

# North Carolina Senior Games Participation Analysis

Created by Thomas Grover

---

## Three Motivating Questions:

1. What trends can we see in pickleball participation? Is the sport thriving or declining?
  2. What is the demographic of pickleball players? How old are most players?
- 

## Pickleball Trends Analysis:

The first step in the analysis was subsetting the participation data supplied by North Carolina Senior Games. We needed just records in which EventType was Pickleball. I created a grouping variable to separate the years into “Pre-Covid” and “Post-Covid” categories. This is useful in analysis, and makes logical sense given the real world implications caused by covid and quarantining period.

### Frequencies of Pickleball Participation By Year

(Separated by Pre/Post-Covid Groups)

| <i>Pre Post Covid</i> | <b>Event Year</b> | <b>Count</b> |
|-----------------------|-------------------|--------------|
| <i>Pre-Covid</i>      | 2016              | <b>466</b>   |
|                       | 2017              | <b>467</b>   |
|                       | 2018              | <b>552</b>   |
|                       | 2019              | <b>563</b>   |
| <i>Post-Covid</i>     | 2022              | <b>501</b>   |
|                       | 2023              | <b>719</b>   |
|                       | 2024              | <b>864</b>   |

Count represents the number of pickleball participants for the respective year. We can see the more recent years have higher frequencies. This may lead one to assume that pickleball is clearly trending upwards. But what does the data say?

To determine if there were any statistically significant differences in pickleball participation over the years, we conduct a T-test. Our target parameter is essentially the difference between the average participation before and after covid. (We can think of this as a small sample hypothesis test for a target parameter  $\Theta = \mu_1 - \mu_2$ ).

**It is important to note that this method assumes our observations come from a normal distribution.**

**Table 1: Summary Statistics of the Number of Pickleball Participants per Year by Era**

| Era        | Method        | N | Mean  | Std Dev | Std Err | Minimum | Maximum |
|------------|---------------|---|-------|---------|---------|---------|---------|
| Post-Covid |               | 3 | 694.7 | 182.7   | 105.5   | 501.0   | 884.0   |
| Pre-Covid  |               | 4 | 512.0 | 52.7320 | 26.3660 | 466.0   | 563.0   |
| Diff (1-2) | Pooled        |   | 182.7 | 122.8   | 93.6129 |         |         |
| Diff (1-2) | Satterthwaite |   | 182.7 |         | 108.7   |         |         |

**Table 2: T-Test Results**

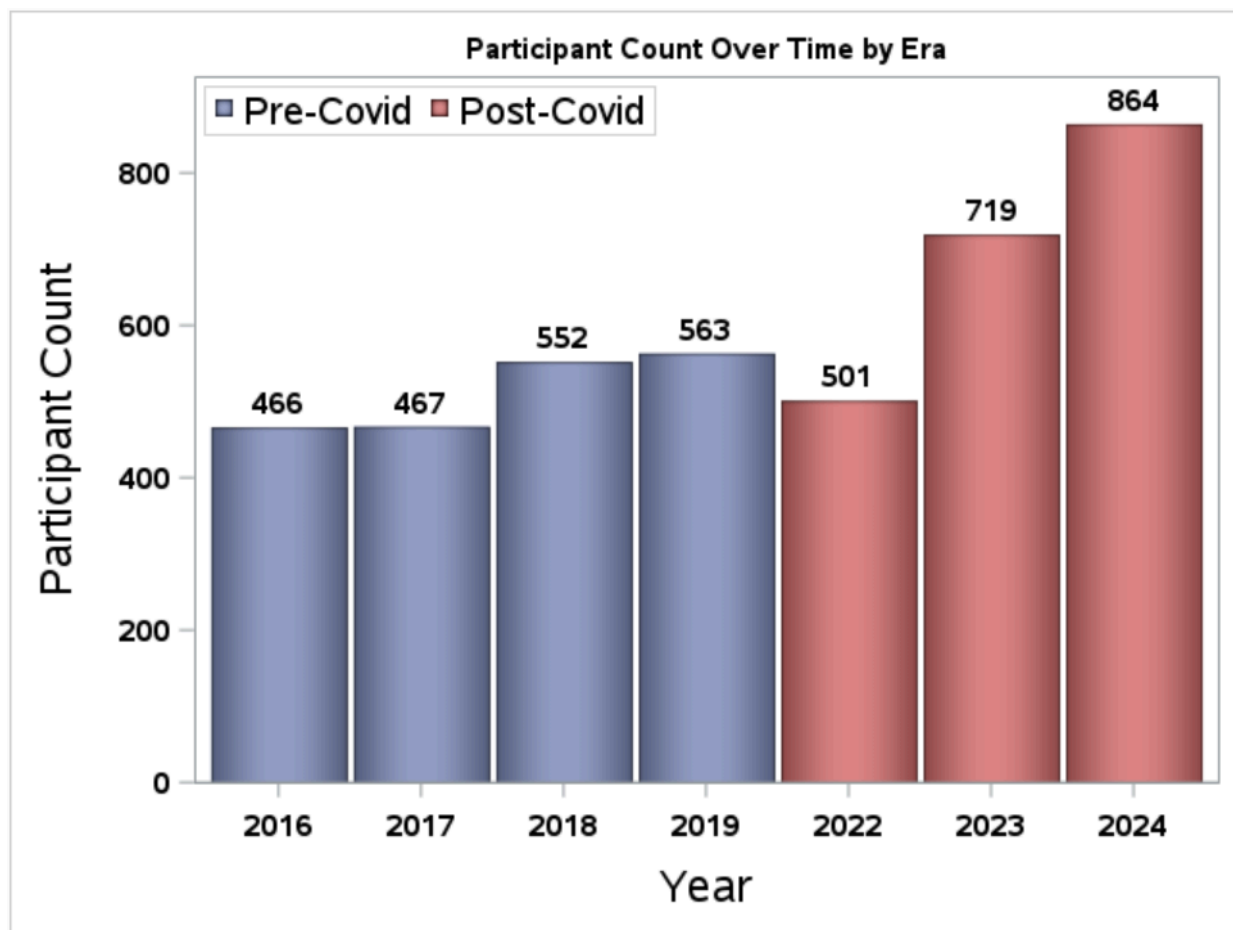
| Method        | Variances | DF     | t Value | Pr >  t |
|---------------|-----------|--------|---------|---------|
| Pooled        | Equal     | 5      | 1.95    | 0.1085  |
| Satterthwaite | Unequal   | 2.2518 | 1.68    | 0.2209  |

| Equality of Variances |        |        |         |        |
|-----------------------|--------|--------|---------|--------|
| Method                | Num DF | Den DF | F Value | Pr > F |
| Folded F              | 2      | 3      | 12.01   | 0.0740 |

From the analysis in SAS, we can see that with a pooled variance test we get a p-value of  $p=0.1085$ , and when we don't assume equal variances we get a p-value of  $p=0.2209$ . We

can also see the p-value for equality of variances  $p=0.0740$ , meaning according to the data it is extremely unlikely that the two eras have the same variances.

We need to consider this in the context of the real world though, and especially with how limited our data is (test between samples of 3 and 4 observations respectively). I would argue that assuming equal variances actually makes sense here (same populations, same organizations data, same event), but clearly based on the very low p-value when testing for equality of variances, the data doesn't support this. So what can we say about the trends of pickleball participation? A visual will support our insights.



This graph shows the frequencies we calculated earlier in a visual manner, with Era's separated by color. The uptrend is visually significant, but our data doesn't provide clear statistical evidence of a difference between pre and post-covid years. However, in my

best judgement with all of this information taken into account, and the understanding of national and global pickleball trends (growing rapidly), *I would tell NCSG that pickleball is one of their events trending upwards, and that they should market towards it more and invest more resources into it.*

---

### **Pickleball Demographic Analysis:**

The next goal was gaining insight into the demographic of pickleball players. The first steps of analyzing Age involved getting basic summary statistics about the distribution of the variable. First, acquiring the basic measures of **center and variability**. And second, by taking a look at **quantile data** and **extreme observations** to get a better understanding of the context of the data.

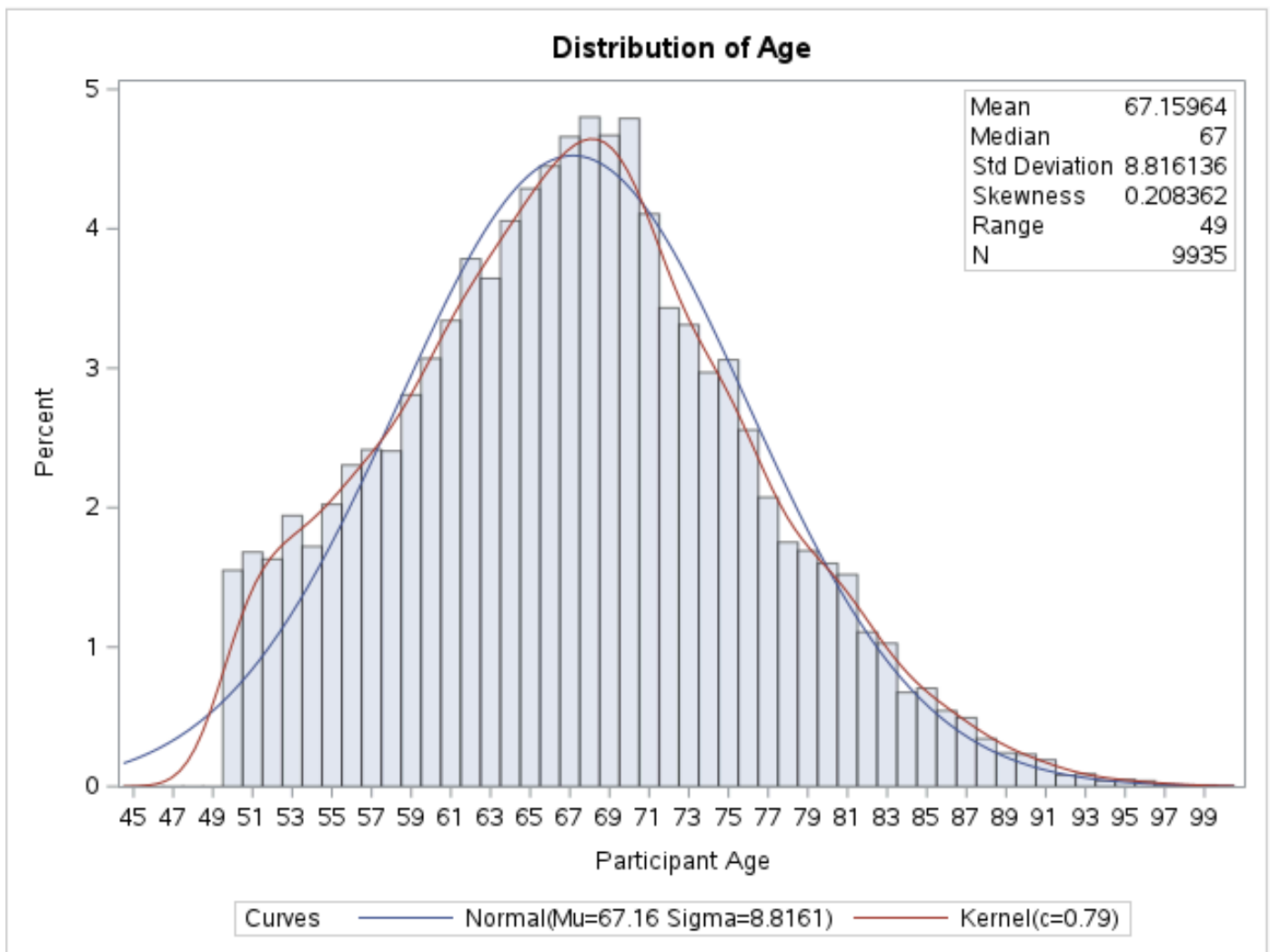
| Basic Statistical Measures |          |                     |          |
|----------------------------|----------|---------------------|----------|
| Location                   |          | Variability         |          |
| Mean                       | 67.15964 | Std Deviation       | 8.81614  |
| Median                     | 67.00000 | Variance            | 77.72426 |
| Mode                       | 68.00000 | Range               | 49.00000 |
|                            |          | Interquartile Range | 12.00000 |

| Quantiles (Definition 5) |          |
|--------------------------|----------|
| Level                    | Quantile |
| 100% Max                 | 99       |
| 99%                      | 88       |
| 95%                      | 82       |
| 90%                      | 79       |
| 75% Q3                   | 73       |
| 50% Median               | 67       |
| 25% Q1                   | 61       |
| 10%                      | 55       |
| 5%                       | 53       |
| 1%                       | 50       |
| 0% Min                   | 50       |

| Extreme Observations |      |         |      |
|----------------------|------|---------|------|
| Lowest               |      | Highest |      |
| Value                | Obs  | Value   | Obs  |
| 50                   | 9883 | 96      | 6864 |
| 50                   | 9877 | 96      | 9019 |
| 50                   | 9838 | 97      | 5426 |
| 50                   | 9818 | 98      | 7301 |
| 50                   | 9815 | 99      | 7170 |

The next step I took in this was creating a visual representation of the overall distribution of players, to get a feel for who was actually playing pickleball. Here we have a histogram of the distribution of age, where **each bar represents a single age**. The Y-Axis represents the percentage of all pickleball players that specific age makes up. We also overlay a normal distribution with the **same mean and standard deviation we observed**, as well as a kernel distribution, for comparison.

Notice the **skewness value** in the legend, indicating we have **right skewed data with its positive value**.



So what can we conclude about our demographic?

1. All measures of center indicate that the middle of our **age distribution is somewhere between 67-68 years of age.**
  2. Our **standard deviation of 8.816** indicates that about **68% of total pickleball players are 58-76 years old.**
  3. When we take into account the minimum age of 50 for participation, it appears that given our large sample, the age of participants converges in distribution to a Normal distribution with mean  $\mu = 67.16$  and variance  $= 8.816^2$ .  
(Let  $Y_i$  = age of participant,  $Y_i \sim N(67.16, 8.816^2)$  approximately)
-