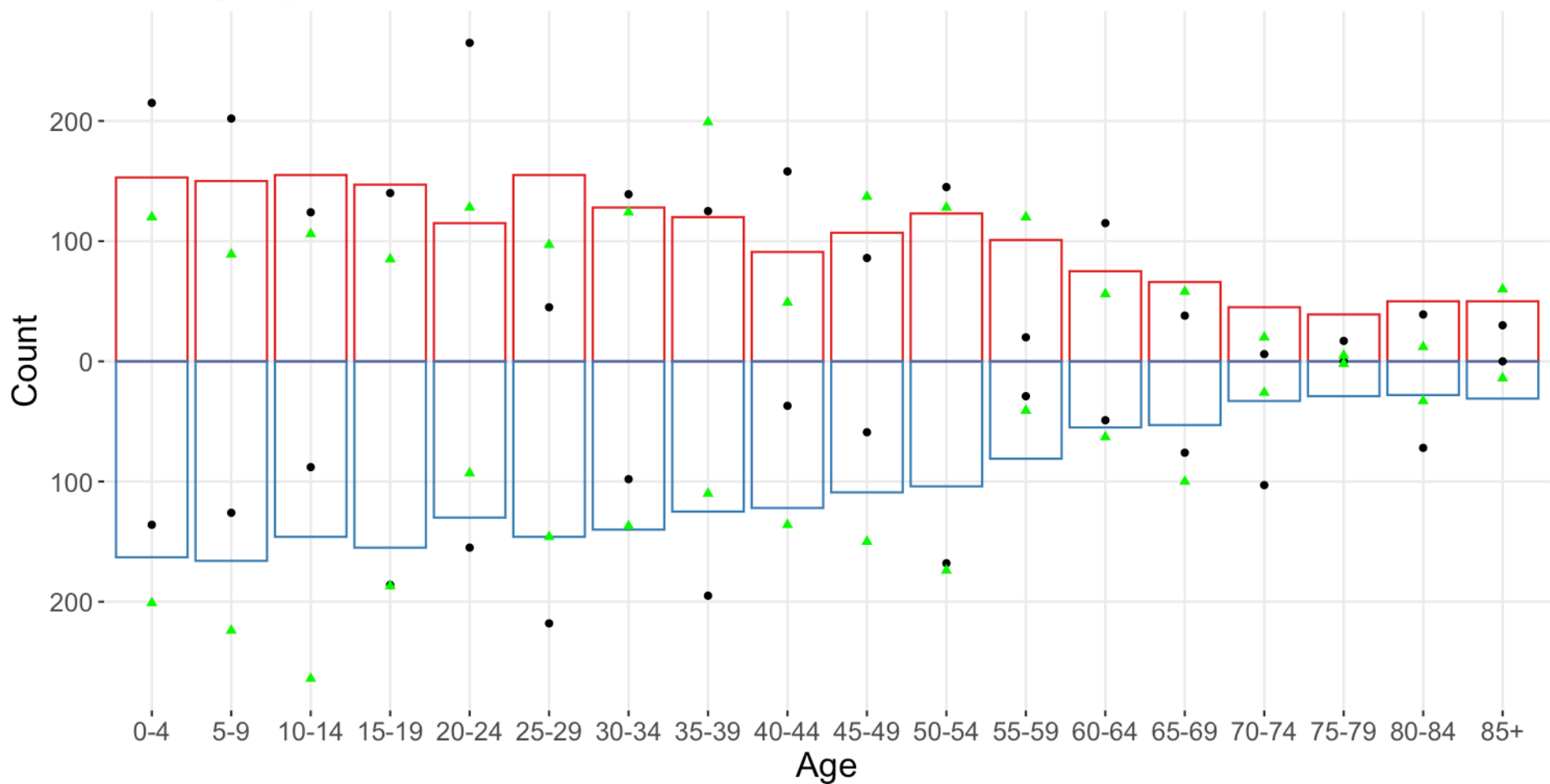
Source: US Census Bureau 2011; US Census Bureau 2019; Van Riper et al. 2020

# Sex by Age: G270053068818





# Sex by Age: G2701230030703



# What's next?

- Census will release another demo. dataset
  - This week?
  - Only supports redistricting tables

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  - Only supports redistricting tables
- Invariants will be set next week
- Final privacy loss budget and allocation set in early January