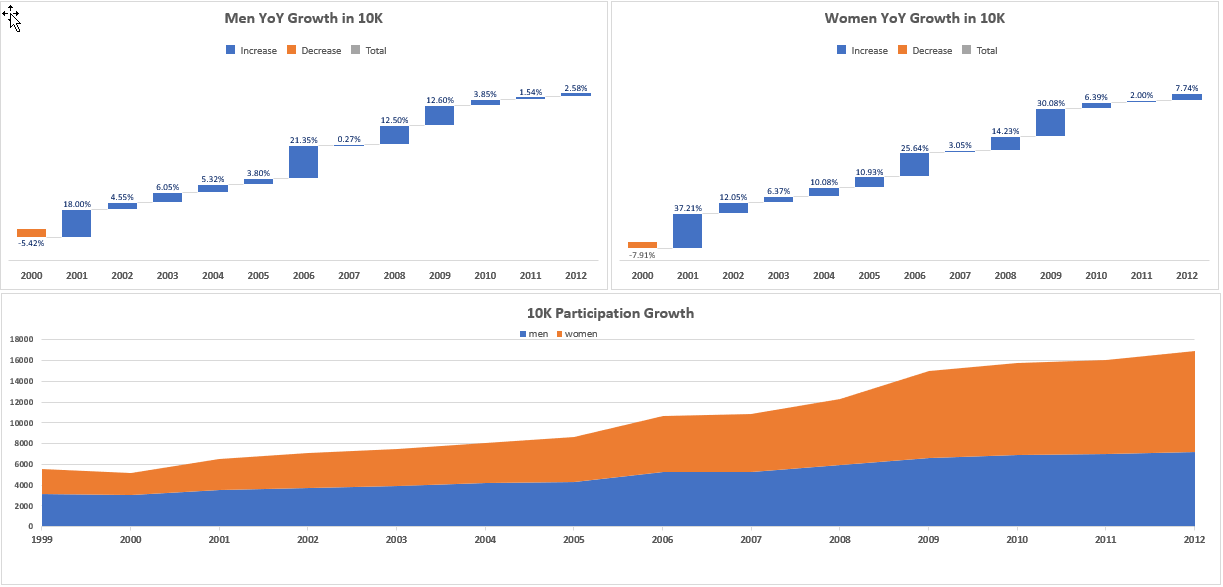
**Case 2: Cherry Blossom 10 Miler**

**Introduction and Overview**

The Cherry Blossom 10 Mile race in DC is one of the most popular races in the area, and it is part of the Professional Road Running Organization race circuit. In the years from 1999 to 2012, the race has seen growth in the number of both female and male runners. While there was a slight dip in participation between 2000 and 2001, the year over year growth has been as high as 25-30%, as seen in **Figure 1**. Overall, the Cherry Blossom 10 miler has seen a greater increase in women runners (314% from 1999 to 2012), compared to a smaller increase of male runners (126% over the same time period). The bottom graph of **Figure 1** shows a larger proportion of women over the years, and by 2012, the race had growth to nearly 17,000 participants.

**Figure 1: Year Over Year Participation by Gender**

The annual event has also grown in complexity, with the seeding of elite runners for guaranteed entry, fundraising opportunities to guarantee a race spot for non-seeded runners, a team entry system and competition, and a lottery system to award entry to additional runners. In 2012, the slowest race times were well past two and a half hours, plus the many hours needed for set up and tear down. The city of Washington DC has put pressure on Cherry Blossom race organizers to limit the time of the race. Given the other Cherry Blossom activities happening around the same time, it has become more difficult to keep roads closed that length of time. Therefore, the race needs to take less time in the years ahead.

The race committee has asked our team to assess the past 13 years of race results to understand how the race has grown, how the age distribution of runners has changed, and how the pace of runners has changed. This information will become the foundation of changes to the Cherry Blossom 10 Mile race.

**Data Preparation (Q7)**

The team collected data by web scrapping information from the Cherry Blossoms 10 Miler website. The results from 1999 to 2012 for both female and male runners were obtained by accessing each individual year’s website in the directory found at: <http://www.cherryblossom.org/results>. We found that several years of data were stored in slightly different formats (i.e. wider page headers, variable spaces between lines of data) and had different names for the same attribute (i.e. time vs net time vs gun time to represent the total race time for each runner).

*Data Scraping and Parsing*

Our team designed and implemented a function that accommodated these format variations to effectively scrape the data from the website. The comprehensive methods we employed can be found in the respective R code and Python notebooks. See **Code Directory** at the conclusion of this write up. The following are highlights of issues encountered by the team.

Women’s Data

* For most years, the data was fairly consistent but there were anomalies that required additional specialized treatment in some years.
* For the 1999 data, the **“//pre”** node worked to obtain data, while for 2000, the **“//font”** node was needed. Furthermore, in 1999, the linefeed carriage return character **“\r”** was missing, and the newline character **“\n”** was used instead.
* Additionally, the header line **“====”** was not present in all years, making it rather difficult to find the header and the spacer columns. This required manual intervention to establish the variable lengths in 2001 and 2002.
* Various functions were used to trim leading and trailing whitespaces and replace extraneous characters and spaces.

Men’s Data

* Similar issues existed for the men’s data as the women’s data with some exceptions.
* The biggest challenge in scraping the men’s data was for year 2009. It required additional work to effectively scrape the data and then to structure, clean, and transform as needed.
* Lastly, 2009 data was formatted in a Word-like document on the website, where the end of each data line was indicated by **</pre><pre>.** This required a separate function to scrape, parse, and clean the data.

*Data Cleaning*

After compiling all the data, our initial review revealed additional inconsistencies and missing data. The team invested time to find missing ages by cross referencing against the “Searchable Results” page on the website; information was updated or marked as missing (NA) for more than 85 records. We also specifically reviewed times for participants under the age of 12; while times for some of these young runners may be questionable, the information published was taken at face value, with the assumption that there were too few to significantly impact the analysis.

Additional data cleaning and preparation included:

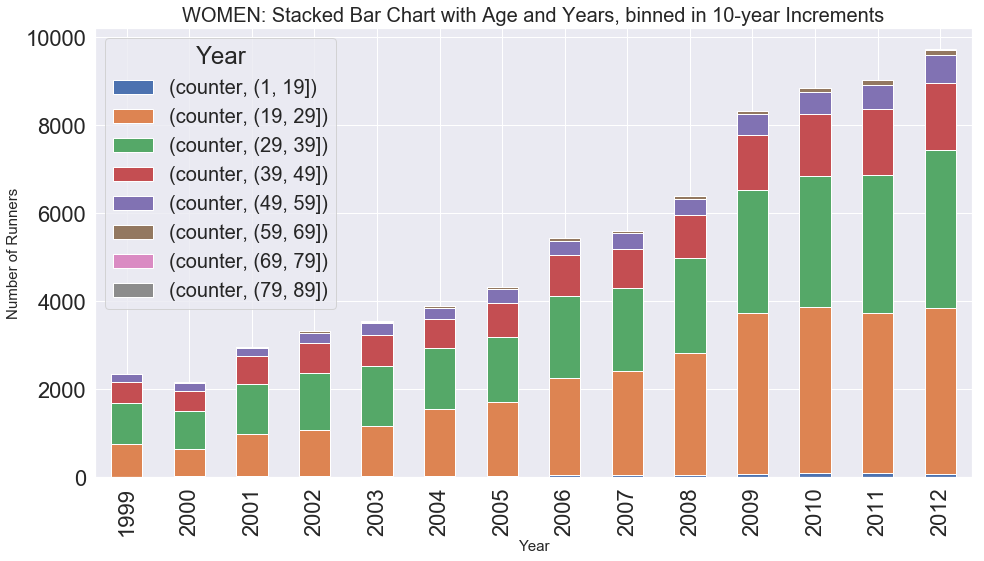
* Runners without race time were dropped from the data set.
* Race time was presented as “Time,” “Net Time,” or “Gun Time,” depending on race year. For consistency in analysis, we used “Net Time” when available, followed by “Gun Time” or “Time.” Details of these definitions can be found in **Appendix Figure A**, our data dictionary.
* Participations were binned into 10-year age brackets (with the exception of under 19 and above 80) to assist in understanding age trends.
* We created variables to analyze race accurately, given the variation in hours, minutes, seconds, and milliseconds of the time variable. Race pace was calculated mathematically (as the Cherry Blossom website had errors).

**Analysis**

With the objective to understand how participants’ age and race time have changed over the years, the team first looked at women and men separately.

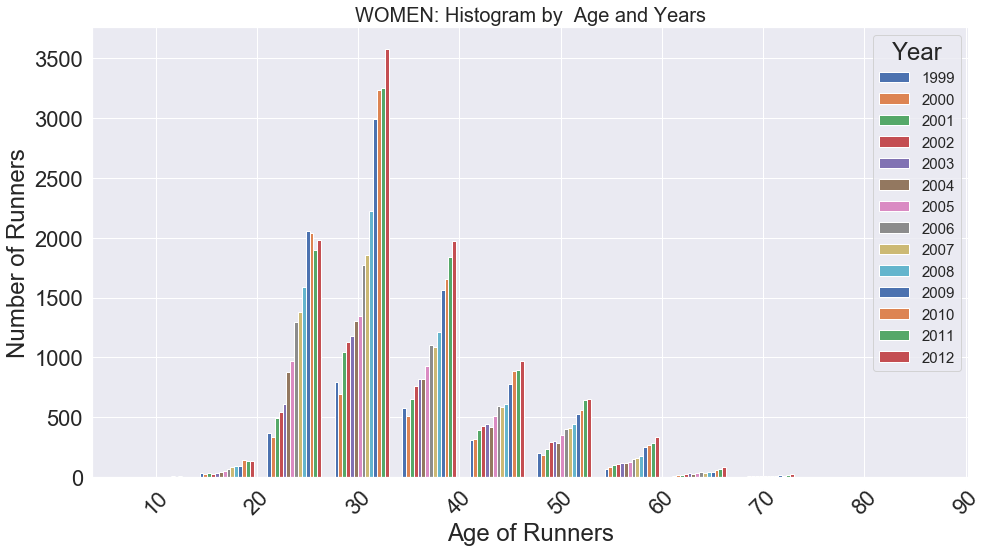
*WOMEN*

The first assessment by the team was to look at how the women in each of the 10-year age bins changed over time. As seen in **Figure 2**, the growth of women participants is most notable in the 19-29 and 29-39 categories age groups, while the other age categories grew more proportionally. Starting in 2009, this 20-year age span reflected a significant percentage of female participants (nearly 78% in 2009).

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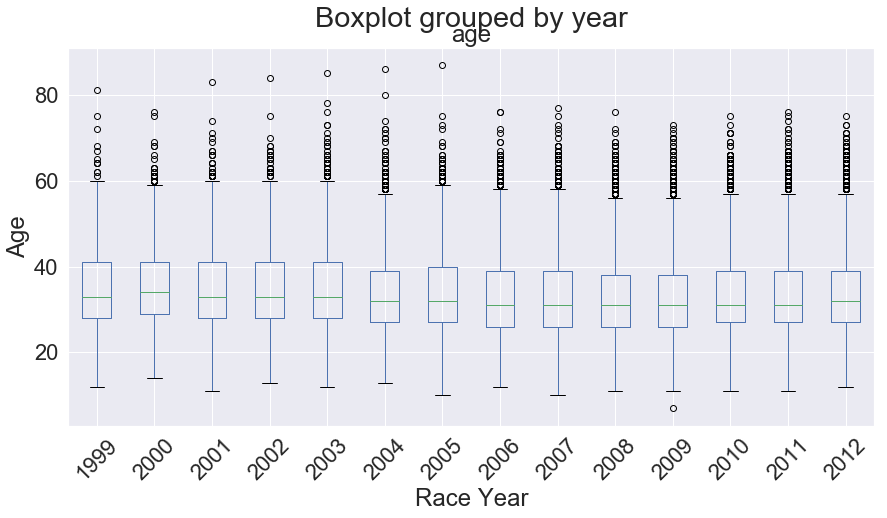
**Figure 2: 10-Year Age Bins by Year**

We looked at the same information presented another way: the number of Runners Group by Age for each year. **Figure 3** illustrates the significant increase in runners roughly age 20-40.



**Figure 3: Runner Count by Age Groups Each Year**

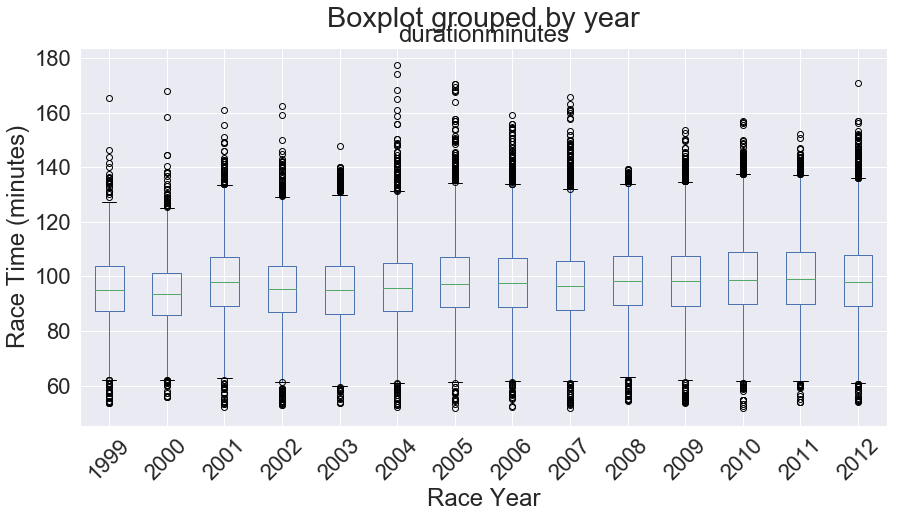
Then, we looked at two boxplots that compared women’s age by year (**Figure 4**) and race time by year (**Figure 5**). The boxplots provided better insights into the trends for age and time. The red dotted line is equal to the median age in 2000, which saw the highest average of the oldest runners. The median age decreases slightly year over year, with a noticeable drop in 2006. There were no late 70 or 80+ aged runners in 2006, which also might contribute to a change in average age. Overall, there are not dramatic shifts in the distribution of the plots but certainly a younger trend at all quartiles.

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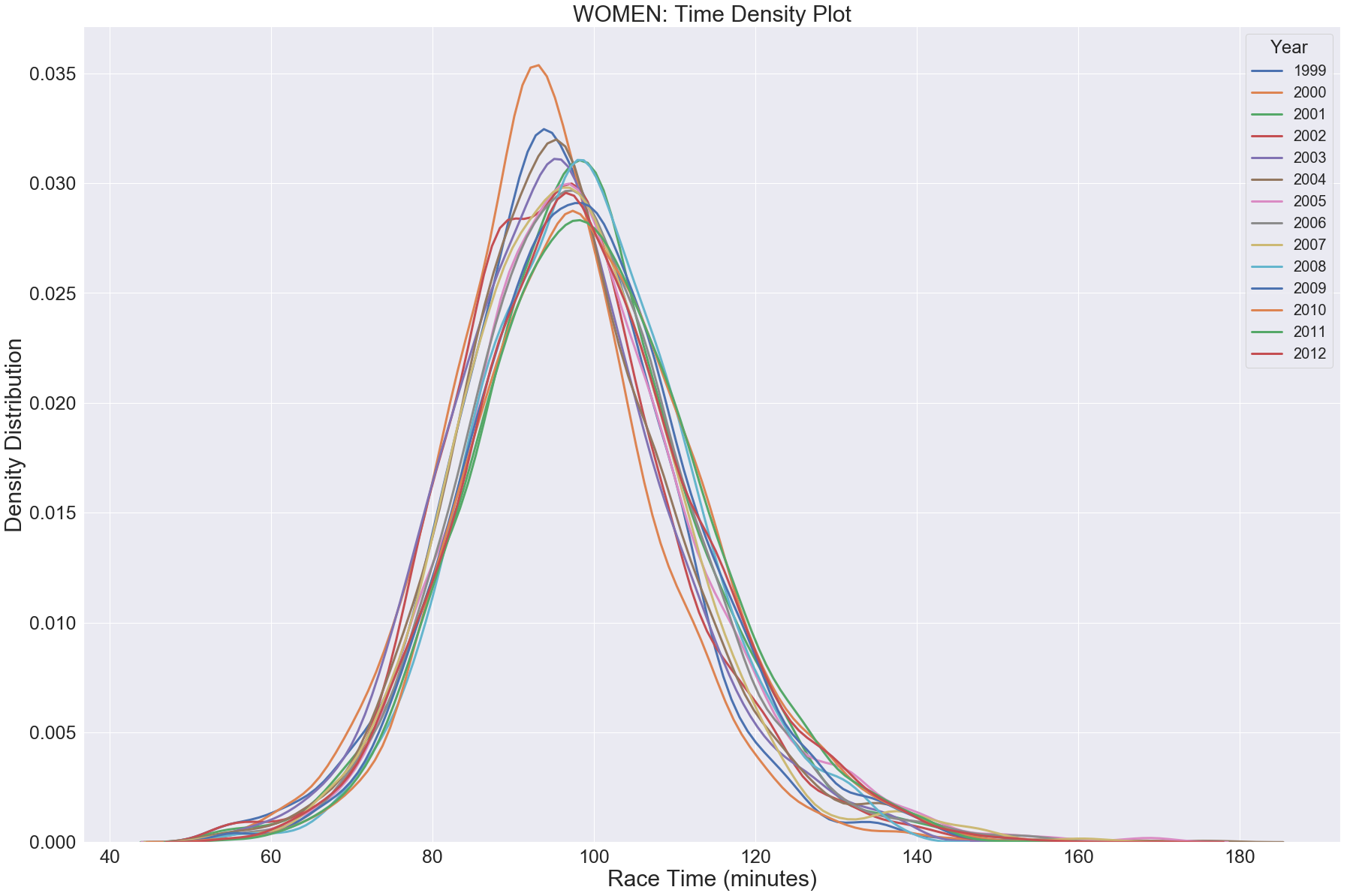
**Figure 4: Boxplot of Year and Age**

The second boxplot the team analyzed was a comparison of race time in minutes over the years to see if changes are occurring with the speed of female runners. The red dotted line reflects the median 2001 – a slow race from the early years – against the average race times for all years. As the years progress, the average race time does continue to lengthen by one or two minutes; upper quartiles show even larger variations in time. It also does not appear that the women are getting faster over the years, as seen in **Figure 5**.

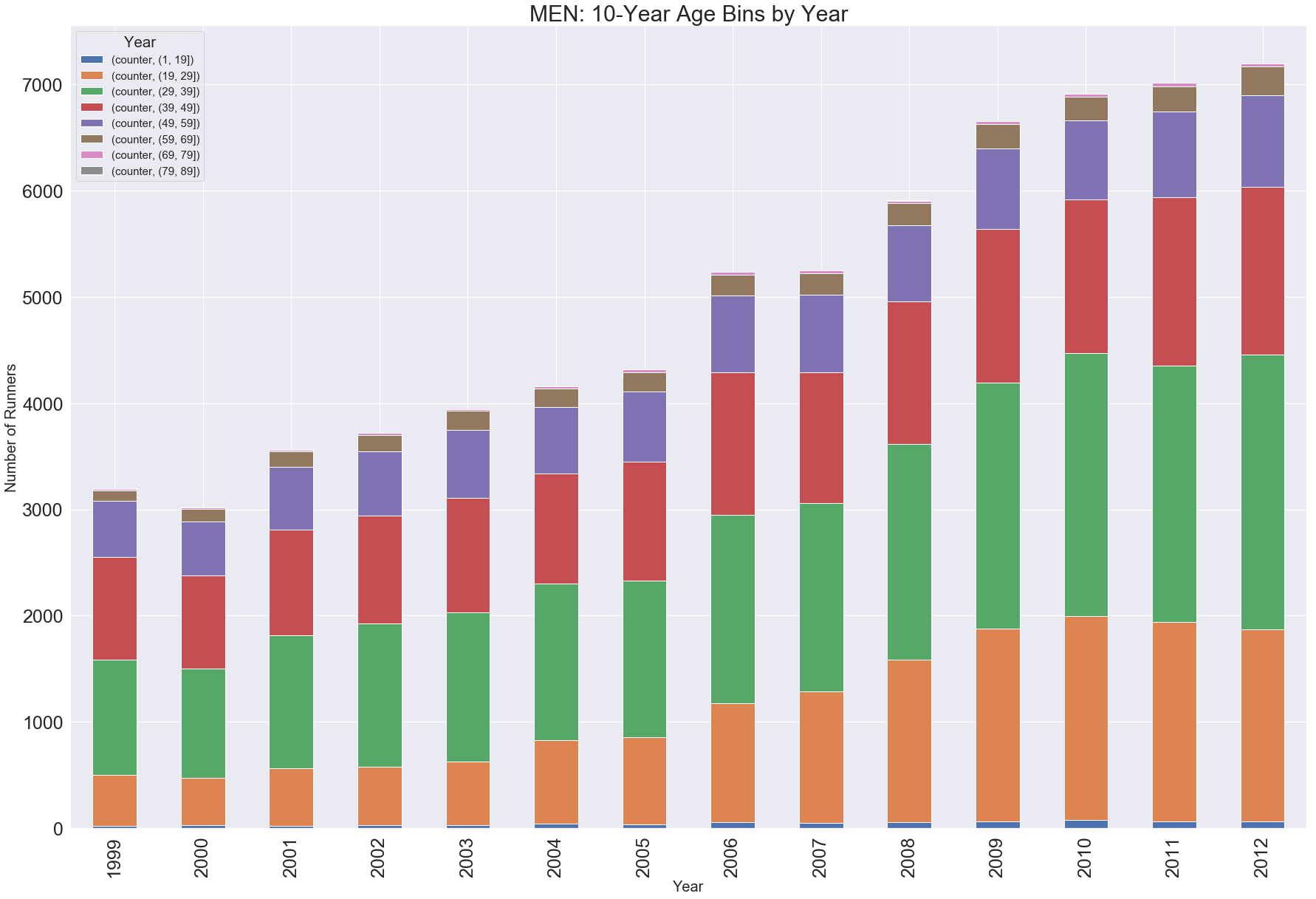
To assess speed year over year, the team also looked at a density plot of female race times. **Figure 6** shows a better comparison of the distribution of race times over the years. While 2000 does show a marked difference from the others – it was the fastest year in the data set – the other years are roughly consistent when the distributions are plotted in this way.



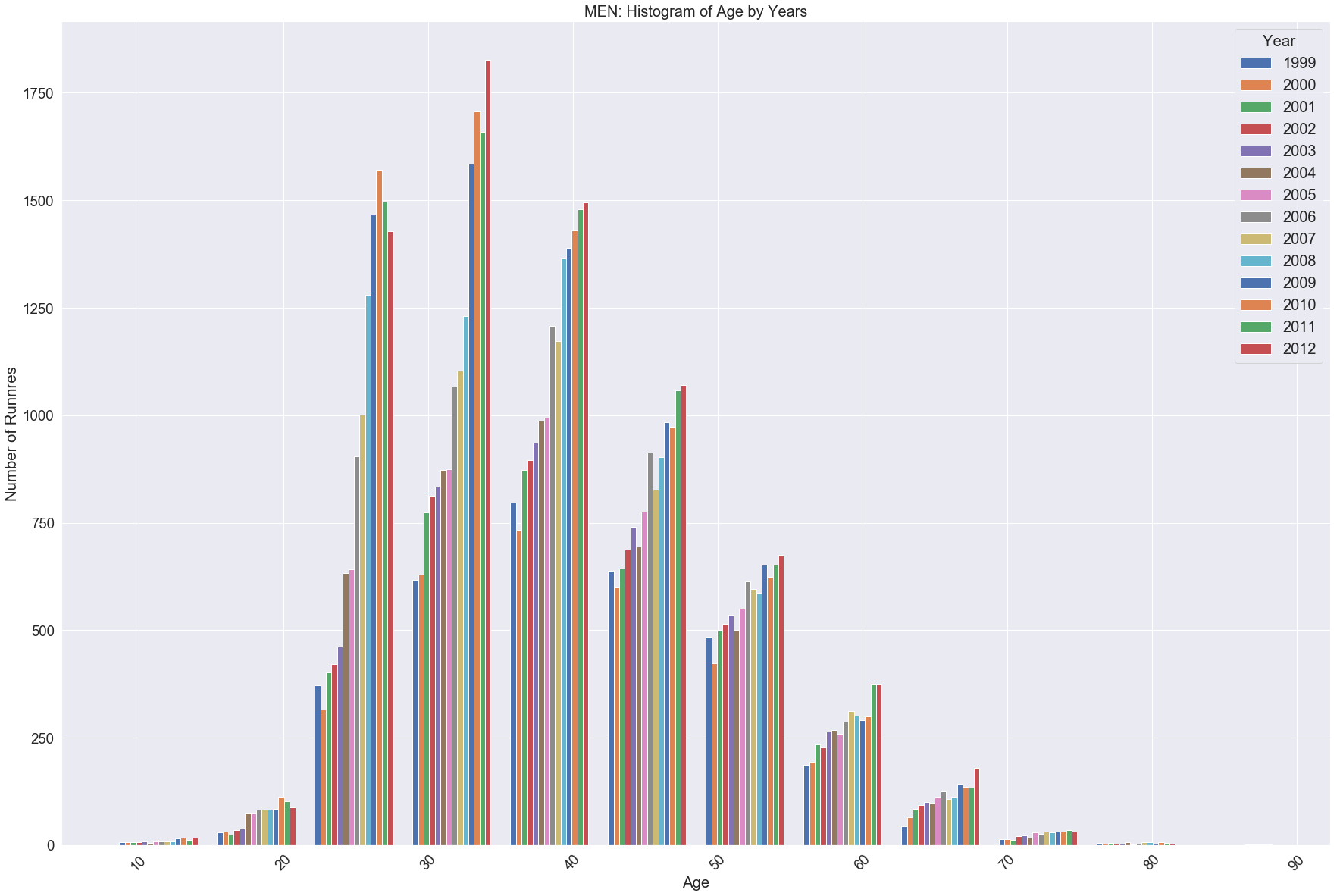
**Figure 5: Boxplot of Year and Race Time**

**Figure 6: Density Plot of Time**

*MEN*

****In parallel to the analysis of age and time for female runners, the team looked at the same set charts for men. The first assessment was to look at how the men in each of the 10-year age bins changed over time. As seen in **Figure 7**, the growth of men participants is most prominent in the 19-29 and 29-39 age groups, but the changes are not nearly as dramatic as the female changes. The proportions of age categories over 40 as a reflection of the whole population remain relatively consistent.

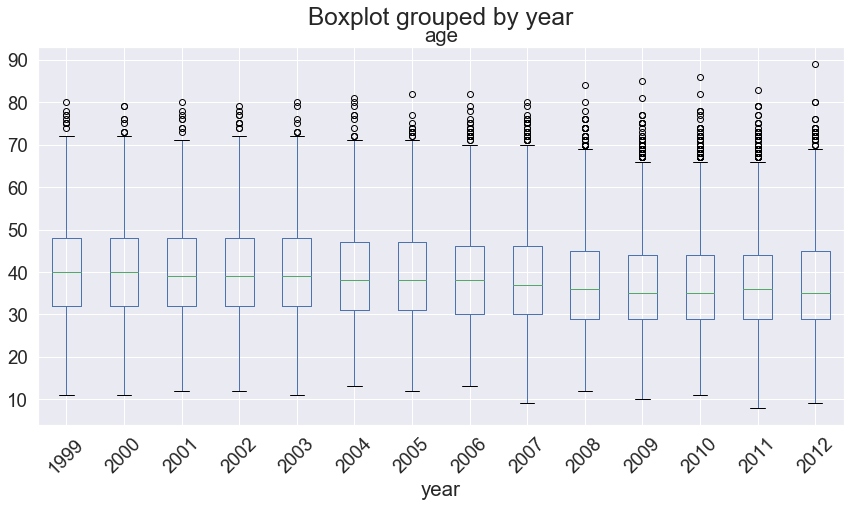
**Figure 7: 10-Year Age Bins by Year**

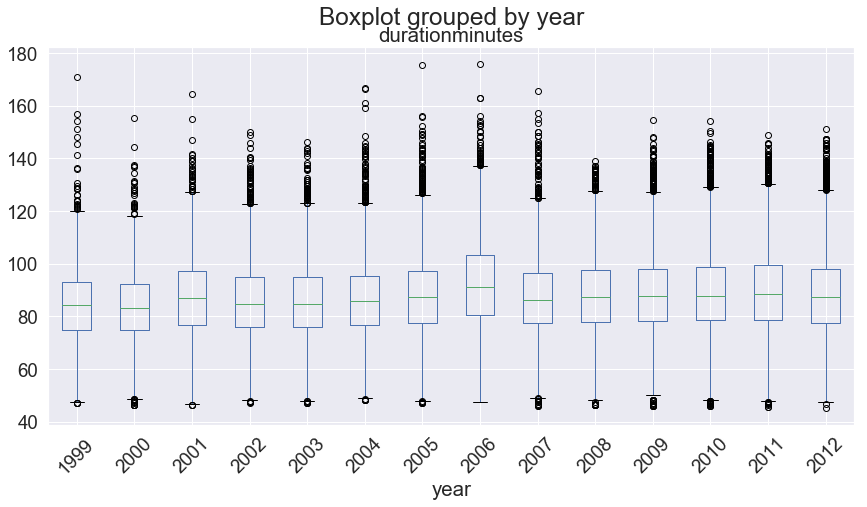
**Figure 8** highlights the growth in runners. Comparing each bar to those around it reveals the changes from year to year. The biggest peaks are in runners age 20-40, especially in recent years.

**Figure 8: Runner Count by Age Groups Each Year**

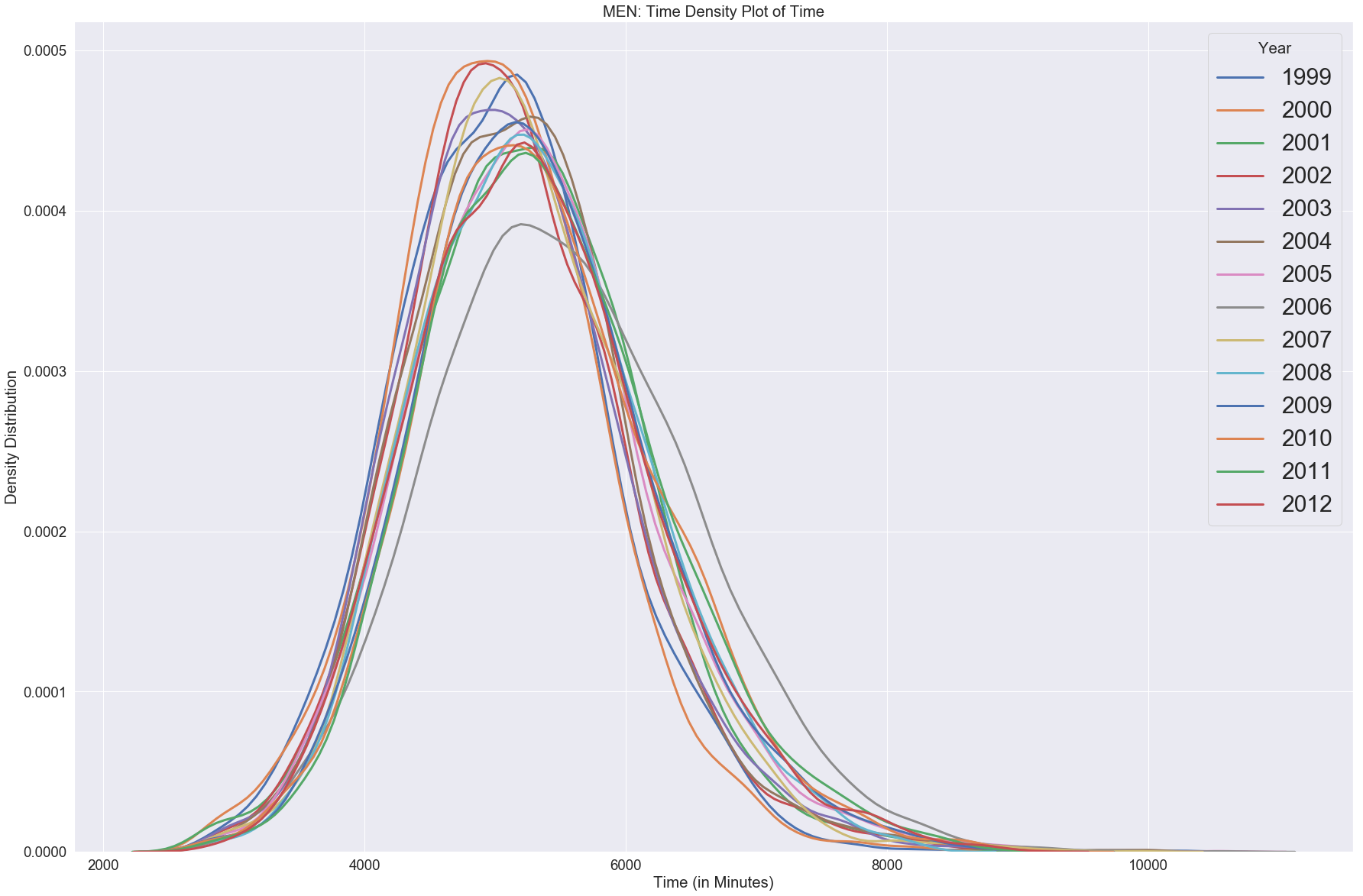
The two boxplots (**Figure 9** and **Figure 10**) compared men’s age by year and race time by year. The boxplots provided better insights into the trends for age and time. In **Figure 9**, the red dotted line is equal to the median age in 1999 and 2000, which saw the oldest average runner age. The median age decreases slightly year over year. Overall, there are not dramatic shifts in the distribution of the plots but certainly a younger trend at all quartiles.

**Figure 10** analyses race time in minutes over the years to see if changes are occurring in the race time of male runners. The red dotted line reflects the median of the slowest year’s race – 2006. It also does not appear that the men are getting faster over the years. While 2004 to 2007 had some slow outliers, the general trend overall from **Figure 11** reflects roughly consistent pace among male runners. As before, the distribution of race time is mostly consistent. However, 2006 appears to be an odd year; this year’s curve is slightly flatter and shifted to the right, indicating a slower distribution of time.

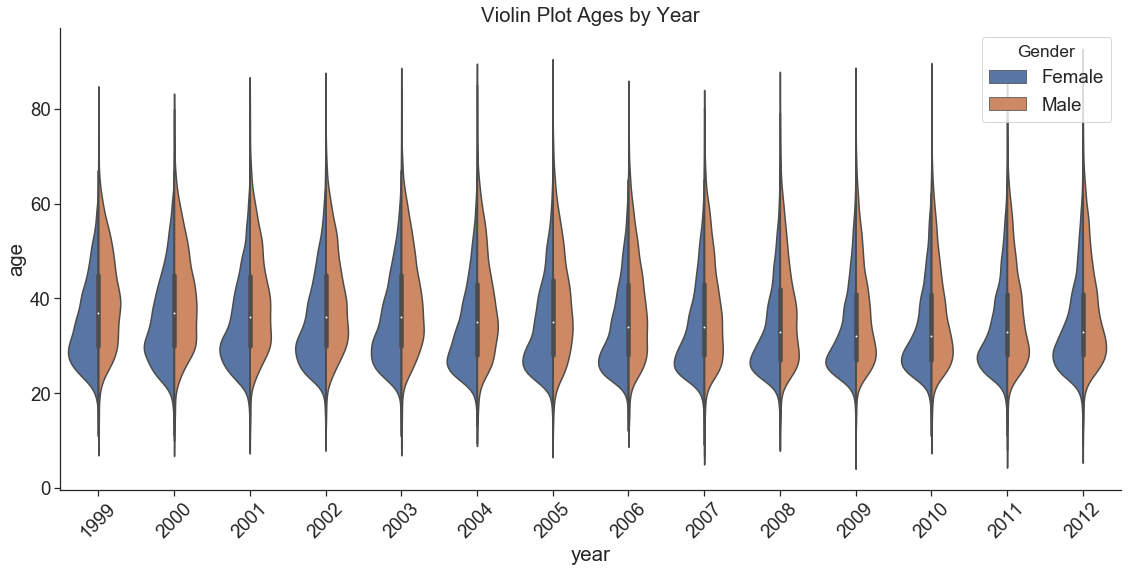
**Figure 9: Boxplot of Year and Age**

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**Figure 10: Boxplot of Year and Race Time**

**Figure 11: Density Plot of Time**

*COMPARISON OF WOMEN AND MEN*

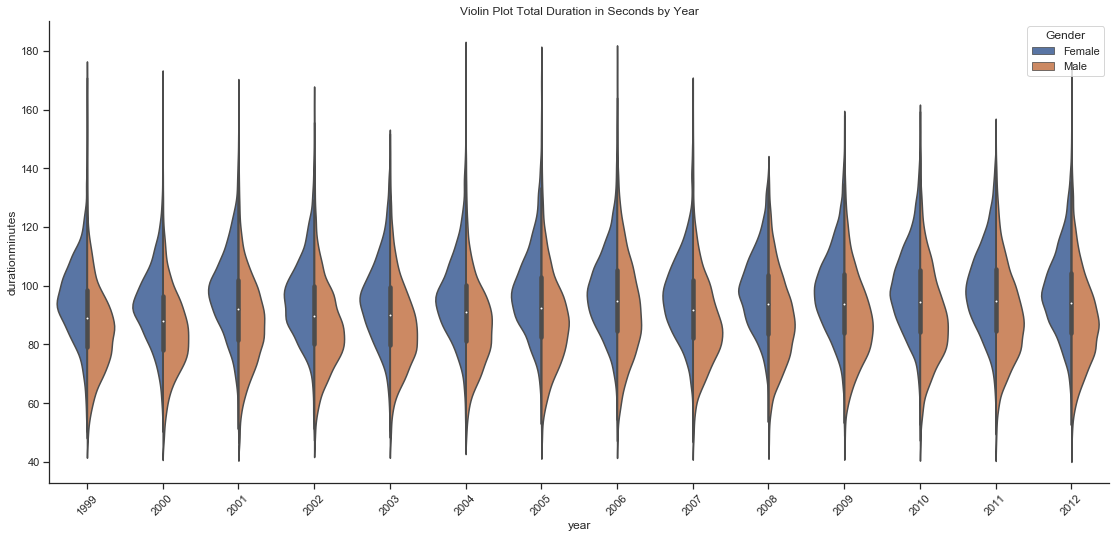
To compare men and women, we created violin plots help to demonstrate the differences. The age distribution of men compared to woman appears to be wider throughout the years, indicating a larger age spread. The distribution for women is skewed younger than men. More recent trends show that the female age distribution is less skewed in later years, as compared to earlier years. This skewness indicates that competition for the slots in the 30s age group for females could be high, given the increase in size of that age group.

**Figure 12: Violin Plot of Age**

*COMPARISON OF WOMEN AND MEN*

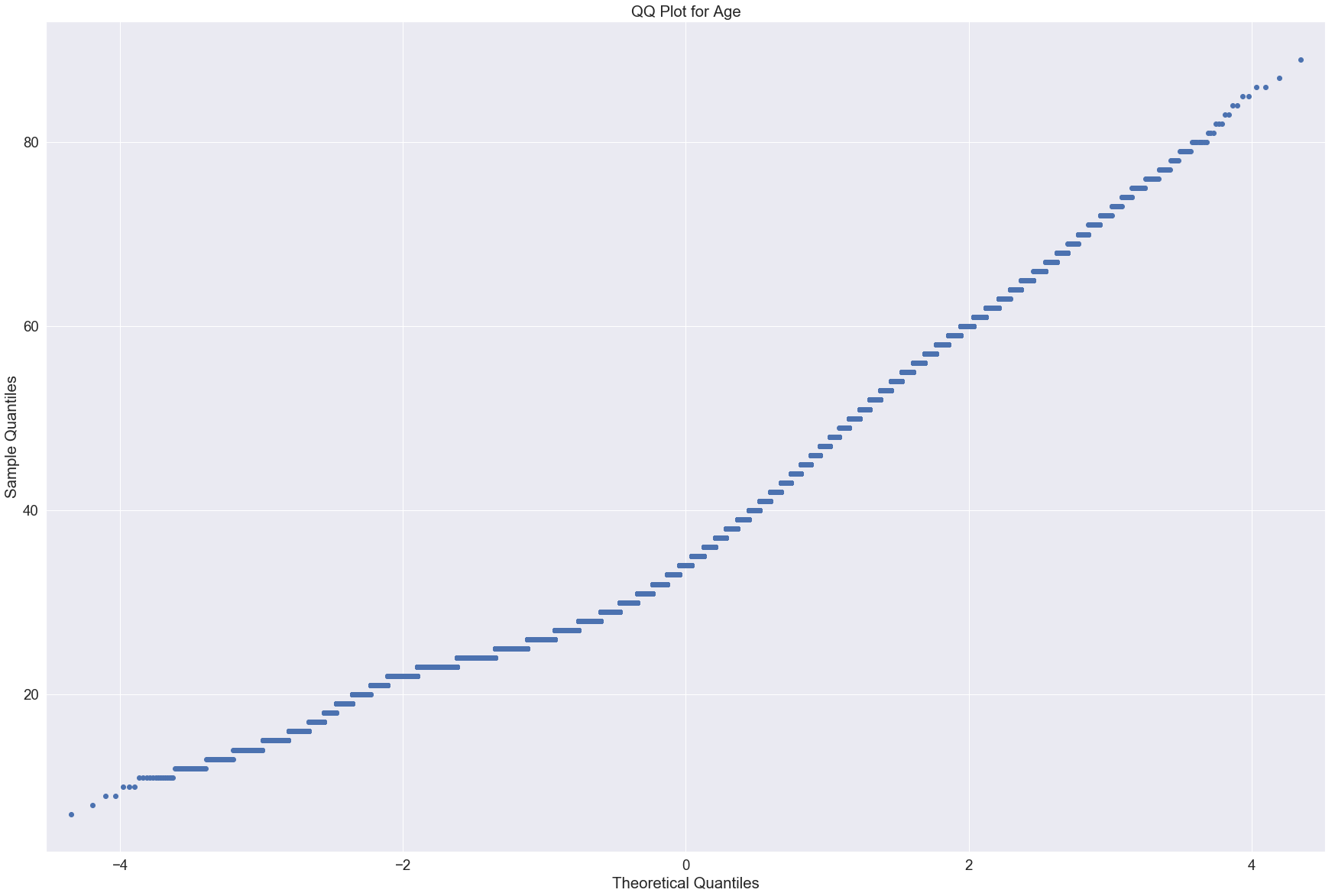
To compare men and women race times (in secs), we created violin plots to help demonstrate the differences. The race time distribution of men compared to woman appears to be wider throughout the years, indicating a larger race time spread. The distribution for the women displays less spread in race times. The race times appear to be fairly normal for both men and women. It is also evident from the plot that the men have a lower race completion time than women.

Reinforce the finds earlier, demonstrating generally consistent times over the years

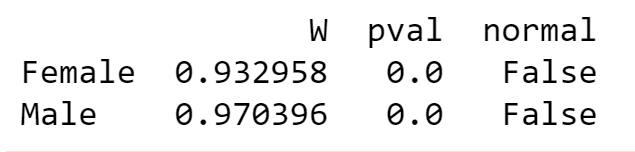


**Figure 13: Violin Plot of Time**

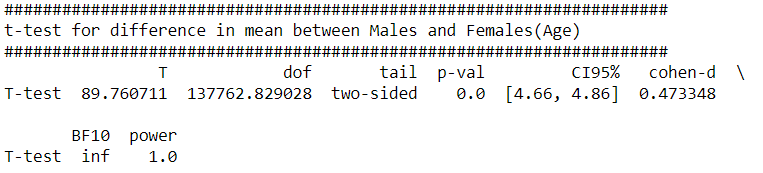
*Statistical Analysis*

We also conducted statistical analysis to understand if the age difference between women and men is significant. First, we ran a preliminary QQ Plot for both age and time for the entire data sets to visually check for normality. A QQ Plot illustrates the distribution of the data against a normal distribution; a straight line at 45 degrees indicates the data is normally distributed. **Figure 14** is a QQ Plot for age data and shows the data does not exhibit normality. This will help us select appropriate statistical tests.

**Figure 14: QQ Plot for Age**

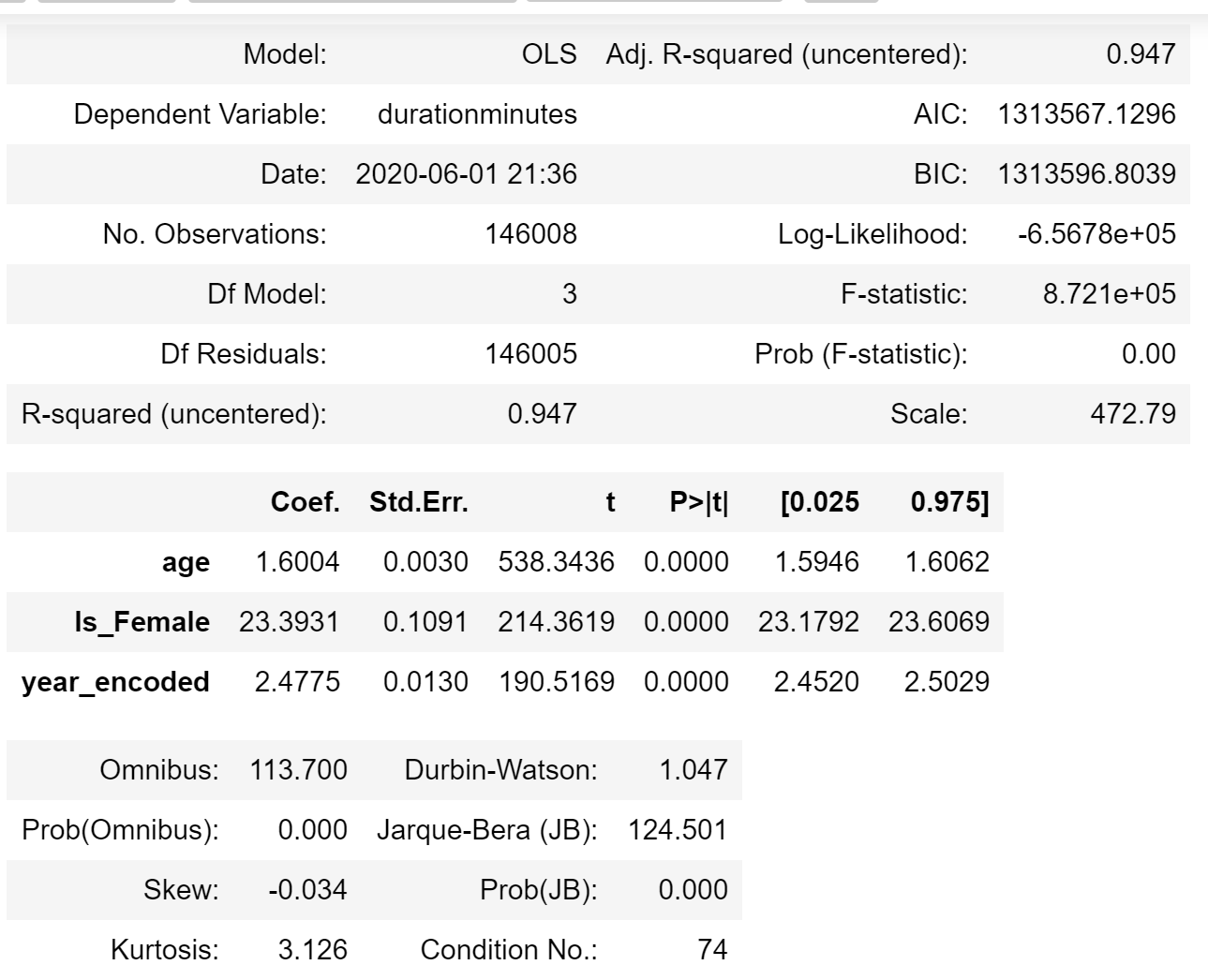
We further investigated normality with the Shapiro-Wilk test. The null-hypothesis of this test is that ages are normally distributed for each gender. The test shows strong evidence that the distribution is not normal. Results shown in **Figure 15**. This backs the findings of both QQ plots.

**Figure 15: Shapiro-Wilk Test for Age**

To complete our analysis, we tested the difference in age means between male and female runners using a t-test. The evidence suggests that the data is not normally disturbed and has unequal sample sizes between males and females. The Welch–Satterthwaite equation is used to approximate the adjusted degrees of freedom to help correct for this.

**Figure 16: Welch-Satterthwaite Test for Age**

Our null hypothesis (no difference in average age between men and women) is rejected, showing strong evidence (p-value = 0.0) that the means are different. Results shown in **Figure 16.**

A simple OLS regression model helps us further understand the relationship of age, time, and gender. The results of the regression test indicate that age, female, and year are all significant (small p-values less than 0.000) for each dependent variable. The R-squared value is also extremely high at 0.947, which indicates an incredibly strong model. [WHICH MEANS WHAT]. However, the results seem too perfect, so further investigation with models that do not rely as on data set assumptions should be done.

**Figure 17: Results of Regression Model**

**Conclusion (Q10)**

Through statistical analysis and visual analysis of supporting charts, our team assessed how the race has grown from 1999 to 2012, and how the age and time distributions have changed during that timeframe. Our analysis shows that gender does have a significant impact on race time, as the regression model (**Figure 17**) and violin chart (**Figure 14**) indicate. As expected, the race times do slightly increase as age increases. It is taking runners, on average, approximately 2.5 minutes longer each year. This is more evident in the results of female runners than male. Given the larger proportion of female runners in recent years, their slower times may be overshadowing any gains in time by male runners. These changes are all very gradual, which could be driven by a number of factors such as the how many elite runners participate, the performances by runners in the largest age groups, or the weather on any given race day.

Across women runners, the average age is decreasing between 1999 and 2012. This can be attributed both to the growth of participants in the 20-40 age range and fewer older participants in recent years. The change was gradual, except between 2005 and 2006 when there was a drop in older runners.

Men runners, on average, are also getting young, likely due in part to the growth in participants age 20-40 as well. This change was also gradual. Interestingly, recent years have more male older outlier runners than women.

**CODE DIRECTORY**

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**APPENDIX**

**Appendix Figure A:**

**Data Dictionary**

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| --- | --- | --- |
| **Variable** | **Description** | **Action Taken** |
| Gender | Male or Female; gender of participants; scraped from individual Cherry Blossom website pages | Scraped, no cleaning necessary, data label |
| Year | 1999 to 2012; scraped from individual Cherry Blossom website pages | Scraped, no cleaning necessary, data label |
| Place | Place in each year’s race, separated by Gender and Year | Scraped, no cleaning necessary, did not use |
| Div\_Total | Place of Each Participant in Age Division; for each year | Scraped, no cleaning necessary, did not use |
| Name | Participant Name | Scraped, no cleaning necessary, did not use individual names |
| Age | Age of Each Participant | Scraped, cleaned up missing or outlier ages |
| Hometown | Home of Each Participants, either City ST or Country | Scraped, no cleaning, did not use |
| Time | One of: Time, Gun Time, Net Time; or Comb Time; Time – Overall Participant Time; Gun Time – Time of Participant from Gun to Individual Finish; Net Time – Time of Participant from Start Line to Finish Line. | Scraped, needed to assess which times provided; used in order – Net Time, Gun Time, or Comb Time; type depended on race year |
| Pace | Average Mile Per Hour for Each Participant | Scraped, discovered odd values; did not use |
| Num ID | Cherry Blossom Participant ID, not available for all years | Scraped, no cleaning necessary, did not use |
| Net Time | Net Time for Each Participant, where available | Scraped, where available |
| Comb Time | Combined Time of Each Participant, usually reflects Gun Time | Scraped, where available |
| Time Length | Created Variable to help calculate time | Created to assist in managing time format for calculations |
| F Combine | Created Variable | Created to assist in managing time format for calculations |
| Final Time | Created Variable | Created to assist in managing time format for calculations |
| Hour | Created Variable to separate time: hours | Created to assist in managing time format for calculations |
| Minutes | Created Variable to separate time: minutes | Created to assist in managing time format for calculations |
| Seconds | Created Variable to separate time: seconds | Created to assist in managing time format for calculations |
| Dur Secs | Calculated Variable; Total Race Time in Seconds | Created to assist in managing time format for calculations |
| Duration Min | Calculated Variable; Total Race Time in Minutes | Created to assist in managing time format for calculations |
| Calc Pace | Calculated Pace from Time | Created to assist in managing time format for calculations |
| Counter | Created Variable; tally of 1 for each participant | Created to assist in managing participant count |