



CREDIT UNION **CHERRY BLOSSOM** **10M RUN**

Who are the future runners?



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FUTURE RUNNERS

The popularity of the Cherry Blossom run has grown over the years. What can be done to improve the race for our runners to encourage more participation and further growth?

HIGH-LEVEL OBJECTIVES:

Understand the changing profile of runners

- …Is more seeding time structure needed?
- …Are runners younger/older, faster/slower?
- …What is the overall trend, year over year?



MODELING FOR SUCCESS

DELIVERING BUSINESS VALUE

AMONG SEVERAL REASONS, THE FOLLOWING ARE A FEW MAJOR PURPOSES OF PROPER RACE TIMING AND RACER PROFILE UNDERSTANDING

- **Safety:** it is important to ensure that there aren't too many runners at once in the starting interval in order to *reduce liability*
- **Concessions:** Staging can affect flow of runners finishing the race. Improper staging may *bring revenue to concession stands* by creating unnecessarily long lines
- **Entertainment:** Like concessions, timing of live music and performances is important to ensure that *all participants can enjoy live entertainment*.

DATA EXTRACTION

/ WEB SCRAPE



Leveraged R to pull data iteratively from www.cballtimeresults.org to capture all 14 years of female run result data

/ NO RECORDS



20 records were removed for missing age or location (less than 0.02 % of total data)



/ CLEAN



Converted to proper data types and scrubbed extraneous field strings



/ ENGINEERING

Split columns for further analysis and understanding (e.g. Year, Hometown > City + State)

WEB SCRAPER

Leveraging online data

1999 10M Event Results for Overall Women

Name	Age	Time	Pace	PIS/TIS Ⓢ	Division	PID/TID Ⓢ	Hometown
Jane Omoro (W)	26	0:53:37	5:22	1/2358	W2529	1/559	Kenya
Jane Ngotho (W)	29	0:53:38	5:22	2/2358	W2529	2/559	Kenya
Lidiya Grigoryeva (W)	NR	0:53:40	5:22	3/2358	NR	NR	Russia
Eunice Sagero (W)	20	0:53:55	5:24	4/2358	W2024	1/196	Kenya
Alla Zhilyayeva (W)	29	0:54:08	5:25	5/2358	W2529	3/559	Russia
Teresa Wanjiku (W)	24	0:54:10	5:25	6/2358	W2024	2/196	Kenya
Elana Viazova (W)	38	0:54:29	5:27	7/2358	W3539	1/387	Ukraine
Gladys Asiba (W)	NR	0:54:50	5:29	8/2358	NR	NR	Kenya
Nnenna Lynch (W)	27	0:55:39	5:34	9/2358	W2529	4/559	Concord, MA
Margaret Kagiri (W)	30	0:55:43	5:34	10/2358	W3034	1/529	Kenya
Susannah Beck (W)	30	0:56:13	5:37	11/2358	W3034	2/529	Eugene, OR
Kelly Keeler (W)	37	0:57:23	5:44	12/2358	W3539	2/387	Bloomington, MN
Marie Boyd (W)	39	0:57:24	5:44	13/2358	W3539	3/387	Albuquerque, NM
Betsy Kempter (W)	32	0:57:51	5:47	14/2358	W3034	3/529	Chapel Hill, NC
Naoko Ishibe (W)	30	0:58:05	5:49	15/2358	W3034	4/529	Washington, DC
Bea Marie Altieri (W)	31	0:58:36	5:52	16/2358	W3034	5/529	Columbia, MD
Connie Buckwalter (W)	NR	0:59:36	5:58	17/2358	NR	NR	Lancaster, PA
Sharon Servidio (W)	25	0:59:42	5:58	18/2358	W2529	5/559	Alexandria, VA
Donna Moore (W)	38	0:59:48	5:59	19/2358	W3539	4/387	Silver Spring, MD
Marian Huizing (W)	31	1:00:30	6:03	20/2358	W3034	6/529	Rockville, MD

prev

next

- Tables were parsed from website <https://www.cherryblossom.org/>
- Tools comprised of R's "tidyverse", "XML2", and more
- Created an algorithm to parse an unknown number of pages for a known time year period
- Performed additional splits and age binning

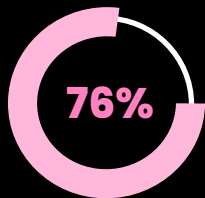
Year	Name	Age	Time	Pace	PIS	TIS	Division	PID	TID	City	State	Country	TotalTimeSec	TotalTimeMin	PaceTimeSec	PaceTimeMin	AgeGroup
1	Jane Omoro	26	0:53:37	5:22	1/2358	W2529	1/559	Kenya	3217	53.617	322	5.367	28s				
2	Jane Ngotho	29	0:53:38	5:22	2/2358	W2529	2/559	Kenya	3218	53.633	322	5.367	28s				
3	Lidiya Grigoryeva	NR	0:53:40	5:22	3/2358	NR	NR	Russia	3220	53.667	322	5.367	NR				
4	Eunice Sagero	20	0:53:55	5:24	4/2358	W2024	1/196	Kenya	3235	53.917	324	5.400	28s				
5	Alla Zhilyayeva	29	0:54:08	5:25	5/2358	W2529	3/559	Russia	3248	54.133	325	5.417	28s				
6	Teresa Wanjiku	24	0:54:10	5:25	6/2358	W2024	2/196	Kenya	3250	54.167	325	5.417	28s				
7	Elana Viazova	38	0:54:29	5:27	7/2358	W3539	1/387	Ukraine	3269	54.483	327	5.483	38s				
8	Gladys Asiba	NR	0:54:50	5:29	8/2358	NR	NR	Kenya	3290	54.833	329	5.483	NR				
9	Nnenna Lynch	27	0:55:39	5:34	9/2358	W2529	4/559	Concord, MA	3339	55.650	334	5.567	28s				
10	Margaret Kagiri	30	0:55:43	5:34	10/2358	W3034	1/529	Kenya	3343	55.717	334	5.567	38s				
11	Susannah Beck	30	0:56:13	5:37	11/2358	W3034	2/529	Eugene, OR	3373	56.217	337	5.617	38s				
12	Kelly Keeler	37	0:57:23	5:44	12/2358	W3539	2/387	Bloomington, MN	3443	57.383	344	5.733	38s				
13	Marie Boyd	39	0:57:24	5:44	13/2358	W3539	3/387	Albuquerque, NM	3444	57.400	344	5.733	38s				
14	Betsy Kempter	32	0:57:51	5:47	14/2358	W3034	3/529	Chapel Hill, NC	3471	57.850	347	5.783	38s				
15	Naoko Ishibe	30	0:58:05	5:49	15/2358	W3034	4/529	Washington, DC	3485	58.083	349	5.817	38s				
16	Bea Marie Altieri	31	0:58:36	5:52	16/2358	W3034	5/529	Columbia, MD	3516	58.600	352	5.867	38s				
17	Connie Buckwalter	NR	0:59:36	5:58	17/2358	NR	NR	Lancaster, PA	3576	59.600	358	5.967	NR				
18	Sharon Servidio	25	0:59:42	5:58	18/2358	W2529	5/559	Alexandria, VA	3582	59.700	358	5.967	28s				
19	Donna Moore	38	0:59:48	5:59	19/2358	W3539	4/387	Silver Spring, MD	3588	59.800	359	5.983	38s				
20	Marian Huizing	31	1:00:30	6:03	20/2358	W3034	6/529	Rockville, MD	3630	60.500	363	6.050	38s				
21	Nurtha Merz	36	1:00:32	6:03	21/2358	W3539	5/387	Annandale, VA	3632	60.533	363	6.050	38s				
22	Wendy Nelson-Barrett	30	1:00:43	6:04	22/2358	W3034	7/529	Lebanon, PA	3643	60.717	364	6.067	38s				
23	Patti Small	40	1:00:47	6:05	23/2358	W4044	5/386	Ashburn, VA	3647	60.783	365	6.083	48s				
24	Anita Frenes	34	1:01:04	6:06	24/2358	W3034	8/529	Rockville, MD	3664	61.067	366	6.100	38s				
25	Aryn Fahey	22	1:01:15	6:08	25/2358	W2024	3/196	Fairfax, VA	3675	61.250	368	6.133	28s				
26	Debrae Ficker	22	1:01:28	6:08	26/2358	W2024	4/196	Patuxent, CT	3680	61.333	368	6.133	28s				
27	Kate Petricone	37	1:01:59	6:12	27/2358	W3539	6/387	Winnet, PA	3719	61.983	372	6.200	27s				
28	Jennifer Janis	29	1:02:07	6:13	28/2358	W2529	6/559	Milvern, MA	3727	62.117	373	6.217	28s				
29	Christina Henning	34	1:02:12	6:13	29/2358	W3034	9/529	Shorewood, WI	3732	62.200	373	6.217	38s				
30	Patty Fulton	31	1:02:16	6:14	30/2358	W3034	10/529	Silver Spring, MD	3736	62.267	374	6.233	38s				
31	Nancy Watkins	30	1:02:28	6:15	31/2358	W3034	11/529	Stafford, VA	3748	62.467	375	6.250	38s				
32	Kirsten Black	24	1:02:45	6:17	32/2358	W2024	5/196	Arlington, VA	3765	62.750	377	6.283	28s				
33	Dorian Meyer	39	1:02:53	6:17	33/2358	W2529	7/387	Ramson, NJ	3771	62.850	377	6.283	38s				
34	Karen Oudekerk	31	1:03:01	6:18	34/2358	W3034	12/529	Arlington, VA	3781	63.017	378	6.300	38s				
35	Anna Britt	34	1:03:18	6:19	35/2358	W3034	13/529	Summit, NJ	3790	63.167	379	6.317	38s				
36	Denise Erickson	30	1:03:31	6:21	36/2358	W3034	14/529	Baltimore, MD	3811	63.517	381	6.350	38s				
37	Cecily Tyson	30	1:03:33	6:21	37/2358	W3034	15/529	Philadelphia, PA	3813	63.550	381	6.350	37s				
38	Leslie Minix-Wolfe	37	1:03:53	6:23	38/2358	W3539	8/387	Reston, VA	3833	63.883	383	6.383	38s				
39	Mag Ritter	25	1:04:16	6:24	39/2358	W2529	7/559	Richmond, VA	3836	63.933	384	6.400	28s				
40	Debi Bernades	40	1:04:15	6:24	40/2358	W4044	5/386	King George, VA	3855	64.250	386	6.433	48s				
41	Linda Buck	43	1:05:01	6:30	41/2358	W4044	3/386	Germanston, MD	3901	65.017	390	6.500	48s				
42	Patricia Keating	37	1:05:17	6:32	42/2358	W3539	9/387	Clarksville, MD	3917	65.283	392	6.533	38s				

Sample of page 1 of women's 10k runner results for 1999

Parsed table of women's run data, holding more than 75,000 records

DATA

/ AGE



Women in their
20's & 30's

/ TOP STATES

VA, MA & DC



Runners come from just
3 states

/ AVERAGE TIME

01:39:01

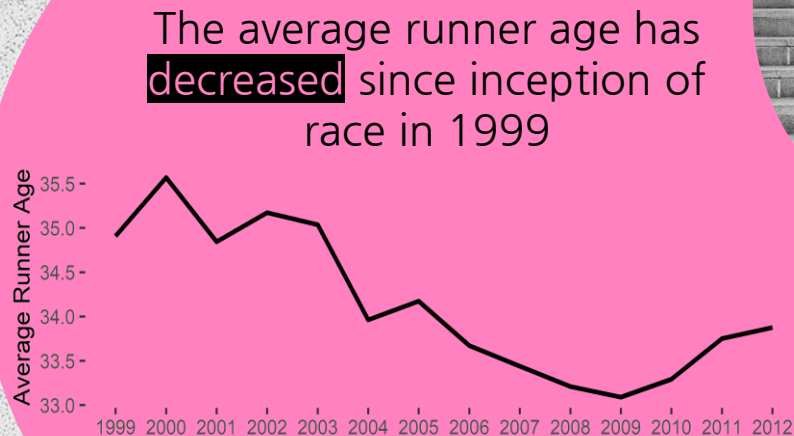


CHERRY BLOSSOM
WOMEN'S 10M RUN

**Data from 2012 run*

RUNNER TRENDS

+0.29%
Average Race Time
Trend (1999-2012)



12%

The average **growth**
of number of runners
each year

SPOTLIGHT

Muffet Chatterton
Crofton, MD

Year	Age	Pace	PiD.TiD
1999	54	8:10	10/114
2000	55	8:13	6/60
2001	56	8:26	10/66
2002	57	9:06	17/72
2003	58	8:43	16/83
2004	59	9:13	17/83
2006	61	8:46	8/48
2007	62	9:22	5/33
2008	63	9:34	19/50
2009	64	9:43	17/56
2010	65	9:57	6/17
2011	66	9:52	12/26
2012	67	10:07	14/35

PiD = Position
in Division
TiD = Total in
Division



Muffet has ran the Cherry Blossom race from 1999-2012. She was unable to attend in 2005 but came back in 2006 with a vengeance posting a blistering 8:46 pace!

She routinely finished in the top 15 in her division year after year and completed the 2012 race at age 67.

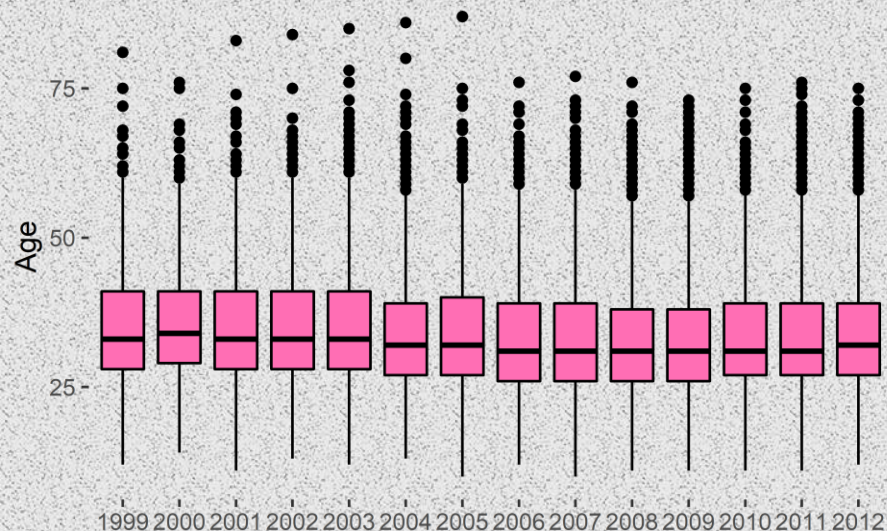
Cheers to you, Muffet!

FAST FACTS



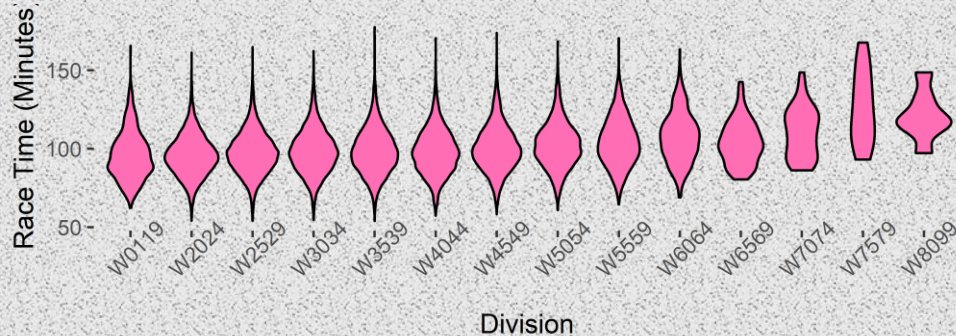
Runner Age Distribution

Runner age distribution has stayed relatively consistent year over year.



Race Time by Division

W3539 sees the widest range of finish times compared to W8099 with the smallest range.



ANOVA METHOD

Analysis of variance, or ANOVA, is a statistical method used to compare averages of more than two groups.

In this analysis, ANOVA is used to compare average race time in minutes across age groups and/or by year.

An ANOVA that has a significant p-value (< 0.05), or a confidence interval that does not include zero, suggests that the averages in comparison are statistically different than one another.



ANOVA RESULTS

AGE COMPARISON	/ AVG TIME DIFF	/ LOWER CONF INT	/ UPPER CONF INT	/ P-VALUE <small>* Tukey Adjusted</small>
/ 01-19/20-29	1.1441	-0.5264	2.8146	0.4307
/ 70-79/80-89	7.0590	-9.8772	23.9952	0.9123
/ 60-69/80-89	14.4883	-1.6633	30.6399	0.1168

The above chart highlights where p-values are **not significant**, indicating there is **no significant difference in mean run times between age groups**. There is a significant difference in run times between every other age combination outside of this table.

At a 95% confidence level, the upper and lower bounds include zero, which is another indication that zero is a plausible difference in mean run time for the groups above.

RECOMMENDATIONS

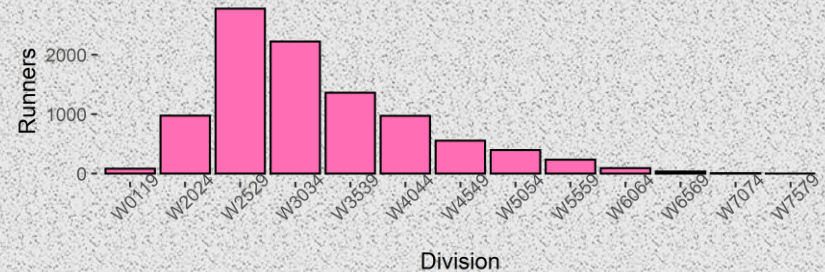
10M RUN – OPERATIONS DAY OF

- Start slow divisions earlier
- Expect approximately 10,000 runners for 2013
- Break larger divisions into multiple gates

NEW DIVISIONS

- Under 30 Club (combines 0-19, 20-24, 25-29)
 - Largest runner population in this group
 - No significant difference in time
- Golden Years (combines 70-74, 75-79, 80-100)

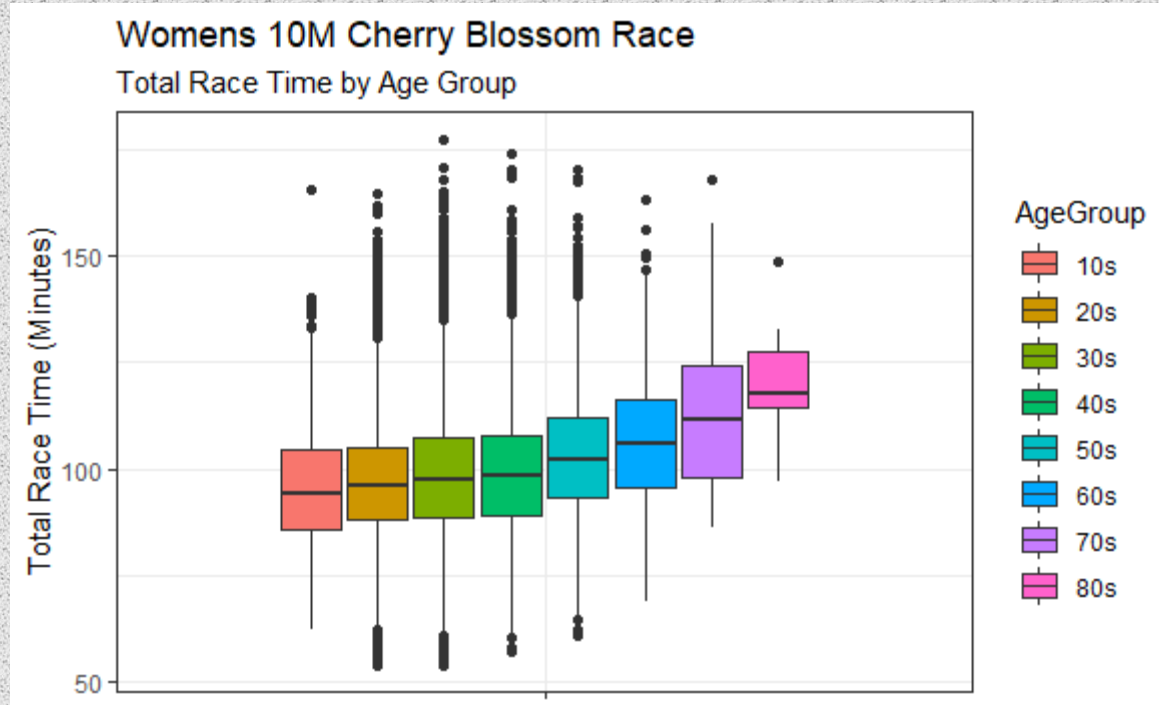
NUMBER OF RUNNERS BY DIVISION



APPENDIX



AVERAGE RACE TIME VARIES BY AGE GROUP



AVERAGE RACE TIME WITHIN AN AGE GROUP

VARIES VERY LITTLE

