

## Plant Resilient USA

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# Miami needs trees.

And so does everywhere else.

#### Motivation

- In 20 years, the Miami climate will not be suitable for many plants native to Miami that are planted at this moment.
- For Miami citizens, trees and shades are crucial to their activities around the city, providing them necessary coolness in extreme hot weather.
- However, these trees may soon be gone as climate change alters the hardiness zones they normally grow in.



We present -

## Climate-Ready Planting: Interactive Map of Future Hardiness Zones

Plant Native, Plant Resilient

#### Collecting Baseline Data

These are the current hardiness zones in USA as of 2023.

0	<pre>df = prepare_data() df</pre>							
<del>_</del>		zipcode	zone	trange	zonetitle	t_low	t_high	
	0	501	7b	5 to 10	7b: 5 to 10	5	10	
	1	544	7b	5 to 10	7b: 5 to 10	5	10	
	2	1001	6b	-5 to 0	6b: -5 to 0	-5	0	
	3	1002	6a	-10 to -5	6a: -10 to -5	-10	-5	
	4	1003	6a	-10 to -5	6a: -10 to -5	-10	-5	
	39916	99363	7b	5 to 10	7b: 5 to 10	5	10	
	39917	99371	7a	0 to 5	7a: 0 to 5	0	5	
	39918	99401	7a	0 to 5	7a: 0 to 5	0	5	
	39919	99402	7a	0 to 5	7a: 0 to 5	0	5	
	39920	99403	7b	5 to 10	7b: 5 to 10	5	10	
	39921 ro	ws × 6 colu	mns					





Quite a lot of data!
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#### Extract Southeastern USA Data

```
[33] def extract zipcodes():
      # read in USPS zipcode database
      zipcodes = pd.read csv('/content/drive/My Drive/hardiness-zone-data/zip code database.csv')
      # Southeastern States: North Carolina, South Carolina, Tennessee, Mississippi, Alabama, Georgia, Florida
      # State abbreviations: NC. SC. TN. MS. AL. GA. FL
      # Extract southeatern US zipcodes
      southeastern_zipcodes = zipcodes[zipcodes['state'].isin(['NC', 'SC', 'TN', 'MS', 'AL', 'GA', 'FL'])]['zip']
      return southeastern zipcodes
[34] def extract original zones(df):
      df.drop(columns=['trange', 't low', 't high'])
      df.to csv('/content/drive/My Drive/hardiness-zone-data/original zones.csv', index=False)
[35] southeastern zipcodes = extract zipcodes()
     # Keep only southeastern states in the dataframe
    southeastern_df = df[df['zipcode'].isin(southeastern_zipcodes)]
     extract original zones(southeastern df)
     print(len(southeastern df))
     southeastern_df.head()
             Still plenty data!
     10190
              27006
                          5 to 10
     10191
                      7b 5 to 10
                                             10
              27007
     10192
              27009
                      8a 10 to 15
                                             15
     10193
              27010
                           5 to 10
                                             10
     10194
              27011
                           5 to 10
                                             10
```

The scale of our project covers the 7 Southeastern states

The project is easily scalable to continental US, Alaska, Hawaii, and Puerto Rico, given the available USDA data.



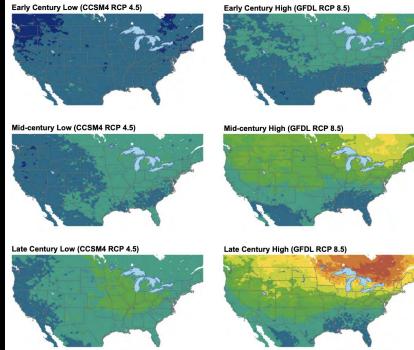
#### Collecte Climate Forecast

USDA reported projected climate change in early-, mid-, and late-century.



	Midwest			Northeast			Southeast		
Change	Early Century	Mid- century	Late Century	Early Century	Mid- century	Late Century	Early Century	Mid- century	Late Century
Min Temp (Low)									
-3	0	0	0	0	0	0	0	0	0
-2	0	0	0	0	0	0	0	0	0
-1	0	0	0	8.0	0	0	0	0	0
0	7.0	0	0	79.3	0	0	46.5	0	0
1	77.1	0	0	19.9	0.2	0	53.2	4.3	0
2	16.0	13.3	0.2	0	55.7	7.8	0.3	77.7	10.8
3	0	75.4	5.0	0	41.4	54.5	0	18.0	55.5
4	0	10.9	60.8	0	2.6	34.4	0	0	30.7
5	0	0.4	34.0	0	0	3.3	0	0	3.1
6	0	0	0.1	0	0	0	0	0	0
7	0	0	0	0	0	0	0	0	0
8	0	0	0	0	0	0	0	0	0
9	0	0	0	0	0	0	0	0	0

Source: Assessing Potential Climate Change Pressures across the Conterminous United States: Mapping Plant Hardiness Zones, Heat Zones, Growing Degree Days, and Cumulative Drought Severity throughout this Century Early: 2010–2039. Mid: 2040–2069. Late: 2070–2099.





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### Update Temperatures By Sampling



	temp_change	early_century	mid_century	late_century
0	0	0.403	0.009	0.001
1	1	0.428	0.328	0.168
2	2	0.136	0.434	0.427
3	3	0.033	0.162	0.199
4	4	0.000	0.058	0.112
5	5	0.000	0.009	0.069
6	6	0.000	0.000	0.023
7	7	0.000	0.000	0.010

13.6% of land in Southeastern USA will experience a 2°C temperature rise by early-century (~2039)

5.8% of Southeastern USA will experience a 4°C temperature rise by mid-century (2040~2069)



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#### Sampling Algorithm

```
Oregon State
University
```

```
def sample zipcodes(southeastern zipcodes, time period, mode);
      if mode == 'low':
        proportions = list(temp changes low df[time period])
      elif mode == 'high':
        proportions = list(temp changes high df[time period])
      sampled zipcodes = []
      for prop in proportions:
        sampled_zipcodes.append(southeastern_zipcodes.sample(frac=prop, random_state = 14, replace=False))
      return sampled_zipcodes
[46] def update temperatures(df. sampled zipcodes):
      df['new t low'] = df['t low']
      df['new_t_high'] = df['t_high']
      temp change = 0
      for zipcode list in sampled zipcodes:
        # 1 degree celcius increase = 1.8 degree F increase
        df.loc[df['zipcode'].isin(zipcode_list), 'new_t_low'] += temp_change * 1.8
        df.loc[df['zipcode'].isin(zipcode_list), 'new t high'] += temp_change * 1.8
        temp_change += 1
      return df
```

We sampled Southeastern zip codes according to the temperature change distribution given by USDA.

Since the USDA Plant
Hardiness Zone Map
(PHZM) dataset is in
Fahrenheit, but their
climate report is in Celsius,
a conversion is necessary
when calculating
temperature changes.



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#### Forecasting Future Hardiness Zones

```
def update hardiness zones(df):
  df['new trange'] = 'null'
  for index. row in df.iterrows():
    left zone = None
    right zone = None
    for zone in hardiness zones:
      left bound, right bound = zone
      if left_bound <= row['new_t_low'] <= right_bound:</pre>
        left zone = zone
      if left bound <= row['new t high'] <= right bound:</pre>
        right zone = zone
      if left zone and right zone:
        if left zone == right zone:
          southeastern_df.at[index, 'new_trange'] = right_zone
          break
    if left_zone and right_zone and left_zone != right_zone:
        left diff = abs(row['new t low'] - left zone[1])
        right diff = abs(row['new_t_high'] - right_zone[0])
        if left diff > right diff:
          df.at[index, 'new trange'] = left zone
        else:
          df.at[index, 'new_trange'] = right_zone
  return df
```

For each new temperature range, we used a simple variant of the sliding window technique to determine where it falls in the hardiness zones.



#### Not good!

zipcode	zone	trange	new_t_low	new_t_high	new_trange	new_zone
27006	7b	5 to 10	15.8	20.8	(15, 20)	8b
27007	7b	5 to 10	8.6	13.6	(10, 15)	8a
27009	8a	10 to 15	13.6	18.6	(15, 20)	8b
27010	7b	5 to 10	8.6	13.6	(10, 15)	8a
27011	7b	5 to 10	5.0	10.0	(5, 10)	7b

### With these data, here's our solution -



#### **Our Solution**

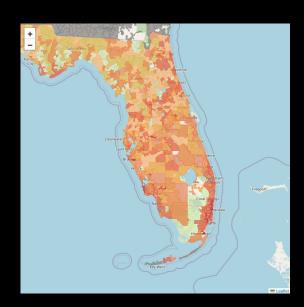
CCSM4 RCP 4.5: Low



Early-century low (2010 - 2039)



Mid-century low (2040 – 2069)



Late-century low (2070 - 2099)



#### **Our Solution**

GFDL RCP 8.5: High







Early-century high (2010 – 2039)

Mid-century high (2040 – 2069)

Late-century high (2070 - 2099)



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#### Future Directions: Connect Plant Database

With our current scalable interactive map, we plan to connect our data to a plant database that details the plants native to a region and their favorable hardiness zones.



With our scalable interactive map,

gardeners and city planners can plant smarter

for a changing climate native, resilient, and built to last.