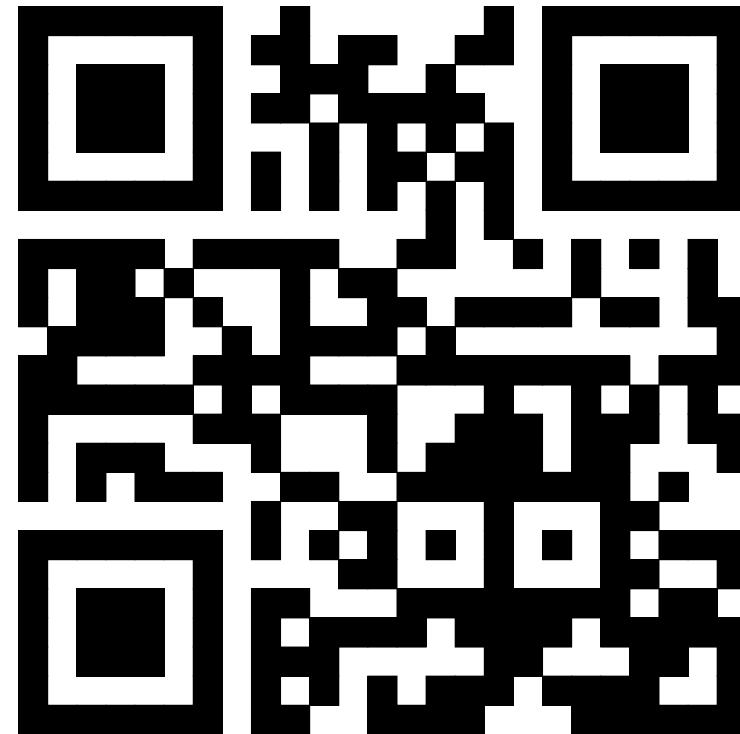


# How to Make a Thousand Plots Look Good: Data Viz Tips for Parameterized Reporting

# Follow Along!



<https://rfor.us/cascadia2024>

## BAKER COUNTY

Total population  
**16,539**

Rural population  
**41%**

Net migration, 2020-2021  
(per 1,000 population)  
**18**

### FEDERALLY RECOGNIZED TRIBES

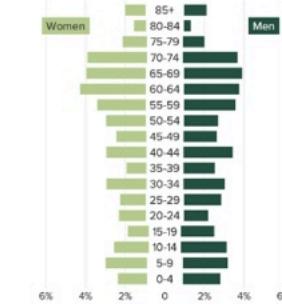


### MEDIAN HOUSEHOLD INCOME



**LIFE EXPECTANCY** 80 years 75 years

### POPULATION BY AGE

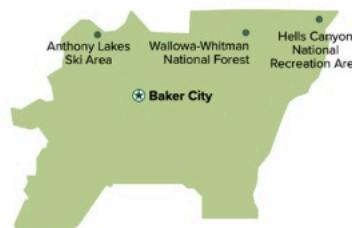


16 Oregon by the Numbers

Total land area  
**3,088 mi<sup>2</sup>**

Public land  
**52%**

Developed/cultivated land  
**8%**



★ County seat ○ Largest community

### HOUSEHOLDS IN FINANCIAL HARDSHIP



### POPULATION BY RACE/ETHNICITY



### TOP EMPLOYMENT INDUSTRIES



2023

## MULTNOMAH COUNTY

Total population  
**810,011**

Rural population  
**1%**

Net migration, 2020-2021  
(per 1,000 population)  
**5**

### FEDERALLY RECOGNIZED TRIBES

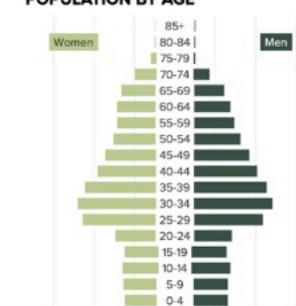


### MEDIAN HOUSEHOLD INCOME



**LIFE EXPECTANCY** 82 years 76 years

### POPULATION BY AGE



66 Oregon by the Numbers

Total land area  
**465 mi<sup>2</sup>**

Public land  
**37%**

Developed/cultivated land  
**42%**



★ County seat ○ Largest community

### HOUSEHOLDS IN FINANCIAL HARDSHIP



### POPULATION BY RACE/ETHNICITY



### TOP EMPLOYMENT INDUSTRIES



2023

## HOUSING

### 2022 Housing Data Profiles

### HARTFORD

SINGLE-FAMILY HOMES AS PERCENT OF ALL HOMES

**19%**

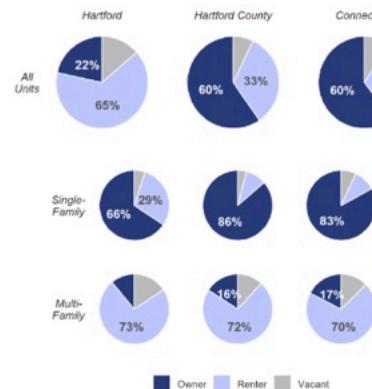


PERCENT OF ALL HOMES OCCUPIED BY OWNERS

**22%**

Overall, 65% of Connecticut's occupied housing stock is comprised of single-family housing, while 35% is multifamily housing (2+ units in structure). Most single-family homes are occupied by homeowners, while most multifamily units are occupied by renters.

In Hartford, 19% of occupied homes are single-family, and 81% are multifamily. Owners live in 66% of Hartford's 10,626 single-family homes, and renters live in 73% of its 44,561 multifamily homes.



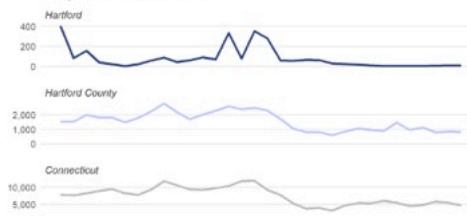
CHANGE IN BUILDING PERMITS, 1990-2021

**-97%**

Growth is slow in the state, which has seen a 41% decrease in building permits between 1990 and 2021.

In Hartford, there were 405 building permits issued in 1990, compared to 11 issued in 2021, representing a 97% decrease.

Number of building permits per year, 1990-2021  
Note: y axis varies between locations



Source: Connecticut Department of Economic and Community Development

## HOUSING

### 2022 Housing Data Profiles

### ANDOVER

SINGLE-FAMILY HOMES AS PERCENT OF ALL HOMES

**90%**

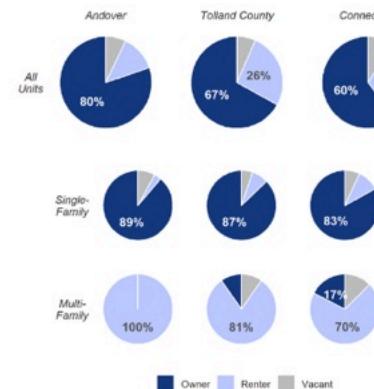


PERCENT OF ALL HOMES OCCUPIED BY OWNERS

**80%**

Overall, 65% of Connecticut's occupied housing stock is comprised of single-family housing, while 35% is multifamily housing (2+ units in structure). Most single-family homes are occupied by homeowners, while most multifamily units are occupied by renters.

In Andover, 90% of occupied homes are single-family, and 10% are multifamily. Owners live in 89% of Andover's 1,161 single-family homes, and renters live in 100% of its 126 multifamily homes.



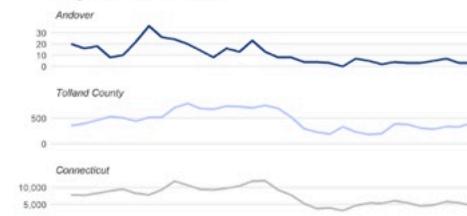
CHANGE IN BUILDING PERMITS, 1990-2021

**-85%**

Growth is slow in the state, which has seen a 41% decrease in building permits between 1990 and 2021.

In Andover, there were 20 building permits issued in 1990, compared to 3 issued in 2021, representing a 85% decrease.

Number of building permits per year, 1990-2021  
Note: y axis varies between locations



Source: Connecticut Department of Economic and Community Development



## AFGHANISTAN

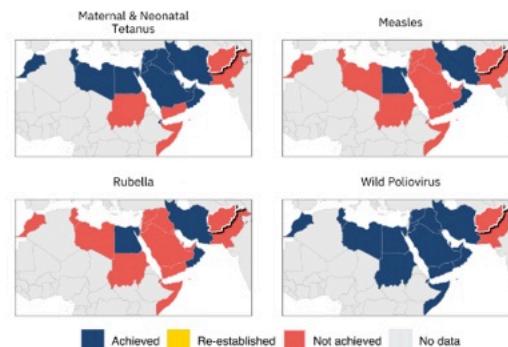
### Impact Goals



#### 1.2 Control, Eliminate & Eradicate Vaccine-Preventable Disease

In 2022, measles elimination was **achieved** in 4 countries in the WHO region, while it was **re-established** in 0 countries and **not achieved** in 17 countries.

The presentation of maps is not an expression of the opinion of Immunization Agenda 2030 regarding the legal status of countries/territories, their governing authorities, or their official borders.



#### Achievement of vaccine preventable disease control, elimination, and eradication targets in Afghanistan:

	Baseline	2021	2022
Measles	✗	✗	✗
Rubella	✗	✗	✗
Wild Poliovirus	✗	✗	✗
Maternal & Neonatal Tetanus	✗	✗	✗

✓ : Achieved

✗ : Not achieved

🟡 : Re-established

Grey : No data

Data source: Verification, certification, and disease-specific committee reports. Targets for vaccine-preventable disease are elimination of measles transmission, elimination of rubella transmission, eradication of wildtype poliovirus, and elimination as a public health problem for maternal and neonatal tetanus. The baseline year is 2020.

## CONGO (DEM. REP.)

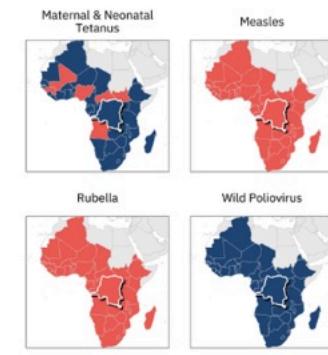
### Impact Goals



#### 1.2 Control, Eliminate & Eradicate Vaccine-Preventable Disease

In 2022, measles elimination was **achieved** in 0 countries in the WHO region, while it was **re-established** in 0 countries and **not achieved** in 47 countries.

The presentation of maps is not an expression of the opinion of Immunization Agenda 2030 regarding the legal status of countries/territories, their governing authorities, or their official borders.



#### Achievement of vaccine preventable disease control, elimination, and eradication targets in Democratic Republic of the Congo:

	Baseline	2021	2022
Measles	✗	✗	✗
Rubella	✗	✗	✗
Wild Poliovirus	✓	✓	✓
Maternal & Neonatal Tetanus	✓	✓	✓

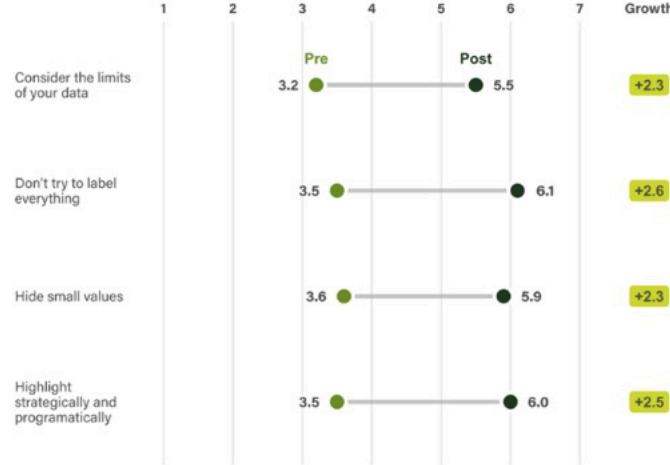
✓ : Achieved

✗ : Not achieved

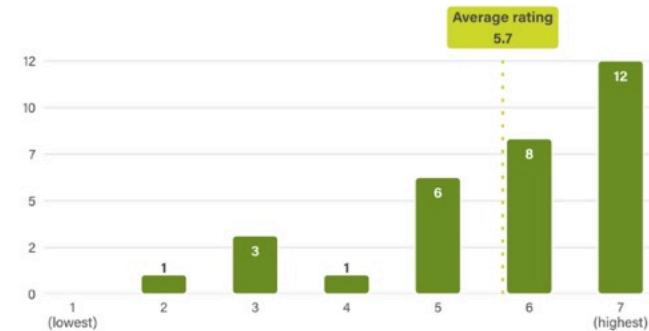
🟡 : Re-established

Grey : No data

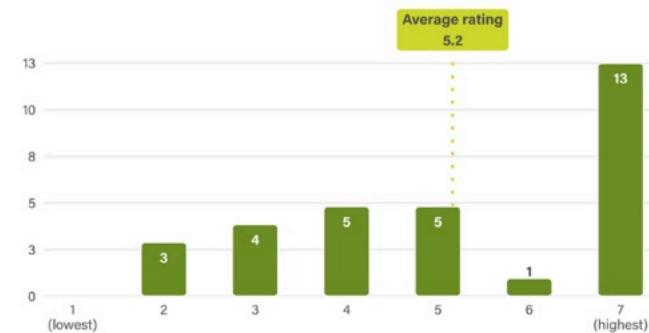
Data source: Verification, certification, and disease-specific committee reports. Targets for vaccine-preventable disease are elimination of measles transmission, elimination of rubella transmission, eradication of wildtype poliovirus, and elimination as a public health problem for maternal and neonatal tetanus. The baseline year is 2020.

**PRE-POST Participant Self Assessed Knowledge Ratings****Scale Ratings**

How well were the identified learning objectives covered?



How effective was the interaction opportunities provided in keeping you engaged in training?



# Data Viz Tips for Parameterized Reporting

# There is No Magic Package



The background of the image is a dark, star-filled night sky. In the foreground, there are several snow-covered mountain peaks, their white slopes contrasting with the dark sky. The text is overlaid on the upper portion of the image.

# Consider the Outer Limits of Your Data

## MULTNOMAH COUNTY

Total population  
**810,011**

Total land area  
**465 mi<sup>2</sup>**

Rural population  
1%



Net migration, 2020-2021  
(per 1,000 population)  
**5**

### FEDERALLY RECOGNIZED TRIBES

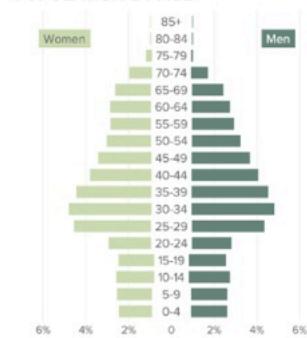


### MEDIAN HOUSEHOLD INCOME



LIFE EXPECTANCY      **82** years      **76** years

### POPULATION BY AGE



Public land  
**37%**      Developed/cultivated land  
**42%**



### HOUSEHOLDS IN FINANCIAL HARSHSHIP



### POPULATION BY RACE/ETHNICITY



### TOP EMPLOYMENT INDUSTRIES



Below  
ALICE

## POPULATION BY RACE/ETHNICITY

American Indian/Alaska Native: 0.6%

Asian: 7.6%

Black/African American: 5.1%

Hispanic/Latino: 12.1%

Native Hawaiian/Pacific Islander: 0.6%

Some other race: 0.4%

Two or more races: 5.5%

White: 68.1%

## TOP EMPLOYMENT INDUSTRIES



```
1 race_ethnicity

# A tibble: 288 × 4
  county population          pct pct_formatted
  <chr>   <fct>           <dbl> <chr>
1 Baker   American Indian/Alaska Native 0.0142  1.4%
2 Baker   Asian              0.00351  0.4%
3 Baker   Black/African American 0.0113  1.1%
4 Baker   Hispanic/Latino      0.0481  4.8%
5 Baker   Native Hawaiian/Pacific Islander 0.00193 0.2%
6 Baker   Some other race     0.000544 0.1%
7 Baker   Two or more races   0.0244  2.4%
8 Baker   White              0.896   89.6%
9 Benton  American Indian/Alaska Native 0.00418  0.4%
10 Benton Asian             0.0729  7.3%
# i 278 more rows
```

```
1 race_ethnicity_bar_chart <- function(county_to_plot) {
2   race_ethnicity |>
3     filter(county == county_to_plot) |>
4     ggplot(
5       aes(
6         x = pct,
7         y = population
8       )
9     ) +
10    geom_col(fill = "#004f39") +
11    ...
12 }
```

```
1 race_ethnicity_bar_chart(county_to_plot = "Multnomah")
```

## Multnomah



```
1 race_ethnicity_bar_chart(county_to_plot = "Multnomah")
```

## Multnomah



```
1 race_ethnicity_bar_chart(county_to_plot = "Baker")
```

## Baker





## MULTNOMAH COUNTY

Total population  
**810,011**

Total land area  
**465 mi<sup>2</sup>**

Rural population  
1%



Net migration, 2020-2021  
(per 1,000 population)  
**5**

### FEDERALLY RECOGNIZED TRIBES

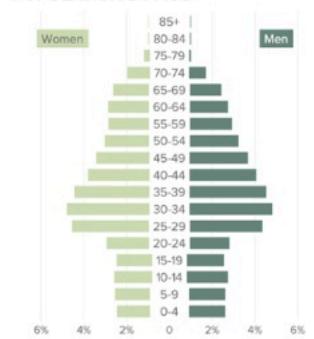


### MEDIAN HOUSEHOLD INCOME



LIFE EXPECTANCY      **82** years      **76** years

### POPULATION BY AGE



Public land  
**37%**      Developed/cultivated land  
**42%**



### HOUSEHOLDS IN FINANCIAL HARSHSHIP



### POPULATION BY RACE/ETHNICITY



### TOP EMPLOYMENT INDUSTRIES



Below  
ALICE

## POPULATION BY RACE/ETHNICITY

American Indian/Alaska Native: 0.6%

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Hispanic/Latino: 12.1%

Native Hawaiian/Pacific Islander: 0.6%

Some other race: 0.4%

Two or more races: 5.5%

White: 68.1%

## TOP EMPLOYMENT INDUSTRIES



```
1 race_ethnicity_bar_chart(county_to_plot = "Multnomah") +
2   geom_col(
3     aes(
4       x = 1
5     ),
6     fill = "transparent",
7     color = "#A9C27F",
8     linetype = "dotted"
9   )
```

## Multnomah



## Baker





## MULTNOMAH COUNTY

Total population  
**810,011**

Total land area  
**465 mi<sup>2</sup>**

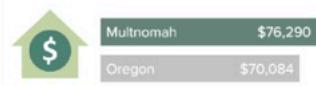
Rural population  
1%

Net migration, 2020-2021  
(per 1,000 population)  
**5**

### FEDERALLY RECOGNIZED TRIBES

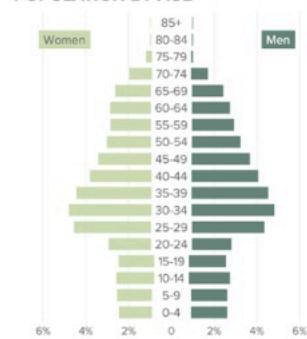


### MEDIAN HOUSEHOLD INCOME



LIFE EXPECTANCY    82 years    76 years

### POPULATION BY AGE



Public land  
**37%**      Developed/cultivated land  
**42%**



### HOUSEHOLDS IN FINANCIAL HARSHSHIP



### POPULATION BY RACE/ETHNICITY



### TOP EMPLOYMENT INDUSTRIES



2023

COQ  
KLA  
SLZ  
UMA  
WSP

## MEDIAN HOUSEHOLD INCOME



Multnomah      \$76,290

Oregon      \$70,084

## LIFE EXPECTANCY

82 years    76 years

## POPULATION BY AGE

82 years    76 years

```
1 median_income

# A tibble: 37 × 3
  geography amount amount_formatted
  <chr>     <dbl> <chr>
1 Oregon      76632 $76,632
2 Baker        51657 $51,657
3 Benton       72882 $72,882
4 Clackamas   95740 $95,740
5 Clatsop      68025 $68,025
6 Columbia     83265 $83,265
7 Coos         57563 $57,563
8 Crook        74969 $74,969
9 Curry         64300 $64,300
10 Deschutes    82042 $82,042
# i 27 more rows
```

```
1 median_income_plot <- function(county_to_plot) {
2   median_income |>
3     filter(geography %in% c(county_to_plot, "Oregon")) |>
4     ggplot(
5       aes(
6         x = amount,
7         y = geography,
8         label = amount_formatted,
9         fill = geography
10        )
11      ) +
12      geom_col() +
13      ...
14 }
```

```
1 median_income_plot(county_to_plot = "Jackson")
```

Jackson	\$67,690
---------	----------

Oregon	\$76,632
--------	----------

Jackson \$67,690

Oregon \$76,632

Harney \$45,462

Oregon \$76,632

Washington \$100,121

Oregon \$76,632

```
1 max_median_income <-
2   median_income |>
3     slice_max(
4       order_by = amount,
5       n = 1
6     ) |>
7     pull(amount)
```

```
1 max_median_income
```

```
[1] 100121
```

```
1 median_income_plot(county_to_plot = "Jackson") +  
2   scale_x_continuous(  
3     limits = c(0, max_median_income)  
4   )
```

Jackson

\$67,690

Oregon

\$76,632

Jackson

\$67,690

Oregon

\$76,632

Harney

\$45,462

Oregon

\$76,632

Washington

\$100,121

Oregon

\$76,632

A close-up photograph of a person's arm and hand resting on a dark, textured surface, possibly a sofa or chair. The lighting is dramatic, with strong highlights on the skin and shadows in the creases. A small portion of a white garment is visible at the bottom left.

**Minimize Text and  
Position it Carefully**



**Don't Try to Label  
Everything**

## POPULATION

2022 Housing Data Profiles

HARTFORD

TOTAL POPULATION

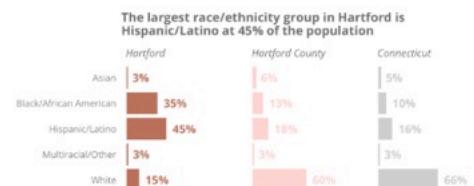
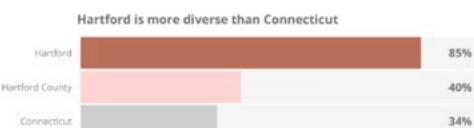
**122,549**



PEOPLE OF COLOR

**85%**

Connecticut population is becoming increasingly diverse, but the BIPOC population is concentrated in certain municipalities, especially Connecticut's cities. In Hartford, 85% of residents are BIPOC, while 15% are white.



MEDIAN AGE

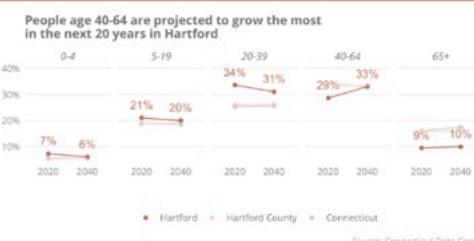
**31**



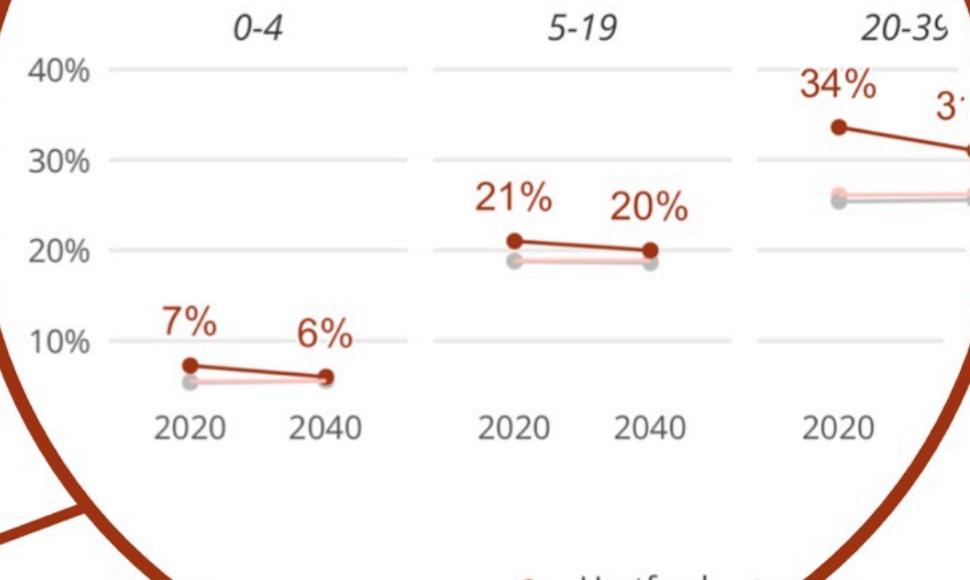
POPULATION CHANGE, 2020 TO 2040

**+0.3%**

In the next twenty years, Hartford's population is projected to grow from 126,443 to 126,846.



**People age 40-64 are projected to grow in the next 20 years in Hartford**



• Hartford

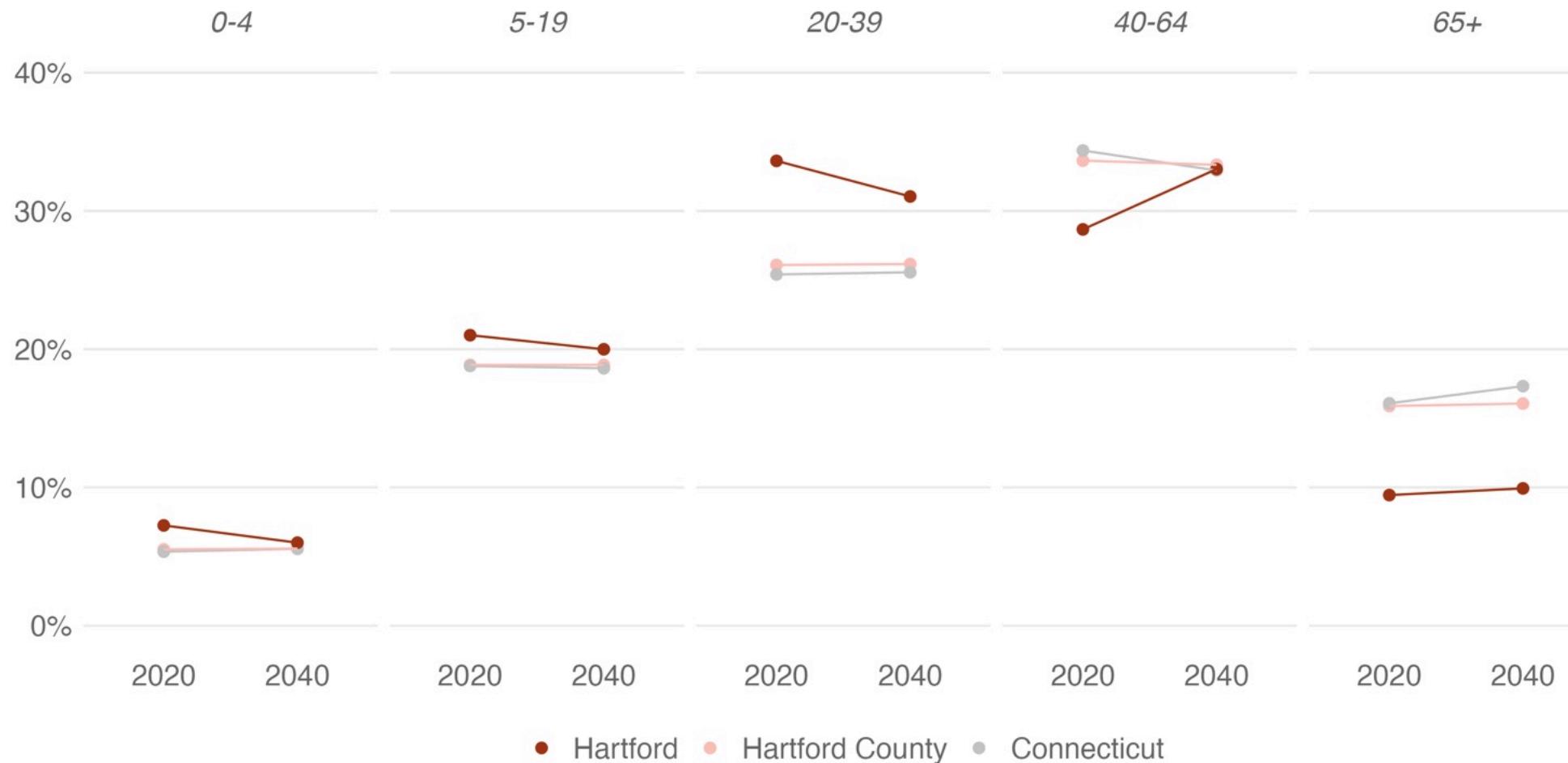


```
1 population_projection |>
2   filter(location == "Hartford")

# A tibble: 10 × 5
  location year age_group     pct pct_formatted
  <chr>    <chr> <fct>      <dbl> <chr>
1 Hartford 2020 0-4        0.0725 7%
2 Hartford 2020 5-19       0.210   21%
3 Hartford 2020 20-39      0.336   34%
4 Hartford 2020 40-64      0.287   29%
5 Hartford 2020 65+        0.0945 9%
6 Hartford 2040 0-4        0.0600 6%
7 Hartford 2040 5-19       0.200   20%
8 Hartford 2040 20-39      0.310   31%
9 Hartford 2040 40-64      0.330   33%
10 Hartford 2040 65+       0.0993 10%
```

```
1 population_projection_plot <- function(town_to_plot, county_to_plot) {
2   population_projection |>
3     filter(location %in% c(town_to_plot, county_to_plot, "Connecticut")) |>
4     ggplot(aes(
5       x = year,
6       y = pct,
7       color = location,
8       group = location
9     )) +
10    geom_point() +
11    geom_line() +
12    ...
13 }
```

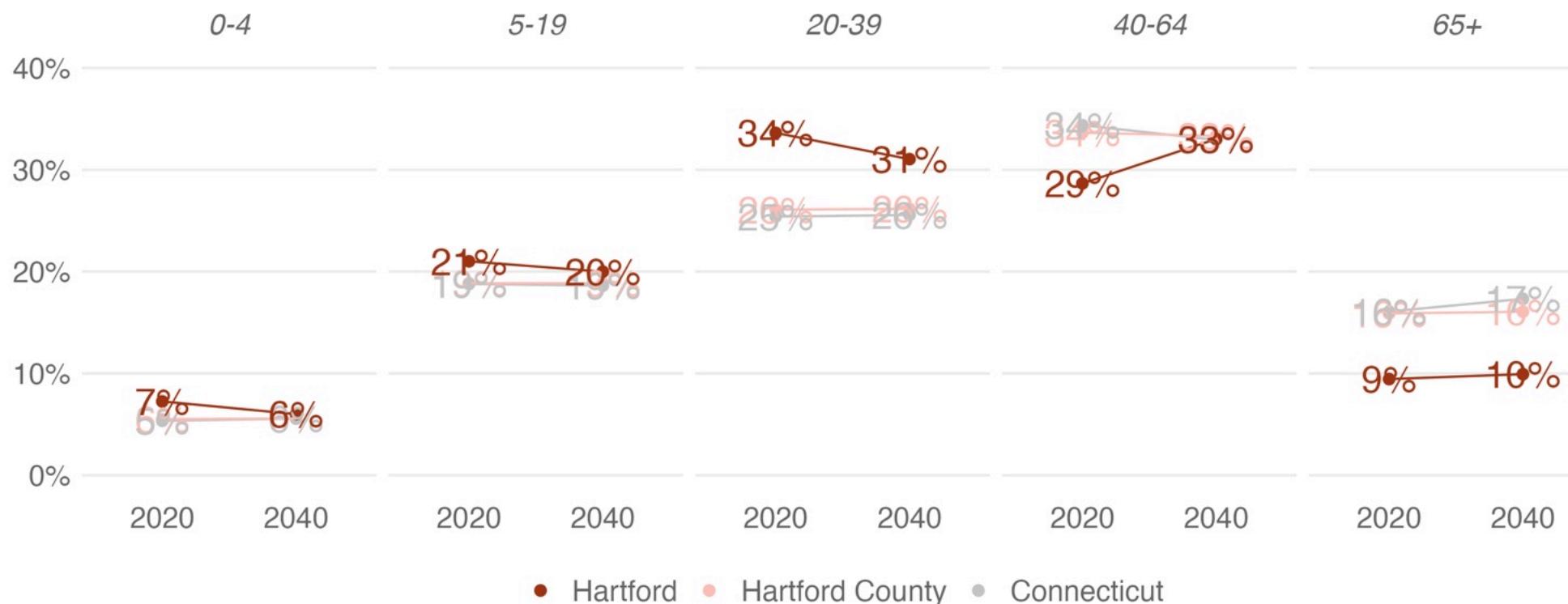
```
1 population_projection_plot(  
2     town_to_plot = "Hartford",  
3     county_to_plot = "Hartford County"  
4 )
```



```

1 population_projection_plot(
2   town_to_plot = "Hartford",
3   county_to_plot = "Hartford County"
4 ) +
5   geom_text(
6     aes(
7       label = pct_formatted
8     )
9   )

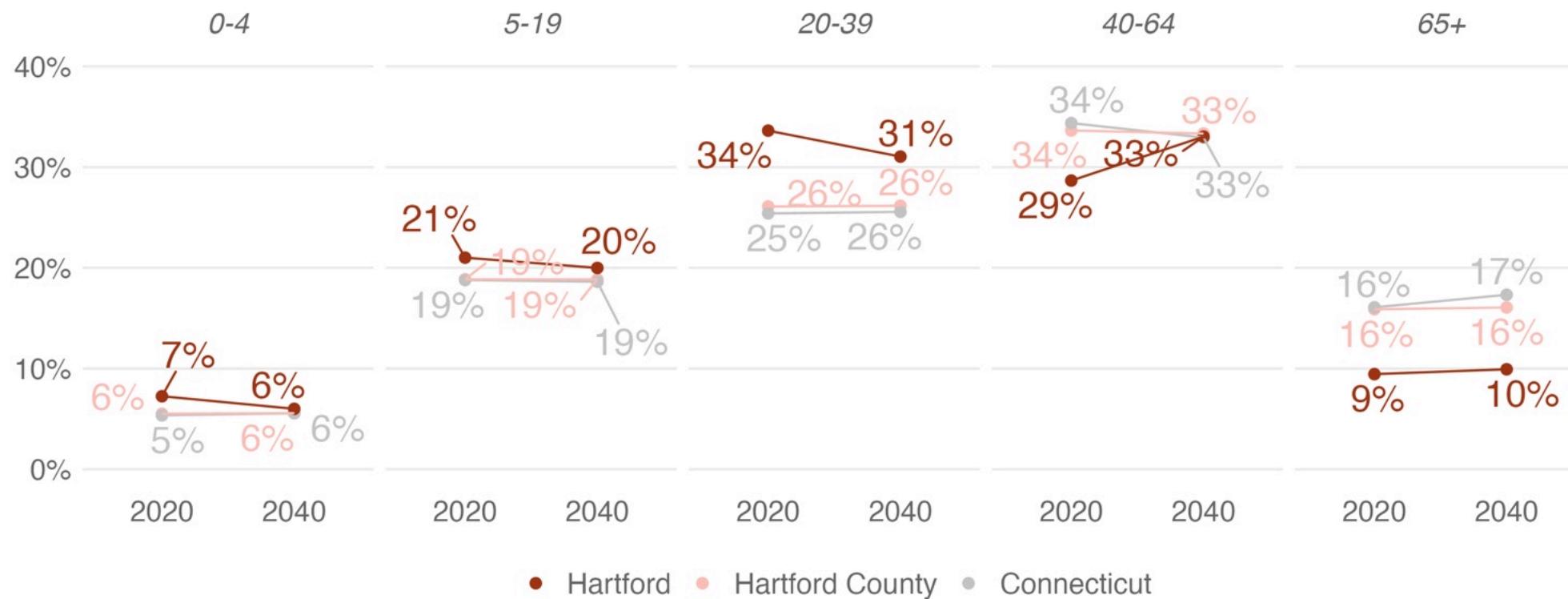
```



```

1 library(ggrepel)
2
3 population_projection_plot(
4   town_to_plot = "Hartford",
5   county_to_plot = "Hartford County"
6 ) +
7   geom_text_repel(
8     aes(
9       label = pct_formatted
10      )
11    )

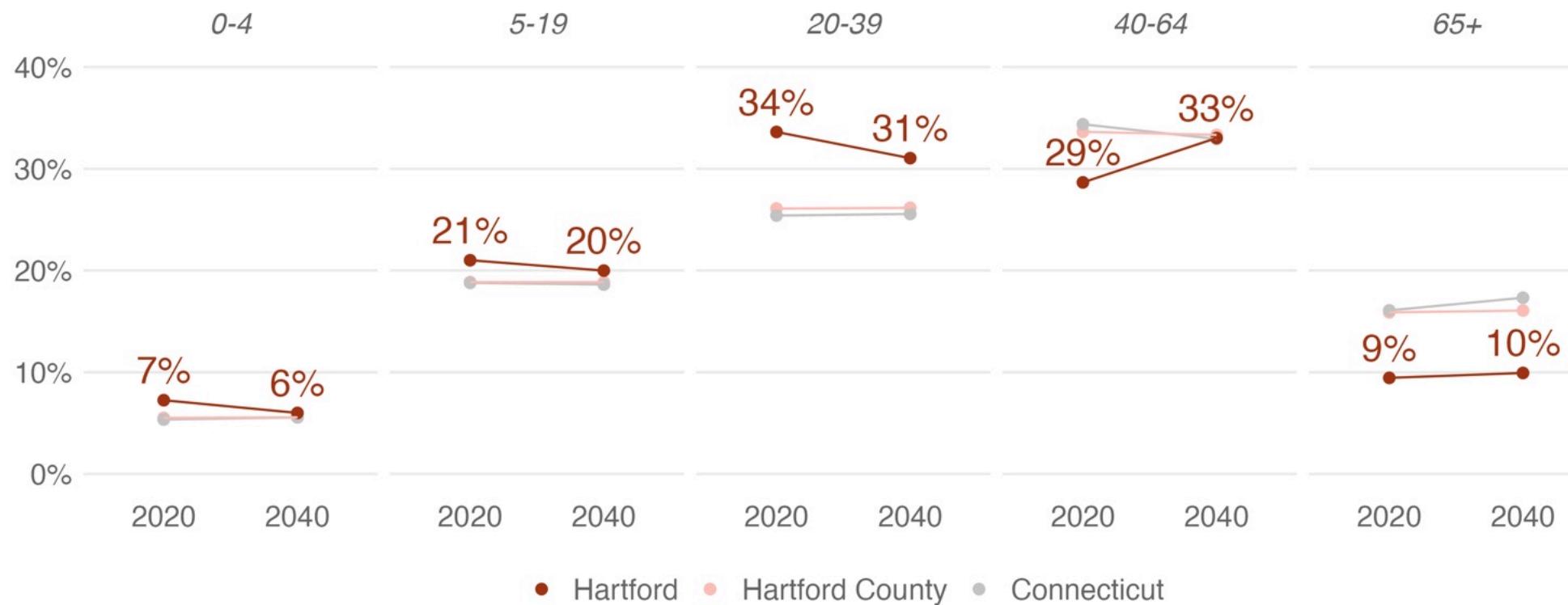
```



```

1 population_projection_plot(
2   town_to_plot = "Hartford",
3   county_to_plot = "Hartford County"
4 ) +
5   geom_text(
6     data = population_projection |> filter(location == "Hartford"),
7     nudge_y = 0.03,
8     aes(
9       label = pct_formatted
10      )
11    )

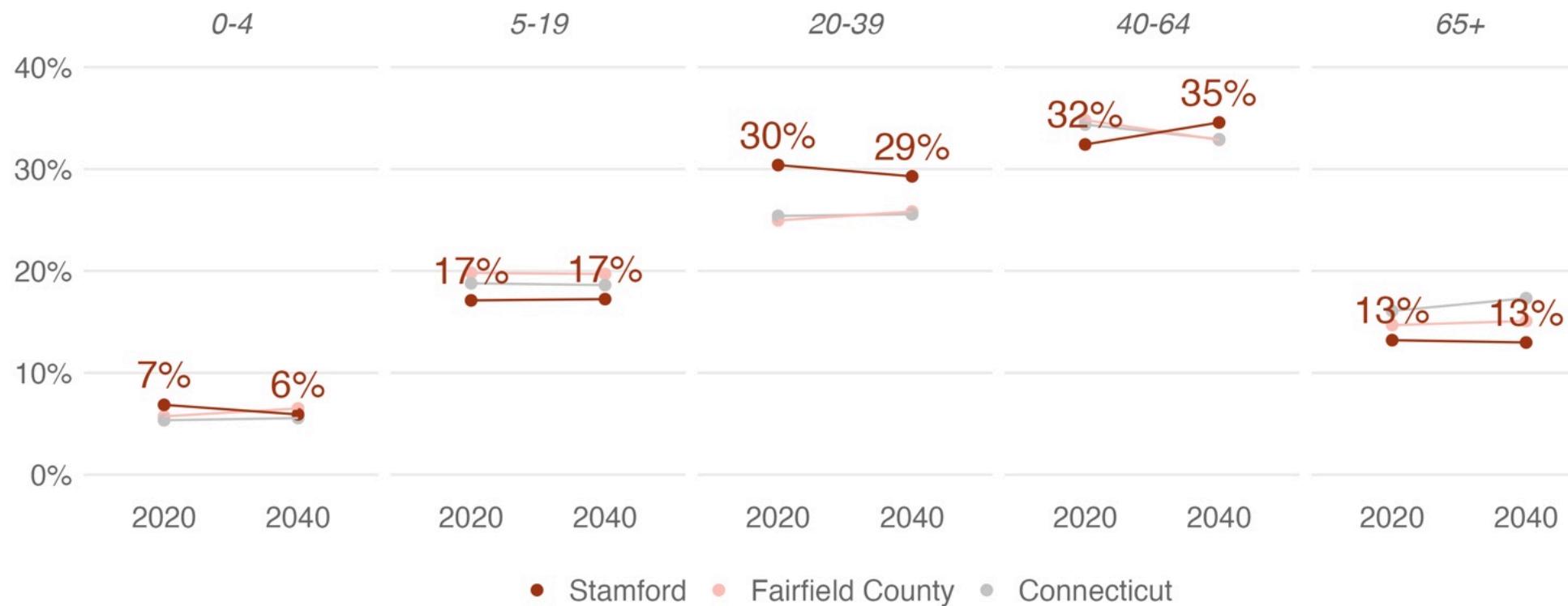
```



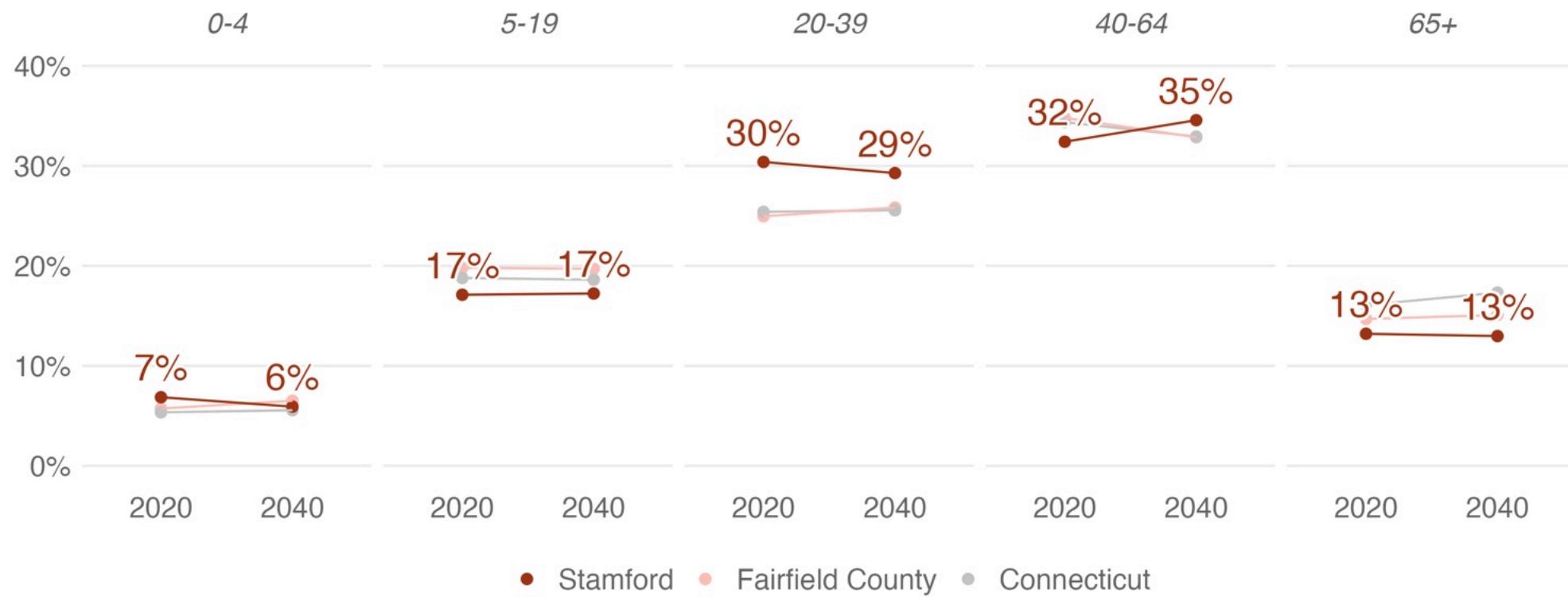
```

1 population_projection_plot(
2   town_to_plot = "Stamford",
3   county_to_plot = "Fairfield County"
4 ) +
5   geom_text(
6     data = population_projection |> filter(location == "Stamford"),
7     nudge_y = 0.03,
8     aes(
9       label = pct_formatted
10      )
11    )

```



```
1 library(shadowtext)
2
3 population_projection_plot(
4   town_to_plot = "Stamford",
5   county_to_plot = "Fairfield County"
6 ) +
7   geom_shadowtext(
8     data = population_projection |> filter(location == "Stamford"),
9     bg.color = "white",
10    nudge_y = 0.03,
11    aes(
12      label = pct_formatted
13    )
14  )
```



1

2

3

# Hide Small Values

## AFFORDABILITY

2022 Housing Data Profiles

HARTFORD

PEOPLE BURDENED BY COST OF HOUSING

**49%**

Households that are cost-burdened spend more than 30% of their income on housing. Severely cost-burdened spend more than 50% on housing.



RENTERS BURDENED BY COST OF HOUSING

**59%**

Housing cost burden for renters			
Hartford	30%	22%	41%
Hartford County	25%	23%	46%
Connecticut	25%	24%	46%

OWNERS BURDENED BY COST OF HOUSING

**39%**

Housing cost burden for owners			
Hartford	17%	22%	61%
Hartford County	15%		75%
Connecticut	11%	16%	72%

Legend: Severe burden (50% or greater) | Moderate burden (Between 30% and 50%) | Not burdened (Less than 30%) | Not Computed

RENTERS' HOUSING COSTS AS PERCENT OF INCOME

**29%**



OWNERS' HOUSING COSTS AS PERCENT OF INCOME

**19%**



## Housing cost burden for renters

30% | 22% | 41%

25% | 23% | 46%

25% | 24% | 46%

## Housing cost burden for owners

17% | 22% | 61%

15% | 75%

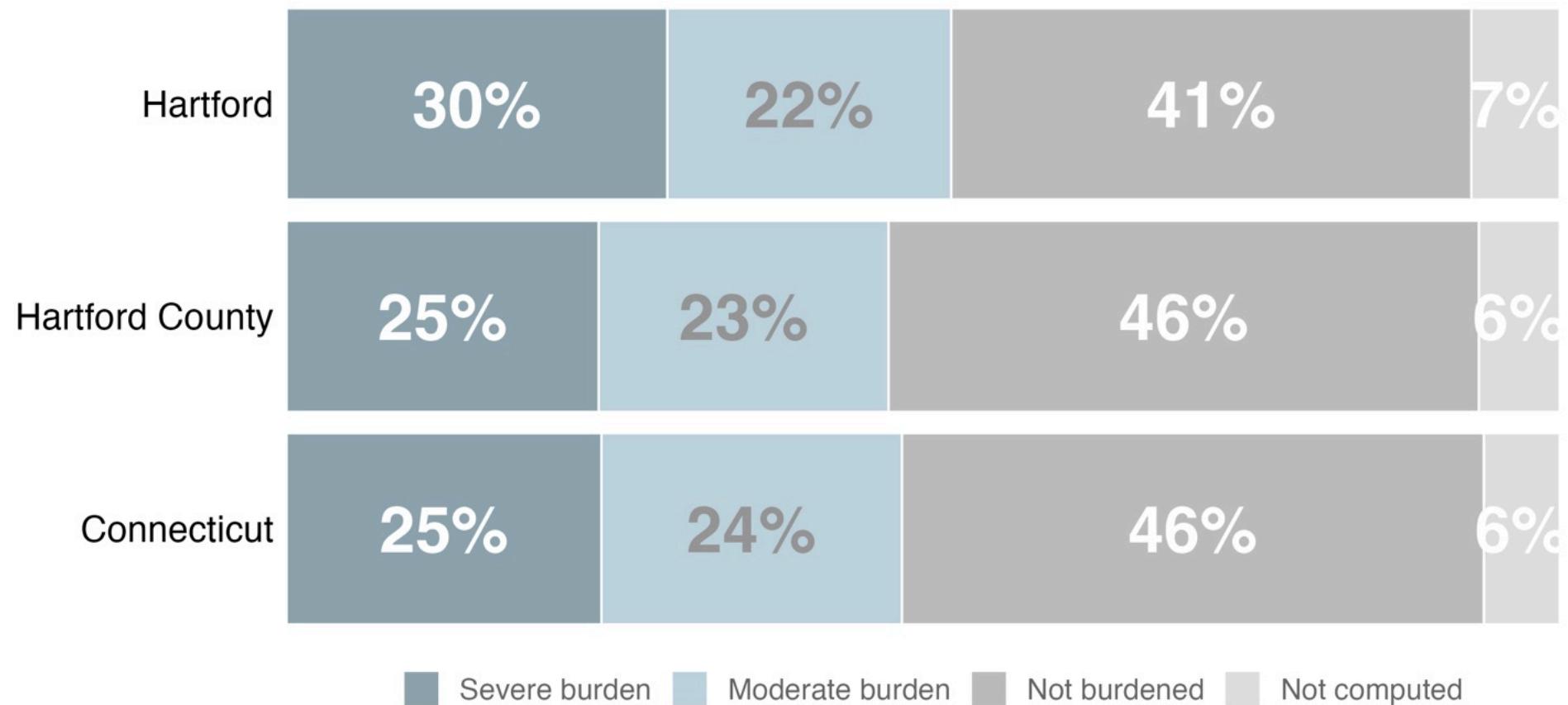
11% | 16% | 72%

```
1 housing_cost_burden

# A tibble: 694 × 4
  location burden_level      pct pct_formatted
  <fct>    <fct>        <dbl> <chr>
1 Bethel   Not burdened  0.381  38%
2 Bethel   Moderate burden 0.352  35%
3 Bethel   Severe burden  0.229  23%
4 Bethel   Not computed  0.0385 4%
5 Bridgeport Not burdened 0.358  36%
6 Bridgeport Moderate burden 0.260  26%
7 Bridgeport Severe burden 0.332  33%
8 Bridgeport Not computed 0.0494 5%
9 Brookfield Not burdened 0.501  50%
10 Brookfield Moderate burden 0.228  23%
# i 684 more rows
```

```
1 housing_cost_burden_plot <- function(town_to_plot, county_to_plot) {
2   housing_cost_burden |>
3     filter(location %in% c(town_to_plot, county_to_plot, "Connecticut")) |>
4     ggplot(aes(
5       x = pct,
6       y = location,
7       fill = burden_level,
8       label = pct_formatted
9     )) +
10    geom_col() +
11    geom_text(position = position_stack(vjust = 0.5)) +
12    ...
13 }
```

```
1 housing_cost_burden_plot(  
2   town_to_plot = "Hartford",  
3   county_to_plot = "Hartford County"  
4 )
```



```
1 housing_cost_burden |>  
2 filter(location == "Hartford")
```

```
# A tibble: 4 × 4  
  location burden_level      pct pct_formatted  
  <fct>    <fct>        <dbl> <chr>  
1 Hartford Not burdened  0.409  41%  
2 Hartford Moderate burden 0.223  22%  
3 Hartford Severe burden  0.299  30%  
4 Hartford Not computed   0.0694 7%
```

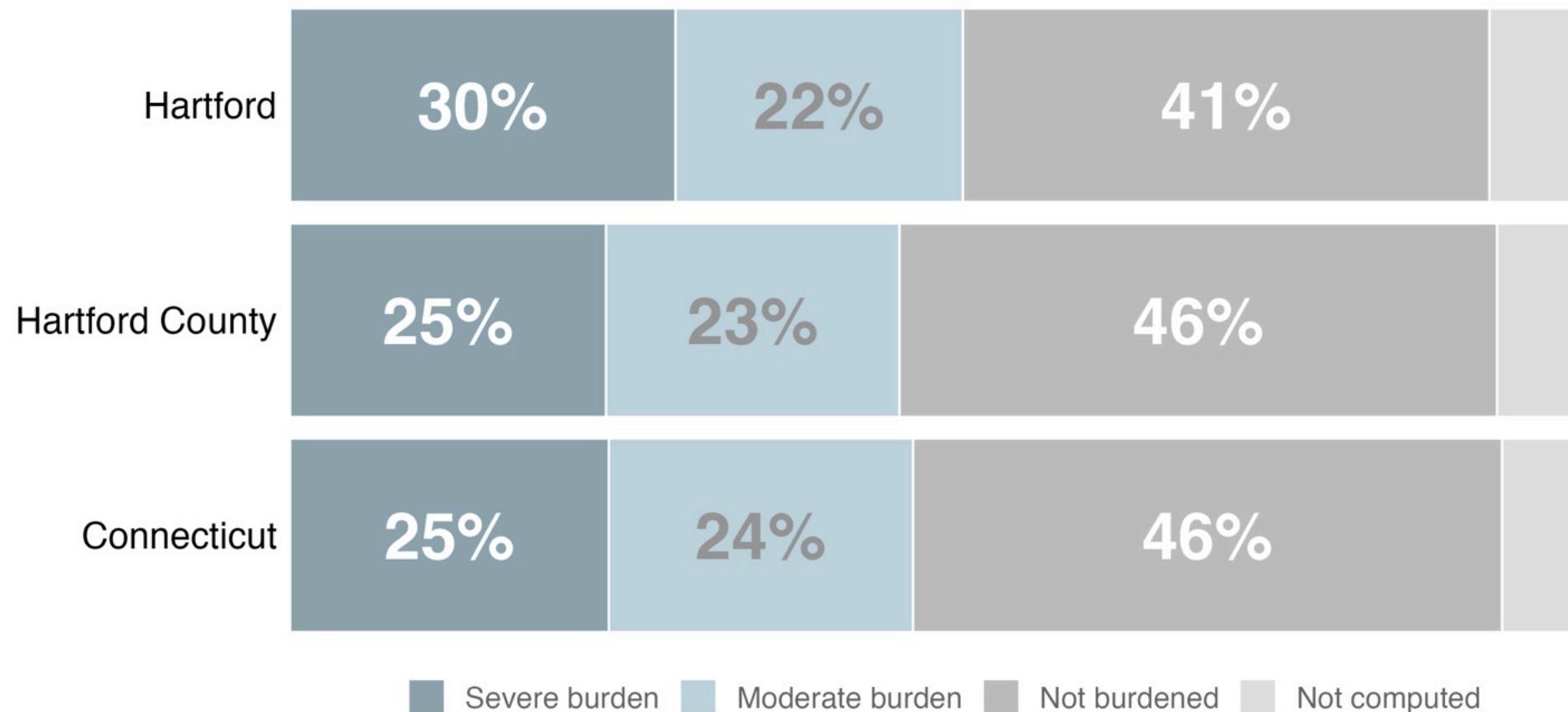
```
1 housing_cost_burden <-  
2 housing_cost_burden |>  
3 mutate(pct_formatted = if_else(pct > 0.07, pct_formatted, NA))
```

```
1 housing_cost_burden |>  
2 filter(location == "Hartford")
```

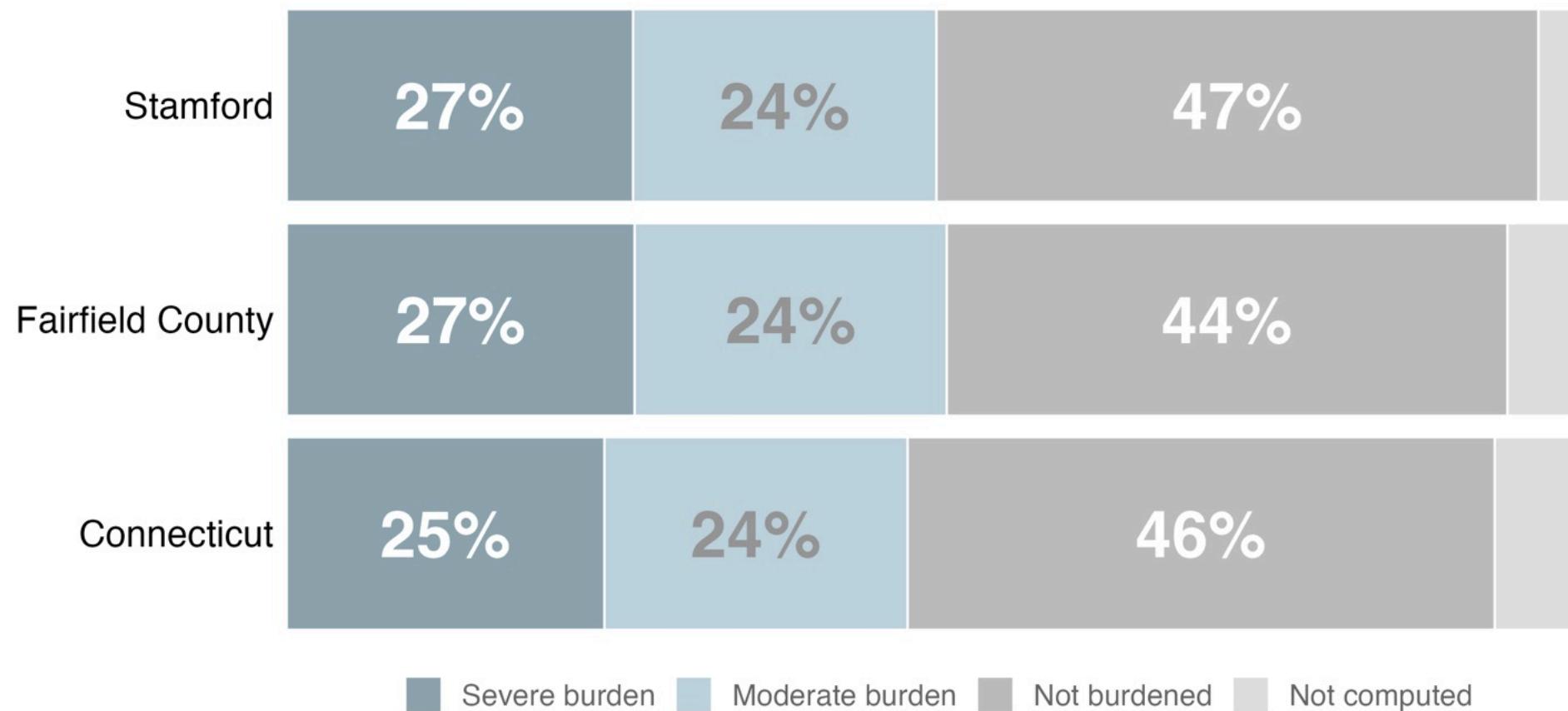
  

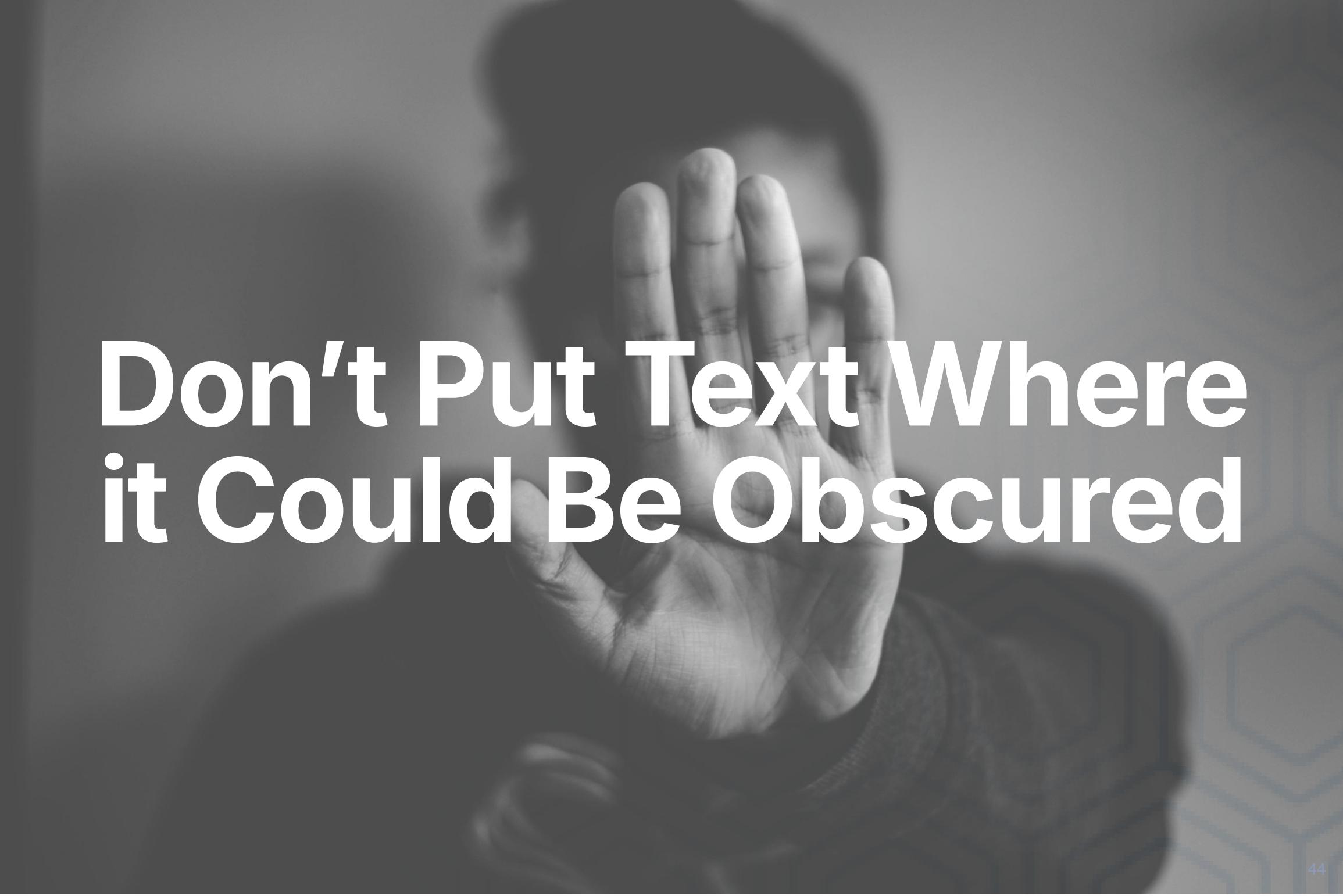
```
# A tibble: 4 × 4  
  location burden_level      pct pct_formatted  
  <fct>    <fct>        <dbl> <chr>  
1 Hartford Not burdened  0.409  41%  
2 Hartford Moderate burden 0.223  22%  
3 Hartford Severe burden  0.299  30%  
4 Hartford Not computed   0.0694 <NA>
```

```
1 housing_cost_burden_plot(  
2   town_to_plot = "Hartford",  
3   county_to_plot = "Hartford County"  
4 )
```

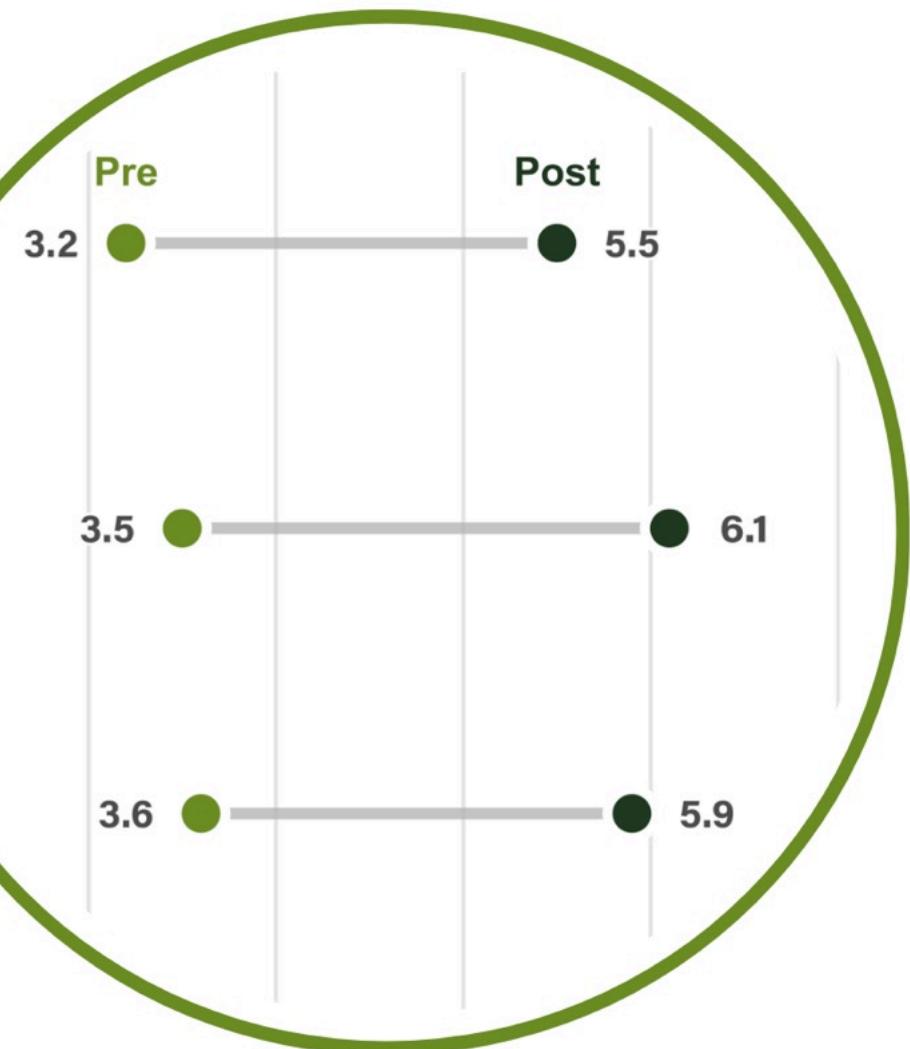
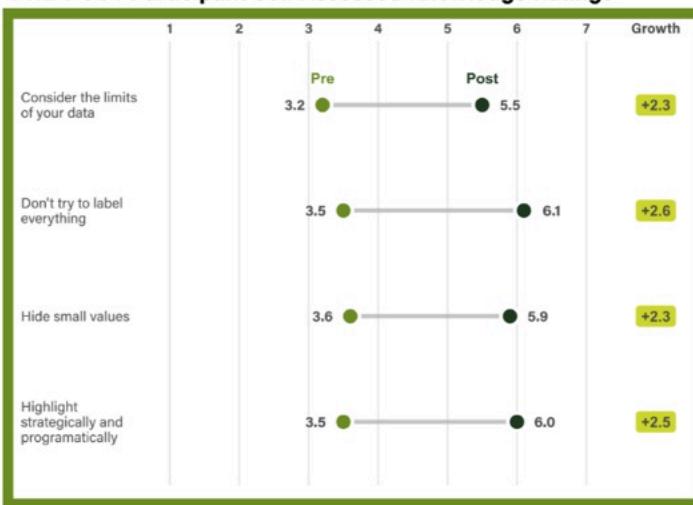


```
1 housing_cost_burden_plot(  
2     town_to_plot = "Stamford",  
3     county_to_plot = "Fairfield County"  
4 )
```



A black and white photograph showing a person's hands covering their face. The fingers are positioned to obscure the eyes, symbolizing how text can be hidden or obscured if placed inappropriately.

Don't Put Text Where  
it Could Be Obscured

**PRE-POST Participant Self Assessed Knowledge Ratings**

```
1 pre_post_data  
  
# A tibble: 2 × 6  
  question timing rating growth growth_formatted growth_text_position  
  <chr>     <chr>    <dbl>    <dbl>           <dbl>                <dbl>  
1 Question 1 Pre        1.6      NA             NA                  NA  
2 Question 1 Post       4.2      2.6            2.6                 2.9
```

```
1 pre_post_plot <- function(df) {  
  df |>  
  ggplot(aes(  
    x = rating,  
    y = question,  
    fill = timing  
  )) +  
  geom_line() +  
  geom_point(shape = 21) +  
  ...  
11 }
```

```
1 pre_post_data |>  
2 pre_post_plot()
```

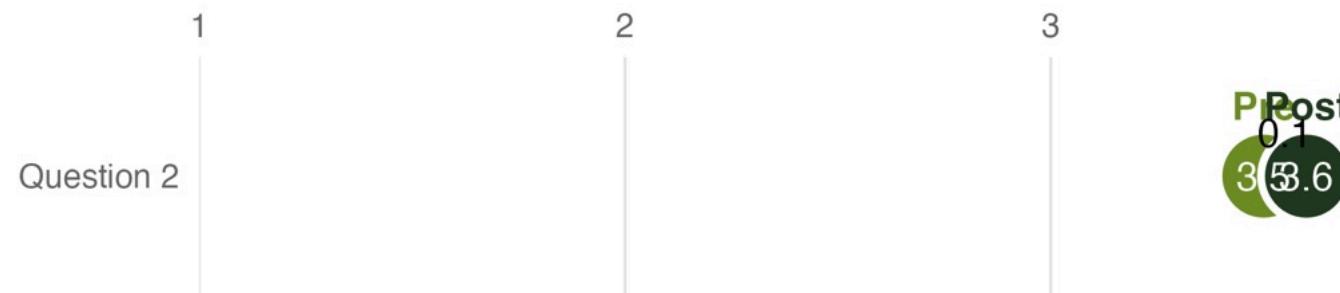


```
1 pre_post_data |>
2   pre_post_plot() +
3   # Ratings within points
4   geom_text(
5     aes(label = rating),
6     color = "white"
7   ) +
8   # Pre/post labels
9   geom_text(
10    aes(
11      label = timing,
12      color = timing
13    )
14  ) +
15 # Growth label
16 geom_text(
17  aes(
18    x = growth_text_position,
19    label = growth_formatted
20  )
21 )
```



```
1 pre_post_data
```

```
# A tibble: 2 × 6
  question timing rating growth growth_formatted growth_text_position
  <chr>     <chr>    <dbl>   <dbl>           <dbl>                  <dbl>
1 Question 2 Pre        3.5     NA             NA                   NA
2 Question 2 Post       3.6    0.100          0.100                 3.55
```



```
1 pre_post_data <-
2   pre_post_data |>
3     mutate(rating_text_position = case_when(
4       timing == "Pre" ~ rating - 0.2,
5       timing == "Post" ~ rating + 0.2
6     ))
```

```
1 pre_post_data |>
2   select(question, timing, rating, rating_text_position)
```

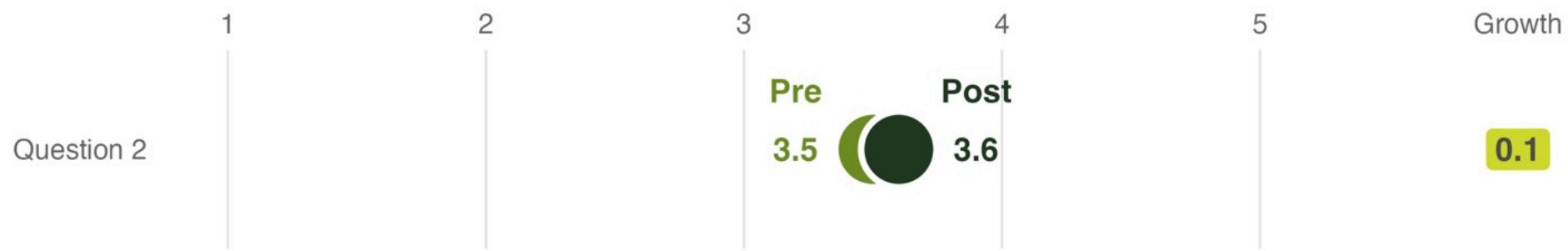
```
# A tibble: 2 × 4
question  timing  rating  rating_text_position
<chr>     <chr>    <dbl>                <dbl>
1 Question 2 Pre      3.5                  3.3
2 Question 2 Post     3.6                  3.8
```

```
1 pre_post_data |>
2   pre_post_plot() +
3   geom_text(
4     aes(
5       x = rating_text_position,
6       label = rating
7     )
8   ) +
9   geom_text(
10    aes(
11      x = rating_text_position,
12      label = timing
13    )
14  ) +
15  ...
```

```
1 pre_post_data |>
2   pre_post_plot() +
3   geom_text(
4     aes(
5       x = rating_text_position,
6       label = rating
7     )
8   ) +
9   geom_text(
10    aes(
11      x = rating_text_position,
12      label = timing
13    )
14  ) +
15 ...
```



```
1 pre_post_data |>
2   pre_post_plot() +
3   geom_label(
4     aes(
5       x = 6,
6       label = growth_formatted
7     )
8   ) +
9   ...
```



# Highlight Strategically and Programmatically

Color

Size

Shadow

Outline

Opacity

A photograph of several crayon sticks standing upright against a dark background. The colors of the crayons visible from left to right are green, blue, red, yellow, magenta, white, brown, and orange. Overlaid on the center of the crayons is the word "Color" in a large, white, sans-serif font.

Color

## HOUSING

2022 Housing Data Profiles

HARTFORD

SINGLE-FAMILY HOMES AS PERCENT OF ALL HOMES

**19%**

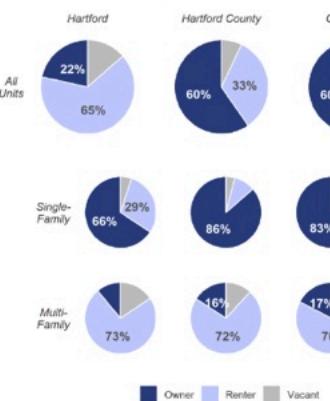


PERCENT OF ALL HOMES OCCUPIED BY OWNERS

**22%**

Overall, 65% of Connecticut's occupied housing stock is comprised of single-family housing, while 35% is multifamily housing (2+ units in structure). Most single-family homes are occupied by homeowners, while most multifamily units are occupied by renters.

In Hartford, 19% of occupied homes are single-family, and 81% are multifamily. Owners live in 66% of Hartford's 10,626 single-family homes, and renters live in 73% of its 44,561 multifamily homes.

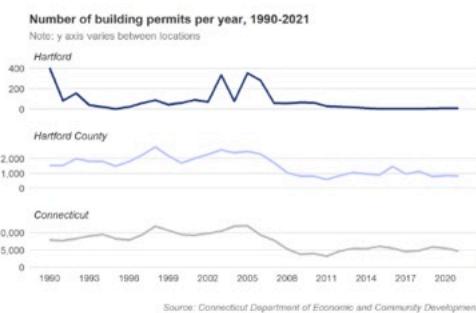


CHANGE IN BUILDING PERMITS, 1990-2021

**-97%**

Growth is slow in the state, which has seen a 41% decrease in building permits between 1990 and 2021.

In Hartford, there were 405 building permits issued in 1990, compared to 11 issued in 2021, representing a 97% decrease.



Data Profiles

25%

50%

Hartford

Hartford County



```
1 single_family_homes

# A tibble: 169 × 2
  location      pct
  <fct>        <dbl>
1 Andover     0.897
2 Ansonia     0.548
3 Ashford     0.850
4 Avon         0.844
5 Barkhamsted 0.956
6 Beacon Falls 0.717
7 Berlin       0.807
8 Bethany      0.958
9 Bethel        0.746
10 Bethlehem   0.944
# i 159 more rows
```

```
1 single_family_homes_plot <- function() {
  single_family_homes |>
    ggplot(
      aes(
        x = pct,
        y = 1
      )
    ) +
    geom_point(
      shape = 124,
      color = "grey80"
    ) +
    ...
  }
```

```
1 single_family_homes_plot()
```

0%

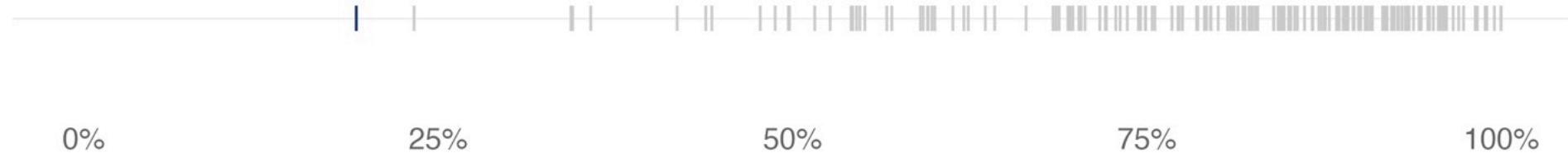
25%

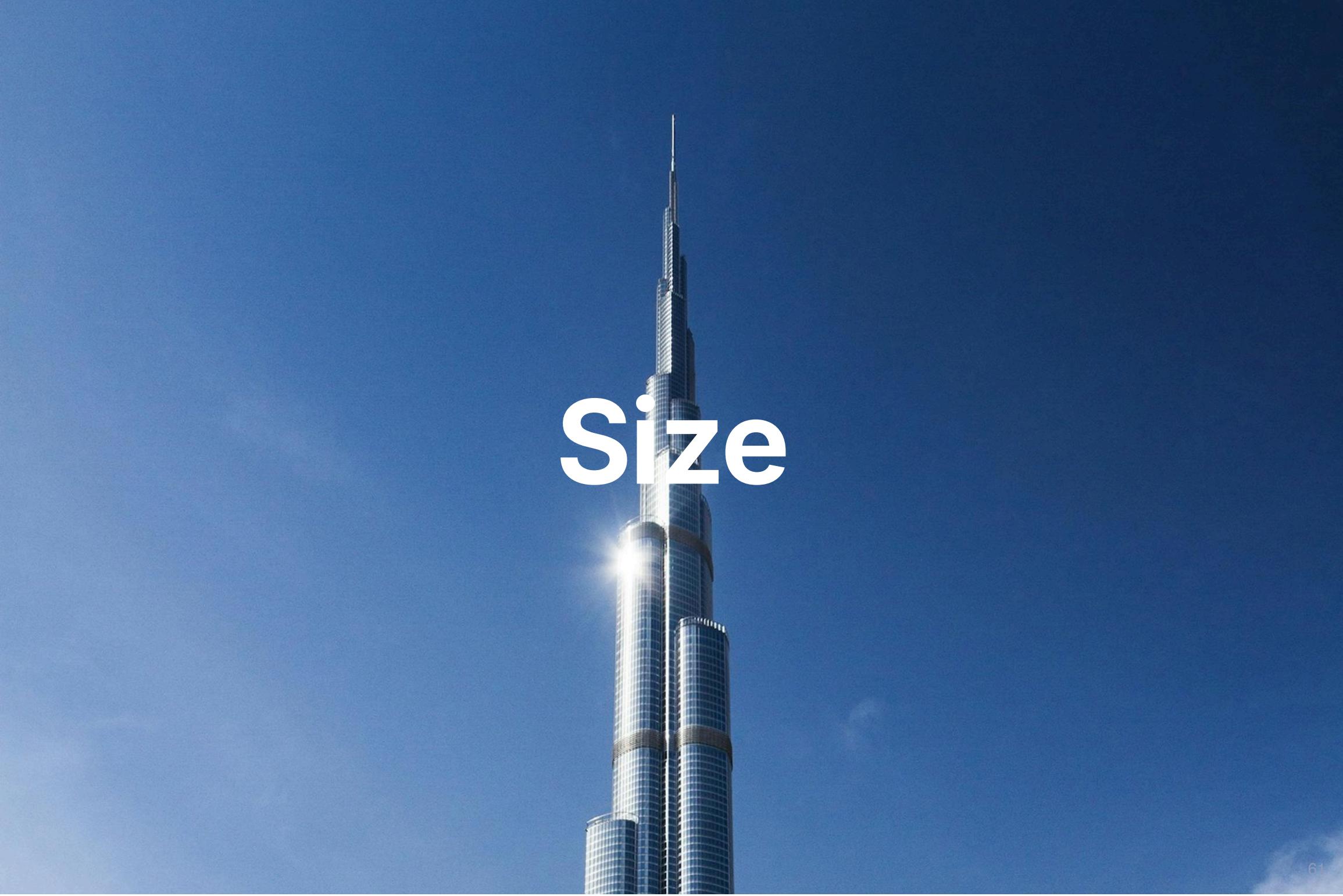
50%

75%

100%

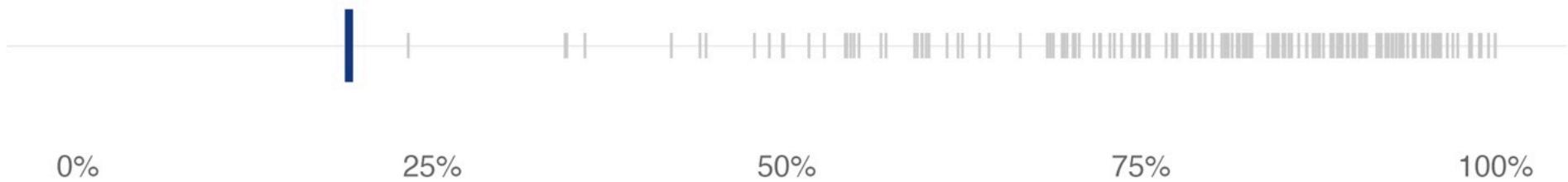
```
1 single_family_homes_plot() +  
2   geom_point(  
3     data = single_family_homes |> filter(location == "Hartford"),  
4     shape = 124,  
5     color = "#15397f"  
6   )
```



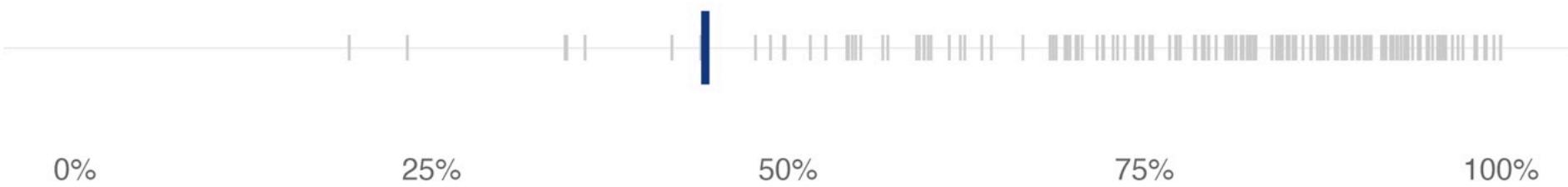
A photograph of the Burj Khalifa, the world's tallest building, set against a clear, deep blue sky. The sun is positioned directly behind the tower, creating a bright, overexposed area at the top where the spire meets the sky.

Size

```
1 single_family_homes_plot() +  
2   geom_point(  
3     data = single_family_homes |> filter(location == "Hartford"),  
4     shape = 124,  
5     color = "#15397f",  
6     size = 15  
7   )
```



```
1 single_family_homes_plot() +  
2   geom_point(  
3     data = single_family_homes |> filter(location == "Stamford") ,  
4     shape = 124 ,  
5     color = "#15397f" ,  
6     size = 15  
7   )
```



# Shadow



# AFGHANISTAN

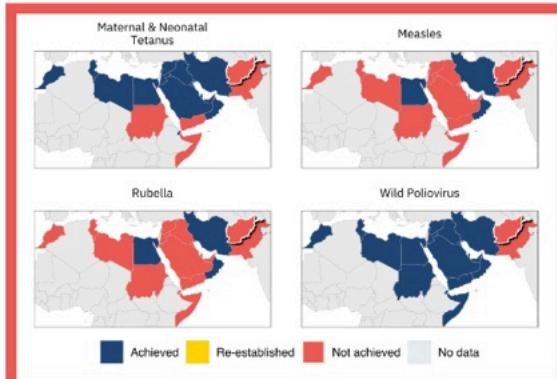
## Impact Goals



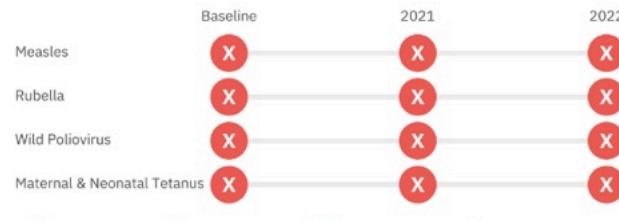
### 1.2 Control, Eliminate & Eradicate Vaccine-Preventable Disease

In 2022, measles elimination was **achieved** in 4 countries in the WHO region, while it was **re-established** in 0 countries and **not achieved** in 17 countries.

The presentation of maps is not an expression of the opinion of Immunization Agenda 2030 regarding the legal status of countries/territories, their governing authorities, or their official borders.

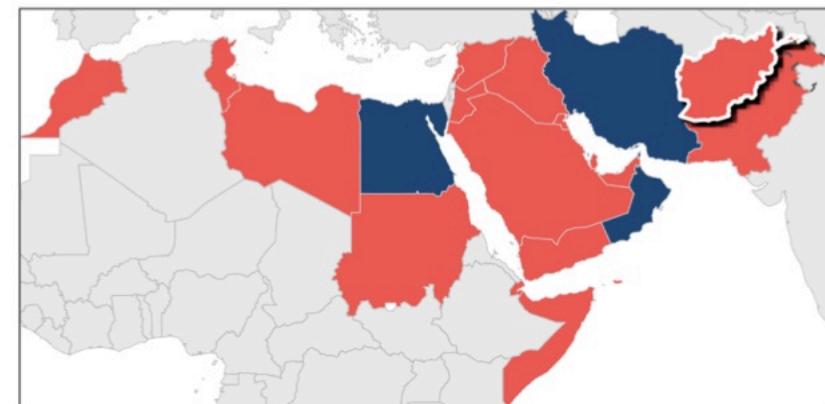


Achievement of vaccine preventable disease control, elimination, and eradication targets in Afghanistan:



Data source: Verification, certification, and disease-specific committee reports. Targets for vaccine-preventable disease are elimination of measles transmission, elimination of rubella transmission, eradication of wildtype poliovirus, and elimination as a public health problem for maternal and neonatal tetanus. The baseline year is 2020.

### Rubella



■ Achieved    ■ Re-establish

```
1 rubella
```

Simple feature collection with 69 features and 3 fields

Geometry type: MULTIPOLYGON

Dimension: XY

Bounding box: xmin: -1466635 ymin: -185219 xmax: 8649202 ymax: 4806756

Projected CRS: WGS 84 / World Mercator

First 10 features:

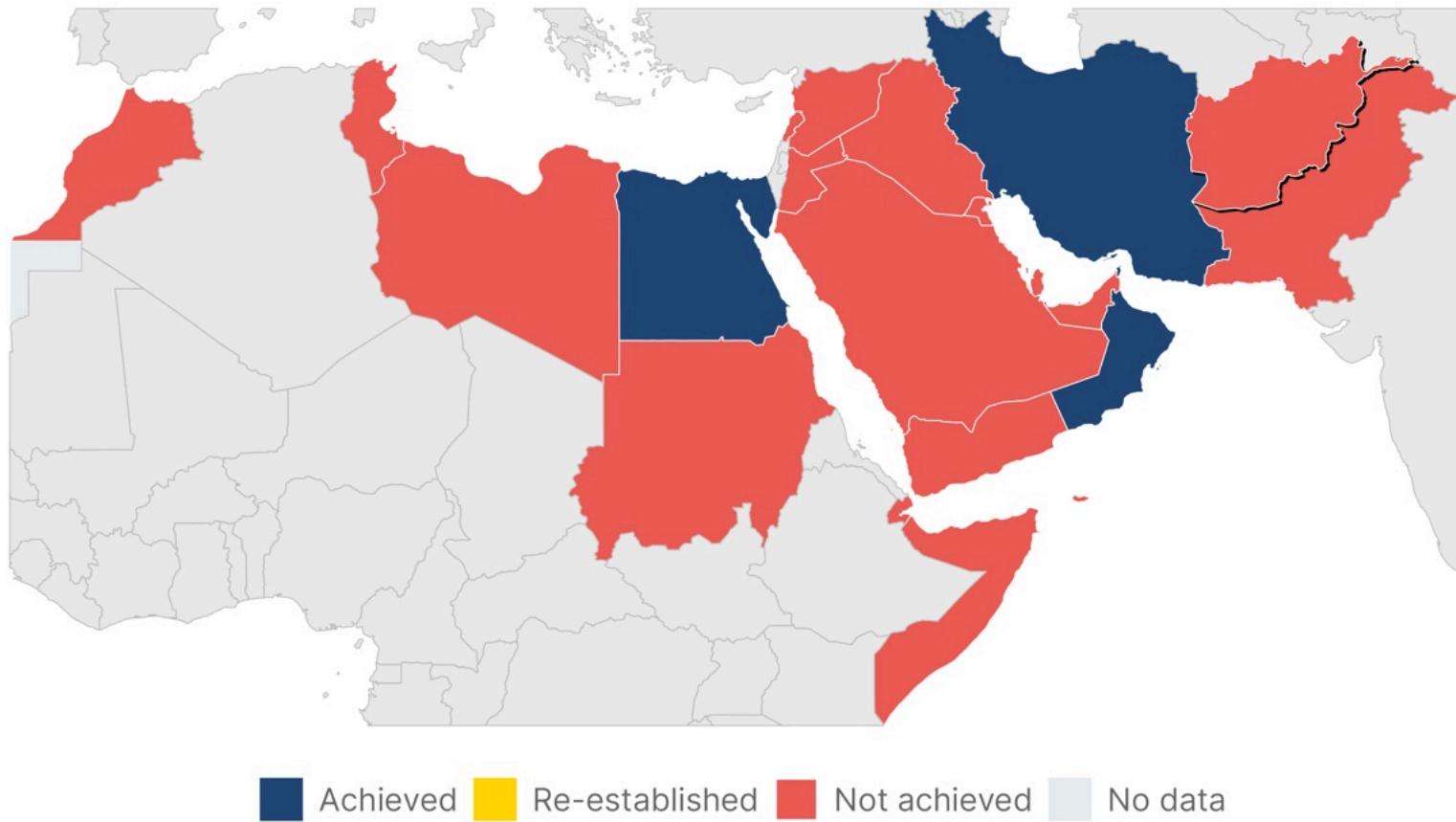
		country	region	status	geometry
1		Afghanistan		1	Not achieved MULTIPOLYGON (((8328599 444...
2	United Arab Emirates			1	Not achieved MULTIPOLYGON (((6004988 275...
3		Benin		0	Not achieved MULTIPOLYGON (((392437.9 12...
4		Burkina Faso		0	Not achieved MULTIPOLYGON (((21538.26 16...
5		Bahrain		1	Achieved MULTIPOLYGON (((5634224 299...
6	Central African Republic			0	Not achieved MULTIPOLYGON (((2701448 953...
7		China		0	Not achieved MULTIPOLYGON (((8649200 420...
8		Cameroon		0	Not achieved MULTIPOLYGON (((1719405 819...
9		Djibouti		1	Not achieved MULTIPOLYGON (((4780514 122...
10		Algeria		0	Not achieved MULTIPOLYGON (((961591.4 43...

```
1 region_map <- function() {  
2   ggplot() +  
3     geom_sf(  
4       data = rubella |> filter(region == 1),  
5       aes(fill = status))  
6     ) +  
7     geom_sf(  
8       data = rubella |> filter(region == 0),  
9       fill = "lightgrey",  
10      alpha = 0.5  
11    ) +  
12    ...  
13 }
```

```
1 region_map()
```

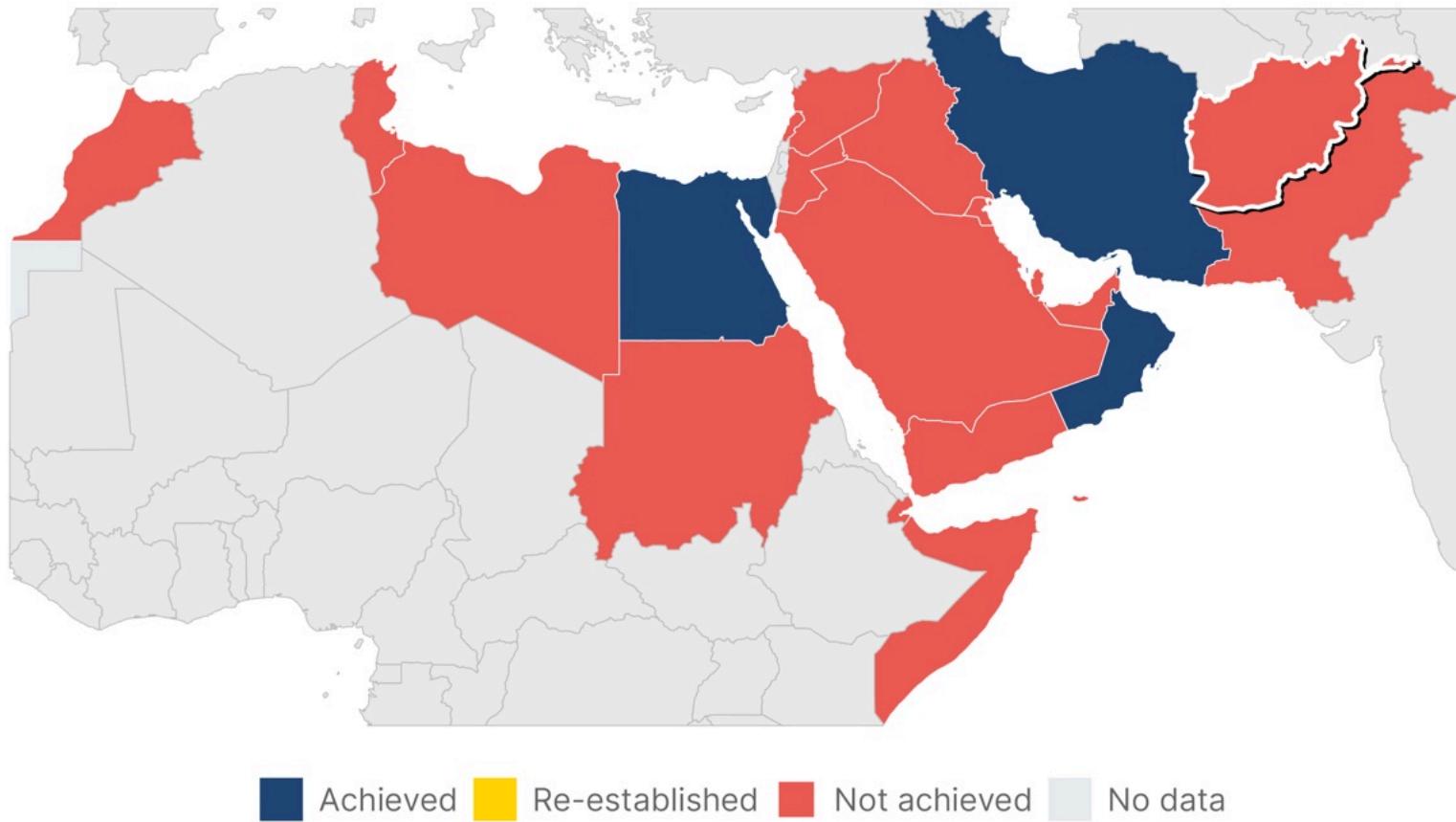


```
1 library(ggfx)
2
3 region_map() +
4   with_shadow(
5     geom_sf(
6       data = rubella |> filter(country == "Afghanistan")
7     ),
8     ...
9   )
```



# Outline

```
1 region_map() +  
2   with_shadow(  
3     geom_sf(  
4       data = rubella |> filter(country == "Afghanistan")  
5     ),  
6     linewidth = 0.8,  
7     color = "white",  
8     ...  
9   )
```



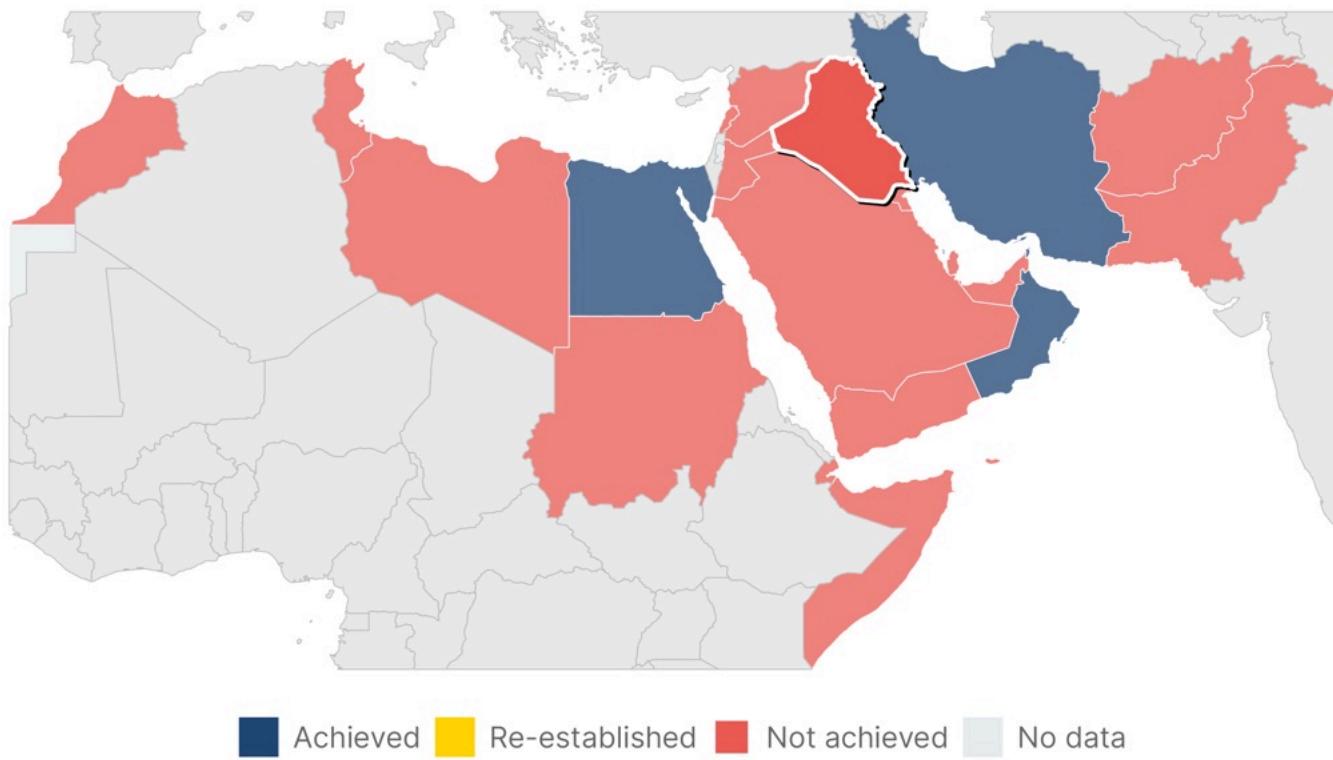
# Opacity

```
1 region_map <- function() {  
2   ggplot() +  
3     geom_sf(  
4       data = rubella |> filter(region == 1),  
5       aes(fill = status),  
6       alpha = 0.75  
7     ) +  
8     geom_sf(  
9       data = rubella |> filter(region == 0),  
10      fill = "lightgrey",  
11      alpha = 0.5  
12    ) +  
13    ...  
14 }
```

```
1 region_map() +  
2   with_shadow(  
3     geom_sf(  
4       data = rubella |> filter(country == "Afghanistan")  
5     ),  
6     linewidth = 0.8,  
7     color = "white",  
8     ...  
9   )
```



```
1 region_map() +  
2   with_shadow(  
3     geom_sf(  
4       data = rubella |> filter(country == "Iraq")  
5     ),  
6     linewidth = 0.8,  
7     color = "white",  
8     ...  
9   )
```



# Conclusion

1. Consider the outer limits of your data
2. Minimize text and position it carefully
3. Highlight strategically and programmatically with a range of aesthetic properties
4. Packages are helpers, but you are the one who has to do the thinking