

# MATH 3070 Lab Project 5

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*Remember: I expect to see commentary either in the text, in the code with comments created using #, or (preferably) both! **Failing to do so may result in lost points!***

## Problem 1 (Verzani problem 4.7)

Use the `subset()` function to return a data frame made from the `Cars93` (**MASS**) data frame consisting only of **non-USA** cars in origin, with 4 cylinders and a maximum price of \$15,000 or less.

```
library(MASS)

# Subset the Cars93 data frame based on three conditions:
# - Origin should be "non-USA"
# - The car should have 4 cylinders
# - The price should be less than or equal to 15000
non_usa_4cylinders_max15000 <- subset(Cars93,
                                     subset = (Origin == "non-USA" & Cylinders == 4 & Max.Price <= 15.0))

# Output the resulting subset data frame
non_usa_4cylinders_max15000
```

##	Manufacturer	Model	Type	Min.Price	Price	Max.Price	MPG.city	MPG.highway
## 40	Geo	Storm	Sporty	11.5	12.5	13.5	30	36
## 44	Hyundai	Excel	Small	6.8	8.0	9.2	29	33
## 45	Hyundai	Elantra	Small	9.0	10.0	11.0	22	29
## 46	Hyundai	Scoupe	Sporty	9.1	10.0	11.0	26	34
## 53	Mazda	323	Small	7.4	8.3	9.1	29	37
## 54	Mazda	Protege	Small	10.9	11.6	12.3	28	36
## 62	Mitsubishi	Mirage	Small	7.7	10.3	12.9	29	33
## 64	Nissan	Sentra	Small	8.7	11.8	14.9	29	33
## 81	Subaru	Loyale	Small	10.5	10.9	11.3	25	30
## 84	Toyota	Tercel	Small	7.8	9.8	11.8	32	37

## 88	Volkswagen	Fox	Small	8.7	9.1	9.5	25	33
##	AirBags	DriveTrain	Cylinders	EngineSize	Horsepower	RPM	Rev.per.mile	
## 40	Driver only	Front	4	1.6	90	5400	3250	
## 44	None	Front	4	1.5	81	5500	2710	
## 45	None	Front	4	1.8	124	6000	2745	
## 46	None	Front	4	1.5	92	5550	2540	
## 53	None	Front	4	1.6	82	5000	2370	
## 54	None	Front	4	1.8	103	5500	2220	
## 62	None	Front	4	1.5	92	6000	2505	
## 64	Driver only	Front	4	1.6	110	6000	2435	
## 81	None	4WD	4	1.8	90	5200	3375	
## 84	Driver only	Front	4	1.5	82	5200	3505	
## 88	None	Front	4	1.8	81	5500	2550	
##	Man.trans.avail	Fuel.tank.capacity	Passengers	Length	Wheelbase	Width		
## 40	Yes	12.4	4	164	97	67		
## 44	Yes	11.9	5	168	94	63		
## 45	Yes	13.7	5	172	98	66		
## 46	Yes	11.9	4	166	94	64		
## 53	Yes	13.2	4	164	97	66		
## 54	Yes	14.5	5	172	98	66		
## 62	Yes	13.2	5	172	98	67		
## 64	Yes	13.2	5	170	96	66		
## 81	Yes	15.9	5	175	97	65		
## 84	Yes	11.9	5	162	94	65		
## 88	Yes	12.4	4	163	93	63		
##	Turn.circle	Rear.seat.room	Luggage.room	Weight	Origin	Make		
## 40	37	24.5	11	2475	non-USA	Geo Storm		
## 44	35	26.0	11	2345	non-USA	Hyundai Excel		
## 45	36	28.0	12	2620	non-USA	Hyundai Elantra		
## 46	34	23.5	9	2285	non-USA	Hyundai Scoupe		
## 53	34	27.0	16	2325	non-USA	Mazda 323		
## 54	36	26.5	13	2440	non-USA	Mazda Protege		
## 62	36	26.0	11	2295	non-USA	Mitsubishi Mirage		
## 64	33	26.0	12	2545	non-USA	Nissan Sentra		
## 81	35	27.5	15	2490	non-USA	Subaru Loyale		
## 84	36	24.0	11	2055	non-USA	Toyota Tercel		
## 88	34	26.0	10	2240	non-USA	Volkswagen Fox		

## Problem 2 (Verzani problem 4.10)

Find the standard deviation of each variable in the *mtcars* data set.

```
# Apply the sapply function to get the standard deviation of each column in mtcars
sapply(mtcars, sd)
```

##	mpg	cyl	disp	hp	drat	wt
##	6.0269481	1.7859216	123.9386938	68.5628685	0.5346787	0.9784574
##	qsec	vs	am	gear	carb	
##	1.7869432	0.5040161	0.4989909	0.7378041	1.6152000	

## Problem 3 (Verzani problem 4.11)

Find the standard deviation for each numeric variable in *Cars93* (MASS).

```
# Subset Cars93 to only numeric variables and calculate the standard deviation
sapply(Cars93[sapply(Cars93, is.numeric)], sd)
```

```
##      Min.Price      Price      Max.Price      MPG.city
##      8.746029      9.659430      11.030457      5.619812
##      MPG.highway      EngineSize      Horsepower      RPM
##      5.331726      1.037363      52.374410      596.731690
##      Rev.per.mile      Fuel.tank.capacity      Passengers      Length
##      496.506525      3.279370      1.038979      14.602382
##      Wheelbase      Width      Turn.circle      Rear.seat.room
##      6.819674      3.778986      3.223265      NA
##      Luggage.room      Weight
##      NA      589.896510
```

## Problem 4

You can find data describing various statistics on 2012 Olympic participating nations such as population and medal counts in the comma-separated value file (*olympic-medals2012.csv*). Load this dataset into R in a data frame and do the following:

1. Create a new data frame containing only the country's ISO code (a three-letter code for the country; USA is the United States of America, and DZA Algeria), name, population, and counts for gold, silver, and bronze. This data frame should be called *olympic2012*.

```
# Load the olympic-medals2012.csv file
olympic_data <- read.csv("C:/Users/Prachi/OneDrive/Documents/olympic-medals2012.csv")

# Create a new data frame containing selected columns: ISO code, Country name, Population, Gold, Silver, Bronze
olympic2012 <- olympic_data[, c("ISO", "Country.name", "pop.2010", "Gold", "Silver", "Bronze")]

# Display the olympic2012 data frame
olympic2012
```

```
##      ISO      Country.name      pop.2010      Gold      Silver      Bronze
## 1  AFG      Afghanistan      34,385,000      0      0      1
## 2  ALB      Albania      3,205,000      0      0      0
## 3  DZA      Algeria      35,468,000      1      0      0
## 4  ASM      American Samoa      68,420      0      0      0
## 5  AND      Andorra      84,864      0      0      0
## 6  AGO      Angola      19,082,000      0      0      0
## 7  ATG      Antigua and Barbuda      88,000      0      0      0
## 8  ARG      Argentina      40,412,000      1      1      2
## 9  ARM      Armenia      3,092,000      0      1      2
## 10 ABW      Aruba      108,000      0      0      0
## 11 AUS      Australia      22,299,000      7      16      12
## 12 AUT      Austria      8,390,000      0      0      0
## 13 AZE      Azerbaijan      9,054,000      2      2      6
## 14 BHS      Bahamas      343,000      1      0      0
## 15 BHR      Bahrain      1,262,000      0      0      1
## 16 BGD      Bangladesh      148,692,000      0      0      0
## 17 BRB      Barbados      274,000      0      0      0
```

## 18	BLR	Belarus	9,490,000	2	5	5
## 19	BEL	Belgium	10,896,000	0	1	2
## 20	BLZ	Belize	345,000	0	0	0
## 21	BEN	Benin	8,850,000	0	0	0
## 22	BMU	Bermuda	64,600	0	0	0
## 23	BTN	Bhutan	726,000	0	0	0
## 24	BOL	Bolivia	9,929,000	0	0	0
## 25	BIH	Bosnia and Herzegovina	3,760,000	0	0	0
## 26	BWA	Botswana	2,007,000	0	1	0
## 27	BRA	Brazil	194,946,000	3	5	9
## 28	VGB	British Virgin Islands	88,000	0	0	0
## 29	BRN	Brunei	399,000	0	0	0
## 30	BGR	Bulgaria	7,534,000	0	1	1
## 31	BFA	Burkina Faso	16,468,000	0	0	0
## 32	MMR	Burma	47,963,000	0	0	0
## 33	BDI	Burundi	8,382,000	0	0	0
## 34	KHM	Cambodia	14,139,000	0	0	0
## 35	CMR	Cameroon	19,599,000	0	0	0
## 36	CAN	Canada	34,126,000	1	5	12
## 37	CPV	Cape Verde	496,000	0	0	0
## 38	CYM	Cayman Islands	56,230	0	0	0
## 39	CAF	Central African Republic	4,401,000	0	0	0
## 40	TCD	Chad	11,227,000	0	0	0
## 41	CHL	Chile	17,113,688	0	0	0
## 42	CHN	China	1,338,300,000	38	27	23
## 43	COL	Colombia	46,295,000	1	3	4
## 44	COM	Comoros	735,000	0	0	0
## 45	COG	Congo-Brazzaville	4,043,000	0	0	0
## 46	COK	Cook Islands	20,000	0	0	0
## 47	CRI	Costa Rica	4,659,000	0	0	0
## 48	HRV	Croatia	4,418,000	3	1	2
## 49	CUB	Cuba	11,258,000	5	3	6
## 50	CYP	Cyprus	659,350	0	1	0
## 51	CZE	Czech Republic	10,520,000	4	3	3
## 52	COD	Democratic Republic of Congo	65,965,000	0	0	0
## 53	DNK	Denmark	5,547,000	2	4	3
## 54	DJI	Djibouti	889,000	0	0	0
## 55	DMA	Dominica	68,000	0	0	0
## 56	DOM	Dominican Republic	9,927,000	1	1	0
## 57	TLS	East Timor	1,124,000	0	0	0
## 58	ECU	Ecuador	14,465,000	0	0	0
## 59	EGY	Egypt	81,121,000	0	2	0
## 60	SLV	El Salvador	6,193,000	0	0	0
## 61	GNQ	Equatorial Guinea	700,000	0	0	0
## 62	ERI	Eritrea	5,254,000	0	0	0
## 63	EST	Estonia	1,340,000	0	1	1
## 64	ETH	Ethiopia	82,950,000	3	1	3
## 65	FJI	Fiji	860,000	0	0	0
## 66	FIN	Finland	5,364,000	0	1	2
## 67	FRA	France	64,895,000	11	11	12
## 68	GAB	Gabon	1,505,000	0	1	0
## 69	GMB	Gambia	1,729,000	0	0	0
## 70	GEO	Georgia	4,452,000	1	3	3
## 71	DEU	Germany	81,777,000	11	19	14

## 72	GHA	Ghana	24,392,000	0	0	0
## 73	GRC	Greece	11,316,000	0	0	2
## 74	GRD	Grenada	104,000	1	0	0
## 75	GUM	Guam	179,000	0	0	0
## 76	GTM	Guatemala	14,389,000	0	1	0
## 77	GIN	Guinea	9,982,000	0	0	0
## 78	GNB	Guinea-Bissau	1,515,000	0	0	0
## 79	GUY	Guyana	755,000	0	0	0
## 80	HTI	Haiti	9,993,000	0	0	0
## 81	HND	Honduras	7,600,000	0	0	0
## 82	HKG	Hong Kong	7,068,000	0	0	1
## 83	HUN	Hungary	10,000,000	8	4	5
## 84	ISL	Iceland	318,000	0	0	0
## 85	IND	India	1,224,615,000	0	2	4
## 86	IDN	Indonesia	239,870,000	0	1	1
## 87	IRN	Iran	73,973,000	4	5	3
## 88	IRQ	Iraq	32,031,000	0	0	0
## 89	IRL	Ireland	4,475,000	1	1	3
## 90	ISR	Israel	7,624,000	0	0	0
## 91	ITA	Italy	60,483,000	8	9	11
## 92	CIV	Ivory Coast	19,738,000	0	0	0
## 93	JAM	Jamaica	2,702,000	4	4	4
## 94	JPN	Japan	127,451,000	7	14	17
## 95	JOR	Jordan	6,047,000	0	0	0
## 96	KAZ	Kazakhstan	16,323,000	7	1	5
## 97	KEN	Kenya	40,513,000	2	4	5
## 98	KIR	Kiribati	100,000	0	0	0
## 99	KWT	Kuwait	2,736,000	0	0	1
## 100	KGZ	Kyrgyzstan	5,448,000	0	0	0
## 101	LAO	Laos	6,201,000	0	0	0
## 102	LVA	Latvia	2,239,000	1	0	1
## 103	LBN	Lebanon	4,227,000	0	0	0
## 104	LSO	Lesotho	2,171,000	0	0	0
## 105	LBR	Liberia	3,994,000	0	0	0
## 106	LBY	Libya	6,355,000	0	0	0
## 107	LIE	Liechtenstein	36,032	0	0	0
## 108	LTU	Lithuania	3,287,000	2	1	2
## 109	LUX	Luxembourg	507,000	0	0	0
## 110	MKD	Macedonia	2,060,000	0	0	0
## 111	MDG	Madagascar	20,714,000	0	0	0
## 112	MWI	Malawi	14,901,000	0	0	0
## 113	MYS	Malaysia	28,401,000	0	1	1
## 114	MDV	Maldives	316,000	0	0	0
## 115	MLI	Mali	15,370,000	0	0	0
## 116	MLT	Malta	416,000	0	0	0
## 117	MHL	Marshall Islands	54,038	0	0	0
## 118	MRT	Mauritania	3,460,000	0	0	0
## 119	MUS	Mauritius	1,281,000	0	0	0
## 120	MEX	Mexico	113,423,000	1	3	3
## 121	FSM	Micronesia	111,000	0	0	0
## 122	MDA	Moldova	3,562,000	0	0	2
## 123	MCO	Monaco	35,407	0	0	0
## 124	MNG	Mongolia	2,756,000	0	2	3
## 125	MNE	Montenegro	632,000	0	1	0

## 126	MAR	Morocco	31,951,000	0	0	1
## 127	MOZ	Mozambique	23,390,000	0	0	0
## 128	NAM	Namibia	2,283,000	0	0	0
## 129	NRU	Nauru	10,372	0	0	0
## 130	NPL	Nepal	29,959,000	0	0	0
## 131	NLD	Netherlands	16,616,000	6	6	8
## 132	NZL	New Zealand	4,368,000	6	2	5
## 133	NIC	Nicaragua	5,789,000	0	0	0
## 134	NER	Niger	15,512,000	0	0	0
## 135	NGA	Nigeria	158,423,000	0	0	0
## 136	PRK	North Korea	24,589,122	4	0	2
## 137	NOR	Norway	4,889,000	2	1	1
## 138	OMN	Oman	2,783,000	0	0	0
## 139	PAK	Pakistan	173,593,000	0	0	0
## 140	PLW	Palau	20,472	0	0	0
## 141	PSE	Palestine	4,152,000	0	0	0
## 142	PAN	Panama	3,517,000	0	0	0
## 143	PNG	Papua New Guinea	6,858,000	0	0	0
## 144	PRY	Paraguay	6,454,000	0	0	0
## 145	PER	Peru	29,076,000	0	0	0
## 146	PHL	Philippines	93,261,000	0	0	0
## 147	POL	Poland	38,184,000	2	2	6
## 148	PRT	Portugal	10,638,000	0	1	0
## 149	PRI	Puerto Rico	3,978,000	0	1	1
## 150	QAT	Qatar	1,759,000	0	0	2
## 151	ROU	Romania	21,438,000	2	5	2
## 152	RUS	Russia	141,750,000	24	26	32
## 153	RWA	Rwanda	10,624,000	0	0	0
## 154	KNA	Saint Kitts and Nevis	52,000	0	0	0
## 155	LCA	Saint Lucia	174,000	0	0	0
## 156	VCT	Saint Vincent and the Grenadines	109,000	0	0	0
## 157	WSM	Samoa	184,000	0	0	0
## 158	SMR	San Marino	31,534	0	0	0
## 159	STP	Sao Tome and Principe	165,000	0	0	0
## 160	SAU	Saudi Arabia	27,448,000	0	0	1
## 161	SEN	Senegal	12,434,000	0	0	0
## 162	SRB	Serbia	7,291,000	1	1	2
## 163	SYC	Seychelles	87,000	0	0	0
## 164	SLE	Sierra Leone	5,867,000	0	0	0
## 165	SGP	Singapore	5,077,000	0	0	2
## 166	SVK	Slovakia	5,430,000	0	1	3
## 167	SVN	Slovenia	2,049,000	1	1	2
## 168	SLB	Solomon Islands	538,000	0	0	0
## 169	SOM	Somalia	9,331,000	0	0	0
## 170	ZAF	South Africa	49,991,000	3	2	1
## 171	KOR	South Korea	48,875,000	13	8	7
## 172	ESP	Spain	46,071,000	3	10	4
## 173	LKA	Sri Lanka	20,860,000	0	0	0
## 174	SDN	Sudan	43,552,000	0	0	0
## 175	SUR	Surinam	525,000	0	0	0
## 176	SWZ	Swaziland	1,056,000	0	0	0
## 177	SWE	Sweden	9,378,000	1	4	3
## 178	CHE	Switzerland	7,826,000	2	2	0
## 179	SYR	Syria	20,447,000	0	0	0

## 180	TWN	Taiwan	23,174,528	0	1	1
## 181	TJK	Tajikistan	6,879,000	0	0	1
## 182	TZA	Tanzania	44,841,000	0	0	0
## 183	THA	Thailand	69,122,000	0	2	1
## 184	TGO	Togo	6,028,000	0	0	0
## 185	TON	Tonga	104,000	0	0	0
## 186	TTO	Trinidad and Tobago	1,341,000	1	0	3
## 187	TUN	Tunisia	10,549,000	1	1	1
## 188	TUR	Turkey	72,752,000	2	2	1
## 189	TKM	Turkmenistan	5,042,000	0	0	0
## 190	TUV	Tuvalu	9,827	0	0	0
## 191	UGA	Uganda	33,424,000	1	0	0
## 192	GBR	UK	62,232,000	29	17	19
## 193	UKR	Ukraine	45,871,000	6	5	9
## 194	ARE	United Arab Emirates	7,512,000	0	0	0
## 195	URY	Uruguay	3,357,000	0	0	0
## 196	USA	US	309,349,000	46	29	29
## 197	UZB	Uzbekistan	28,228,000	1	0	3
## 198	VUT	Vanuatu	240,000	0	0	0
## 199	VEN	Venezuela	28,834,000	1	0	0
## 200	VNM	Vietnam	86,928,000	0	0	0
## 201	VIR	Virgin Islands	110,000	0	0	0
## 202	YEM	Yemen	24,053,000	0	0	0
## 203	ZMB	Zambia	12,927,000	0	0	0
## 204	ZWE	Zimbabwe	12,571,000	0	0	0

2. What proportion of countries earned no medals? Answer this using *olympic2012*.

```
# Your code here
# Find countries with zero medals
no_medals <- olympic2012$Gold == 0 & olympic2012$Silver == 0 & olympic2012$Bronze == 0

# Calculate the proportion of countries that earned no medals
proportion_no_medals <- sum(no_medals) / nrow(olympic2012)

# Output the proportion
proportion_no_medals
```

```
## [1] 0.5833333
```