

Lab 1

Janine Lim

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You should have RStudio installed to edit this file. You will write code in places marked “TO-DO” to complete the problems. Some of this will be a pure programming assignment. The tools for the solutions to these problems can be found in the class practice lectures. I want you to use the methods I taught you, not for you to google and come up with whatever works. You won’t learn that way.

To “hand in” the homework, you should compile or publish this file into a PDF that includes output of your code. Once it’s done, push by the deadline to your repository in a directory called “labs”.

- Print out the numerical constant pi with ten digits after the decimal point using the internal constant pi.

```
options(digits=11)
pi
```

```
## [1] 3.1415926536
```

- Sum up the first 103 terms of the series $1 + 1/2 + 1/4 + 1/8 + \dots$

```
sum(1/2^(0:102))
```

```
## [1] 2
```

- Find the product of the first 37 terms in the sequence $1/3, 1/6, 1/9 \dots$

```
prod(1/seq(from=3, by = 3, length.out = 37))
```

```
## [1] 1.613528728e-61
```

- Find the product of the first 387 terms of $1 * 1/2 * 1/4 * 1/8 * \dots$

```
prod(1/seq(from = 1, by = 1, length.out = 387))
```

```
## [1] 0
```

```
prod(1/2^(0:386))
```

```
## [1] 0
```

Is this answer *exactly* correct? No - this number will be extremely small and close to 0 but is not 0 because we experienced numerical underflow.

- Figure out a means to express the answer more exactly. Not compute exactly, but express more exactly.

```
-log(2)*sum((0:386))
```

```
## [1] -51771.856063
```

- Create the sequence `x = [Inf, 20, 18, ..., -20]`.

```
x = c(Inf, seq(from = 20, to = -20, by = -2))
x
```

```
## [1] Inf 20 18 16 14 12 10 8 6 4 2 0 -2 -4 -6 -8 -10
## [18] -12 -14 -16 -18 -20
```

Create the sequence `x = [log3(Inf), log3(100), log3(98), ... log3(-20)]`.

```
x = c(Inf, seq(from = 100, to = -20, by = -2))
x=log(x, base = 3)
```

```
## Warning: NaNs produced
```

```
x
```

```
## [1] Inf 4.19180654858 4.17341725189 4.15464876786 4.13548512895
## [6] 4.11590933734 4.09590327429 4.07544759936 4.05452163807 4.03310325630
## [11] 4.01116871959 3.98869253500 3.96564727304 3.94200336639 3.91772888179
## [16] 3.89278926071 3.86714702345 3.84076143031 3.81358809222 3.78557852143
## [21] 3.75667961083 3.72683302786 3.69597450568 3.66403300988 3.63092975357
## [26] 3.59657702662 3.56087679501 3.52371901429 3.48497958377 3.44451784579
## [31] 3.40217350273 3.35776278143 3.31107361282 3.26185950714 3.20983167673
## [36] 3.15464876786 3.09590327429 3.03310325630 2.96564727304 2.89278926071
## [41] 2.81358809222 2.72683302786 2.63092975357 2.52371901429 2.40217350273
## [46] 2.26185950714 2.09590327429 1.89278926071 1.63092975357 1.26185950714
## [51] 0.63092975357 -Inf NaN NaN NaN
## [56] NaN NaN NaN NaN NaN NaN
## [61] NaN NaN
```

Comment on the appropriateness of the non-numeric values.

We get NaN values because we can't take the log₃ of negative numbers and so it becomes a NaN value. We also get Inf values because of the Inf values we have from the original data set. * Create a vector of booleans where the entry is true if `x[i]` is positive and finite.

```
!is.nan(x)
```

```
## [1] TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
## [12] TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
## [23] TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
## [34] TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
## [45] TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE FALSE FALSE FALSE
## [56] FALSE FALSE FALSE FALSE FALSE FALSE FALSE
```

```
is.finite(x)
```

```
## [1] FALSE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
## [12] TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
## [23] TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
## [34] TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
## [45] TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE FALSE FALSE FALSE
## [56] FALSE FALSE FALSE FALSE FALSE FALSE FALSE
```

```
y=!is.nan(x) & is.finite(x) & x>0
```

- Locate the indices of the non-real numbers in this vector. Hint: use the `which` function. Don't hesitate to use the documentation via `?which`.

```
which(y==FALSE)
```

```
## [1] 1 52 53 54 55 56 57 58 59 60 61 62
```

- Locate the indices of the infinite quantities in this vector.

```
which(is.infinite(x))
```

```
## [1] 1 52
```

- Locate the indices of the min and max in this vector. Hint: use the `which.min` and `which.max` functions.

```
which.min(x)
```

```
## [1] 52
```

```
which.max(x)
```

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

- Count the number of unique values in `x`.

```
length(unique(x))
```

```
## [1] 53
```

- Cast `x` to a factor. Do the number of levels make sense? The number of levels make sense because we are trying to cast 53 different numeric values as factor variables so they will each be their own level.

```
as.factor(x)
```

```
## [1] Inf 4.19180654857877 4.1734172518943
## [4] 4.15464876785729 4.13548512895119 4.11590933734319
## [7] 4.09590327428938 4.07544759935851 4.05452163806914
## [10] 4.03310325630434 4.01116871959141 3.98869253500376
## [13] 3.96564727304425 3.94200336638929 3.91772888178973
## [16] 3.89278926071437 3.86714702345081 3.84076143030548
## [19] 3.81358809221559 3.78557852142874 3.75667961082847
## [22] 3.72683302786084 3.69597450568212 3.66403300987579
## [25] 3.63092975357146 3.59657702661571 3.56087679500731
## [28] 3.52371901428583 3.48497958377173 3.44451784578705
## [31] 3.40217350273288 3.3577627814323 3.31107361281783
## [34] 3.26185950714291 3.20983167673402 3.15464876785729
## [37] 3.09590327428938 3.03310325630434 2.96564727304425
## [40] 2.89278926071437 2.8135880922156 2.72683302786084
## [43] 2.63092975357146 2.52371901428583 2.40217350273288
## [46] 2.26185950714291 2.09590327428938 1.89278926071437
## [49] 1.63092975357146 1.26185950714291 0.630929753571457
## [52] -Inf NaN NaN
## [55] NaN NaN NaN
## [58] NaN NaN NaN
## [61] NaN NaN
## 53 Levels: -Inf 0.630929753571457 1.26185950714291 ... NaN
```

- Cast `x` to integers. What do we learn about R's infinity representation in the integer data type? Infinity or -Infinity becomes NA values when we cast `x` to integers.

```
as.integer(x)
```

```
## Warning: NAs introduced by coercion to integer range
```

```
## [1] NA 4 4 4 4 4 4 4 4 4 4 3 3 3 3 3 3 3 3 3 3 3
## [24] 3 3 3 3 3 3 3 3 3 3 3 3 3 3 3 2 2 2 2 2 2 2
## [47] 2 1 1 1 0 NA NA NA NA NA NA NA NA NA NA NA
```

- Use `x` to create a new vector `y` containing only the real numbers in `x`.

```
y=x[!is.nan(x) & is.finite(x) & x>0]
y
```

```
## [1] 4.19180654858 4.17341725189 4.15464876786 4.13548512895 4.11590933734
## [6] 4.09590327429 4.07544759936 4.05452163807 4.03310325630 4.01116871959
## [11] 3.98869253500 3.96564727304 3.94200336639 3.91772888179 3.89278926071
## [16] 3.86714702345 3.84076143031 3.81358809222 3.78557852143 3.75667961083
## [21] 3.72683302786 3.69597450568 3.66403300988 3.63092975357 3.59657702662
## [26] 3.56087679501 3.52371901429 3.48497958377 3.44451784579 3.40217350273
## [31] 3.35776278143 3.31107361282 3.26185950714 3.20983167673 3.15464876786
## [36] 3.09590327429 3.03310325630 2.96564727304 2.89278926071 2.81358809222
## [41] 2.72683302786 2.63092975357 2.52371901429 2.40217350273 2.26185950714
## [46] 2.09590327429 1.89278926071 1.63092975357 1.26185950714 0.63092975357
```

- Use the left rectangle method to numerically integrate x^2 from 0 to 1 with rectangle width size $1e-6$.

```
sum(seq(from=0, to = 1-1e-6, by = 1e-6)^2 * 1e-6)
```

```
## [1] 0.33333283333
```

- Calculate the average of 100 realizations of standard Bernoullis in one line using the `sample` function.

```
mean(sample(c(0,1), size=100, replace=TRUE))
```

```
## [1] 0.55
```

- Calculate the average of 500 realizations of Bernoullis with $p = 0.9$ in one line using the `sample` and `mean` functions.

```
mean(sample(c(0,1), size=500, replace=TRUE, prob = c(0.1, 0.9)))
```

```
## [1] 0.918
```

- Calculate the average of 1000 realizations of Bernoullis with $p = 0.9$ in one line using `rbinom`.

```
?rbinom  
mean(rbinom(n=1000, size = 1, prob = 0.9))
```

```
## [1] 0.896
```

- In class we considered a variable `x_3` which measured “criminality”. We imagined $L = 4$ levels “none”, “infraction”, “misdemeanor” and “felony”. Create a variable `x_3` here with 100 random elements (equally probable). Create it as a nominal (i.e. unordered) factor.

```
x_3=as.factor(sample(c("none", "infraction", "misdemeanor", "felony"), size = 100, replace=TRUE))  
head(x_3)
```

```
## [1] misdemeanor infraction infraction infraction infraction felony  
## Levels: felony infraction misdemeanor none
```

- Use `x_3` to create `x_3_bin`, a binary feature where 0 is no crime and 1 is any crime.

```
x_3_bin= x_3 != "none"
```

- Use `x_3` to create `x_3_ord`, an ordered factor variable. Ensure the proper ordinal ordering.

```
x_3_ord = factor(x_3, levels = c("none", "infraction", "misdemeanor", "felony"), order=TRUE)  
x_3_ord
```

```
## [1] misdemeanor infraction infraction infraction infraction
## [6] felony felony none infraction misdemeanor
## [11] felony infraction felony misdemeanor none
## [16] felony none none infraction felony
## [21] felony misdemeanor infraction misdemeanor felony
## [26] misdemeanor misdemeanor misdemeanor none felony
## [31] infraction infraction felony felony none
## [36] none misdemeanor misdemeanor infraction felony
## [41] none felony infraction misdemeanor misdemeanor
## [46] felony none felony felony felony
## [51] felony felony misdemeanor felony felony
## [56] misdemeanor infraction felony felony felony
## [61] felony infraction infraction none misdemeanor
## [66] infraction infraction infraction none felony
## [71] none felony misdemeanor infraction none
## [76] misdemeanor felony none none misdemeanor
## [81] felony none felony infraction felony
## [86] felony infraction none misdemeanor infraction
## [91] misdemeanor misdemeanor felony none infraction
## [96] misdemeanor misdemeanor infraction none infraction
## Levels: none < infraction < misdemeanor < felony
```

- Convert this variable into three binary variables without any information loss and put them into a data matrix.

```
p=3
n=100
matrix_1 = matrix(nrow = n, ncol = p)
matrix_1[,1]=as.numeric(x_3=="infraction")
matrix_1[,2]=as.numeric(x_3=="misdemeanor")
matrix_1[,3]=as.numeric(x_3=="felony")
colnames(matrix_1) = c("infraction", "misdemeanor", "felony")
matrix_1
```

```
##      infraction misdemeanor felony
## [1,]          0           1      0
## [2,]          1           0      0
## [3,]          1           0      0
## [4,]          1           0      0
## [5,]          1           0      0
## [6,]          0           0      1
## [7,]          0           0      1
## [8,]          0           0      0
## [9,]          1           0      0
## [10,]         0           1      0
## [11,]         0           0      1
## [12,]         1           0      0
## [13,]         0           0      1
## [14,]         0           1      0
## [15,]         0           0      0
## [16,]         0           0      1
## [17,]         0           0      0
## [18,]         0           0      0
```

##	[19,]	1	0	0
##	[20,]	0	0	1
##	[21,]	0	0	1
##	[22,]	0	1	0
##	[23,]	1	0	0
##	[24,]	0	1	0
##	[25,]	0	0	1
##	[26,]	0	1	0
##	[27,]	0	1	0
##	[28,]	0	1	0
##	[29,]	0	0	0
##	[30,]	0	0	1
##	[31,]	1	0	0
##	[32,]	1	0	0
##	[33,]	0	0	1
##	[34,]	0	0	1
##	[35,]	0	0	0
##	[36,]	0	0	0
##	[37,]	0	1	0
##	[38,]	0	1	0
##	[39,]	1	0	0
##	[40,]	0	0	1
##	[41,]	0	0	0
##	[42,]	0	0	1
##	[43,]	1	0	0
##	[44,]	0	1	0
##	[45,]	0	1	0
##	[46,]	0	0	1
##	[47,]	0	0	0
##	[48,]	0	0	1
##	[49,]	0	0	1
##	[50,]	0	0	1
##	[51,]	0	0	1
##	[52,]	0	0	1
##	[53,]	0	1	0
##	[54,]	0	0	1
##	[55,]	0	0	1
##	[56,]	0	1	0
##	[57,]	1	0	0
##	[58,]	0	0	1
##	[59,]	0	0	1
##	[60,]	0	0	1
##	[61,]	0	0	1
##	[62,]	1	0	0
##	[63,]	1	0	0
##	[64,]	0	0	0
##	[65,]	0	1	0
##	[66,]	1	0	0
##	[67,]	1	0	0
##	[68,]	1	0	0
##	[69,]	0	0	0
##	[70,]	0	0	1
##	[71,]	0	0	0
##	[72,]	0	0	1

```
## [73,]      0      1      0
## [74,]      1      0      0
## [75,]      0      0      0
## [76,]      0      1      0
## [77,]      0      0      1
## [78,]      0      0      0
## [79,]      0      0      0
## [80,]      0      1      0
## [81,]      0      0      1
## [82,]      0      0      0
## [83,]      0      0      1
## [84,]      1      0      0
## [85,]      0      0      1
## [86,]      0      0      1
## [87,]      1      0      0
## [88,]      0      0      0
## [89,]      0      1      0
## [90,]      1      0      0
## [91,]      0      1      0
## [92,]      0      1      0
## [93,]      0      0      1
## [94,]      0      0      0
## [95,]      1      0      0
## [96,]      0      1      0
## [97,]      0      1      0
## [98,]      1      0      0
## [99,]      0      0      0
## [100,]     1      0      0
```

- What should the sum of each row be (in English)?

The sum of each row should be 0 or 1 because each value in the matrix should only be a 0 or 1 value, and each row should only have at most one 1 value for either infraction, misdemeanor, or felony. If they're all 0's, then the sum would be 0 and that would mean the person has committed no crimes.

Verify that.

```
rowSums(matrix_1)
```

```
## [1] 1 1 1 1 1 1 1 0 1 1 1 1 1 1 0 1 0 0 1 1 1 1 1 1 1 1 1 0 1 1 1 1 1 0
## [36] 0 1 1 1 1 0 1 1 1 1 1 0 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 0 1 1 1 1 0 1
## [71] 0 1 1 1 0 1 1 0 0 1 1 0 1 1 1 1 1 0 1 1 1 1 1 0 1 1 1 1 0 1
```

- How should the column sum look (in English)? Each column sum should be the sum of people who have committed an infraction, then the sum of people who have committed a misdemeanor, and the third sum should be the number of people who have committed a felony. The column sums altogether should add up to the number of people who have committed an infraction, misdemeanor, and felony.

#TO-DO

Verify that.


```
colSums(matrix_1)
```

```
##   infraction misdemeanor      felony
##           25           23           33
```

- Generate a matrix with 100 rows where the first column is realization from a normal with mean 17 and variance 38, the second column is uniform between -10 and 10, the third column is poisson with mean 6, the fourth column in exponential with lambda of 9, the fifth column is binomial with $n = 20$ and $p = 0.12$ and the sixth column is a binary variable with exactly 24% 1's dispersed randomly. Name the rows the entries of the `fake_first_names` vector.

```
fake_first_names = c(
  "Sophia", "Emma", "Olivia", "Ava", "Mia", "Isabella", "Riley",
  "Aria", "Zoe", "Charlotte", "Lily", "Layla", "Amelia", "Emily",
  "Madelyn", "Aubrey", "Adalyn", "Madison", "Chloe", "Harper",
  "Abigail", "Aaliyah", "Avery", "Evelyn", "Kaylee", "Ella", "Ellie",
  "Scarlett", "Arianna", "Hailey", "Nora", "Addison", "Brooklyn",
  "Hannah", "Mila", "Leah", "Elizabeth", "Sarah", "Eliana", "Mackenzie",
  "Peyton", "Maria", "Grace", "Adeline", "Elena", "Anna", "Victoria",
  "Camilla", "Lillian", "Natalie", "Jackson", "Aiden", "Lucas",
  "Liam", "Noah", "Ethan", "Mason", "Caden", "Oliver", "Elijah",
  "Grayson", "Jacob", "Michael", "Benjamin", "Carter", "James",
  "Jayden", "Logan", "Alexander", "Caleb", "Ryan", "Luke", "Daniel",
  "Jack", "William", "Owen", "Gabriel", "Matthew", "Connor", "Jayce",
  "Isaac", "Sebastian", "Henry", "Muhammad", "Cameron", "Wyatt",
  "Dylan", "Nathan", "Nicholas", "Julian", "Eli", "Levi", "Isaiah",
  "Landon", "David", "Christian", "Andrew", "Brayden", "John",
  "Lincoln"
)
n = 100
p = 6
p1 = rnorm(n, mean = 17, sd = sqrt(38))
p2 = runif(n, min = -10, max = 10)
p3 = rpois(n, lambda = 6)
p4 = rexp(n, rate = 9)
p5 = rbinom(n, size = 20, prob = 0.12)
p6 = sample(rep(c(0,1), times = c(0.76*100, 0.24*100)), size = 100)
matrix_2 = matrix(c(p1, p2, p3, p4, p5, p6), n,p )
rownames(matrix_2)=fake_first_names
matrix_2
```

```
##           [,1]           [,2] [,3]           [,4] [,5] [,6]
## Sophia    21.3169727761 -0.096975346096    6 0.1068797099579    2    0
## Emma      19.6198830027  4.636536929756    5 0.0741292919136    3    1
## Olivia    11.4343812761  2.113138409331    5 0.1063200732026    1    0
## Ava       13.1021756953 -5.972205088474    4 0.0020688884817    5    0
## Mia       20.9698482808  8.777321930975    6 0.0496336620094    6    0
## Isabella   9.8480239700 -2.403197004460    0 0.0668853885598    3    1
## Riley     12.0741826443  8.093195031397    7 0.0642986858471    3    0
## Aria      19.3809076164 -7.162421280518   10 0.0840790860299    1    0
## Zoe        9.1963560853 -4.968998203985    8 0.1344232374963    1    0
## Charlotte 24.2542887551  6.605690754950    7 0.0181559816831    4    1
```

## Lily	12.6642489987	-2.173726879992	3	0.0597835512728	0	1
## Layla	21.9711575140	5.695860465057	8	0.1058250007132	5	0
## Amelia	20.0552294910	-0.374140175991	9	0.0510217141774	0	0
## Emily	9.6600227750	6.487084217370	6	0.0897818012116	3	0
## Madelyn	3.0008517254	-5.008912477642	9	0.1421927944193	2	0
## Aubrey	21.7108066494	-6.453156713396	9	0.0230014395072	3	0
## Adalyn	18.6232831435	7.391738751903	6	0.0557242075188	3	1
## Madison	22.6253124456	6.307426267304	6	0.1946437120758	1	0
## Chloe	9.1896896669	-7.077128943056	13	0.1559690630177	2	0
## Harper	14.5667366457	-5.746584222652	11	0.0830368822920	4	1
## Abigail	25.2478406522	-1.097740624100	10	0.0553441676829	5	0
## Aaliyah	21.9932940940	-3.232096279971	1	0.0629020892601	0	0
## Avery	12.1001894112	5.891855424270	6	0.0409601369562	2	0
## Evelyn	5.8618105567	0.567268277518	5	0.0648080106411	1	1
## Kaylee	15.8999223073	-2.418528804556	3	0.0499507752765	1	0
## Ella	22.4294133675	-3.776327325031	8	0.1000748329968	2	0
## Ellie	21.7752955201	9.434805074707	9	0.0018543044312	2	1
## Scarlett	5.2005529174	1.837662965991	5	0.0668993279752	2	0
## Arianna	25.6853270281	7.104822369292	3	0.0221808305439	2	0
## Hailey	17.7612387500	6.188832842745	11	0.2230191181010	0	1
## Nora	21.2480955629	-2.247545211576	9	0.2496409703734	5	0
## Addison	20.9457173028	-5.913596018218	5	0.2275897082262	1	1
## Brooklyn	14.6920121551	3.842462515458	3	0.1312907578170	1	0
## Hannah	18.0931991882	7.779223225079	6	0.0501357404929	2	0
## Mila	25.6525701116	-3.447828916833	5	0.0745909092948	1	0
## Leah	28.3350072010	8.807859770022	6	0.0482411537733	3	1
## Elizabeth	18.2701073594	7.505804151297	5	0.1213401187579	4	0
## Sarah	15.2972454977	6.026624310762	2	0.0488914123012	3	0
## Eliana	16.0286752011	-0.080457162112	4	0.2455739356023	1	0
## Mackenzie	22.2788053146	2.029335345142	9	0.3341913269475	0	0
## Peyton	10.9834109863	-1.177033935674	6	0.2274919268064	1	0
## Maria	20.7519528273	-0.098207639530	5	0.0301265249339	3	0
## Grace	14.6748630711	-7.974054622464	7	0.2355312620334	2	0
## Adeline	30.5604613402	8.335438342765	3	0.1295297257556	0	0
## Elena	28.6921833274	-2.852512164973	6	0.0095686897015	0	1
## Anna	8.0666221326	-3.210826884024	5	0.1556745243890	2	0
## Victoria	19.0686174386	-5.464843013324	6	0.0656315433379	5	0
## Camilla	24.2618287090	-4.660149314441	10	0.0308884397770	0	0
## Lillian	18.3301898105	-5.143688586541	2	0.0658115846001	2	0
## Natalie	27.0082787691	9.865225846879	10	0.0209475377471	3	1
## Jackson	26.1657126455	-7.269071363844	4	0.0337962295717	1	0
## Aiden	17.1527375217	-7.118680868298	7	0.0313095301448	1	0
## Lucas	19.0324436566	6.190170319751	4	0.0365863330145	1	0
## Liam	20.4698411702	6.511795665137	2	0.0140938474279	2	0
## Noah	20.8543005519	-5.909203304909	11	0.0499005918908	2	0
## Ethan	24.7485919448	-4.411310036667	6	0.0352618707771	5	0
## Mason	14.5484875133	-0.694920876995	6	0.2426496859129	1	0
## Caden	22.0404089354	1.062555839308	11	0.0512335383230	1	0
## Oliver	26.8917653152	-7.123933881521	10	0.1245190988559	3	1
## Elijah	23.9539813629	8.581551155075	7	0.0617041711489	2	0
## Grayson	14.4697878275	0.627228887752	7	0.1779532418415	1	1
## Jacob	17.4500475029	7.996172895655	10	0.2254431394560	2	0
## Michael	10.4960139692	-0.887140790001	8	0.1431217168722	2	0
## Benjamin	21.7626198636	-2.217747964896	5	0.1201887754677	3	0

## Carter	17.2666232121	1.130856024101	6	0.0185544244014	2	0
## James	15.5565211965	-8.435157751665	7	0.1339013988312	3	0
## Jayden	30.2636740205	9.609402297065	8	0.0920863127410	3	0
## Logan	22.5064479176	-0.656472281553	10	0.0123666708047	4	1
## Alexander	20.1061886577	6.263528424315	5	0.0512501986490	2	0
## Caleb	22.3765053657	2.730376068503	5	0.0845444917425	1	1
## Ryan	19.6908940988	6.782528911717	4	0.0279928166109	0	0
## Luke	6.4923264266	-8.859407706186	5	0.0822349184719	1	0
## Daniel	7.8478859219	4.927779510617	12	0.0119340078802	1	1
## Jack	15.2170071341	1.240738867782	4	0.0592677213976	6	0
## William	1.6020565062	-9.623404447921	2	0.2273875358939	6	0
## Owen	28.2860719476	1.493492764421	7	0.2286465372797	2	0
## Gabriel	11.3819359452	-8.743146401830	6	0.2022763538184	1	0
## Matthew	13.9465103707	-3.232774757780	5	0.1997518982231	4	0
## Connor	13.3103153410	-8.933860198595	5	0.0571907323061	4	1
## Jayce	19.3237194694	-5.443692808039	7	0.6661271502201	0	0
## Isaac	19.4659059627	-7.388098626398	2	0.1095398988373	1	0
## Sebastian	20.1672088738	-1.625044718385	7	0.0852538415495	3	0
## Henry	18.9097693100	7.118848874234	6	0.0975014184451	5	0
## Muhammad	18.4318262725	0.551281501539	12	0.3012923629194	2	1
## Cameron	21.9160602969	-9.236788768321	8	0.1603627314432	4	0
## Wyatt	11.8696840199	2.797510963865	7	0.0010312741824	4	0
## Dylan	9.4168019538	3.513419907540	6	0.0242790063947	4	1
## Nathan	18.8956792917	-5.019447100349	4	0.0933227507972	4	1
## Nicholas	15.5632242475	-1.909407330677	3	0.0562213121706	4	0
## Julian	15.1909890355	-9.438405893743	8	0.2917465736849	4	0
## Eli	11.0013998895	0.831363312900	6	0.0276798165397	3	0
## Levi	15.9267721327	-2.942378316075	6	0.0572537430045	3	0
## Isaiah	24.3775285922	-0.051264031790	4	0.1807487938046	4	0
## Landon	9.8060973955	8.938338411972	6	0.0808855425520	2	0
## David	15.7471039166	-2.004809840582	10	0.0754230283201	3	1
## Christian	18.5970988937	6.026346068829	3	0.1907861232017	1	0
## Andrew	4.8151344971	-6.303680115379	4	0.0244682714240	1	0
## Brayden	12.4629593121	-7.597714364529	9	0.1814626277228	2	0
## John	22.7932166083	-0.708377817646	9	0.0014854661810	3	0
## Lincoln	18.0021816198	-2.585894037038	11	0.8179805999652	1	1

- Create a data frame of the same data as above except make the binary variable a factor “DOMESTIC” vs “FOREIGN” for 0 and 1 respectively. Use RStudio’s View function to ensure this worked as desired.

```
X_4 = data.frame(matrix_2)
X_4$p6_cat = factor(p6, labels = c("DOMESTIC", "FOREIGN"))
##View(X_4) I couldn't knit the file unless I took this view function out- sorry!
```

- Print out a table of the binary variable. Then print out the proportions of “DOMESTIC” vs “FOREIGN”.

```
table(X_4$p6_cat)
```

```
##
## DOMESTIC FOREIGN
##          76      24
```

```
table(X_4$p6_cat)/100
```

```
##
## DOMESTIC FOREIGN
##      0.76      0.24
```

Print out a summary of the whole dataframe.

```
summary(X_4)
```

```
##           X1                X2                X3
## Min.      : 1.6020565   Min.      :-9.62340445   Min.      : 0.0
## 1st Qu.:13.2582804   1st Qu.: -5.01154613   1st Qu.: 5.0
## Median :18.5144626   Median : -0.67569658   Median : 6.0
## Mean      :17.6102715   Mean      :-0.10353619   Mean      : 6.4
## 3rd Qu.:21.9298346   3rd Qu.: 5.92547809   3rd Qu.: 8.0
## Max.      :30.5604613   Max.      : 9.86522585   Max.      :13.0
##           X4                X5                X6                p6_cat
## Min.      :0.0010312742   Min.      :0.00   Min.      :0.00   DOMESTIC:76
## 1st Qu.:0.0487288477   1st Qu.:1.00   1st Qu.:0.00   FOREIGN :24
## Median :0.0750069688   Median :2.00   Median :0.00
## Mean      :0.1122053568   Mean      :2.33   Mean      :0.24
## 3rd Qu.:0.1462599187   3rd Qu.:3.00   3rd Qu.:0.00
## Max.      :0.8179806000   Max.      :6.00   Max.      :1.00
```

- Let $n = 50$. Create a $n \times n$ matrix R of exactly 50% entries 0's, 25% 1's 25% 2's. These values should be in random locations.

```
n=50
```

```
R = matrix(sample(rep(c(0:2), times = c(0.50*2500, 0.25*2500, 0.25*2500))), size = n*n, nrow = n, ncol = n)
table(R)
```

```
## R
##    0    1    2
## 1250 625 625
```

- Randomly punch holes (i.e. NA) values in this matrix so that an each entry is missing with probability 30%.

```
holes = matrix(nrow = n, ncol = n, sample(c(rep(0, n*n*0.7), rep(3, n*n*0.3))))
for(i in 1:n){
  for(j in 1:n){
    if(holes[i,j]==3){
      R[i,j]=NA
    }
  }
}
R
```

##		[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]	[,8]	[,9]	[,10]	[,11]	[,12]	[,13]
##	[1,]	1	0	2	1	0	0	0	0	NA	2	2	NA	0
##	[2,]	NA	1	1	1	NA	0	2	0	0	1	0	0	NA
##	[3,]	NA	2	NA	2	NA	0	0	0	NA	NA	NA	NA	0
##	[4,]	0	2	2	NA	NA	1	0	1	NA	2	1	NA	2
##	[5,]	0	0	2	0	2	1	NA	2	0	NA	NA	NA	NA
##	[6,]	0	NA	0	2	2	1	NA	NA	NA	0	NA	NA	2
##	[7,]	0	NA	0	1	NA	0	1	0	1	NA	1	NA	NA
##	[8,]	0	2	0	NA	0	2	0	NA	0	NA	0	0	2
##	[9,]	0	NA	NA	NA	NA	2	0	0	2	2	1	NA	NA
##	[10,]	1	0	0	2	NA	0	2	NA	NA	1	2	NA	NA
##	[11,]	2	2	2	1	0	NA	2	NA	1	1	0	1	NA
##	[12,]	0	0	2	0	NA	1	2	2	2	NA	NA	1	NA
##	[13,]	0	0	1	NA	1	0	NA	NA	0	0	2	NA	0
##	[14,]	0	NA	0	0	NA	1	0	2	0	NA	0	0	2
##	[15,]	NA	NA	0	2	0	0	2	1	2	2	0	NA	NA
##	[16,]	2	0	0	2	0	1	2	1	0	1	0	0	1
##	[17,]	0	2	1	0	0	0	NA	2	0	0	NA	NA	1
##	[18,]	0	NA	NA	0	0	NA	NA	NA	0	0	1	NA	0
##	[19,]	1	0	NA	NA	NA	1	0	NA	0	1	1	0	NA
##	[20,]	0	1	1	NA	0	2	1	0	NA	NA	NA	0	1
##	[21,]	NA	0	0	NA	NA	2	0	1	0	NA	2	NA	1
##	[22,]	2	NA	NA	2	1	0	0	1	2	1	1	NA	2
##	[23,]	0	0	0	1	1	1	2	1	0	1	0	0	NA
##	[24,]	1	1	0	1	1	0	NA	NA	0	NA	1	0	NA
##	[25,]	2	0	1	1	0	0	NA	0	0	2	NA	2	2
##	[26,]	1	NA	0	0	NA	2	NA	1	0	0	0	1	0
##	[27,]	NA	0	NA	0	1	0	0	1	NA	0	2	1	2
##	[28,]	NA	1	1	1	1	0	NA	NA	0	0	NA	0	0
##	[29,]	NA	0	NA	0	0	0	0	0	0	0	NA	0	0
##	[30,]	1	NA	NA	0	0	2	1	NA	0	NA	0	1	0
##	[31,]	2	NA	1	0	2	0	NA	2	1	2	NA	2	NA
##	[32,]	1	1	0	NA	0	1	0	2	0	NA	NA	0	NA
##	[33,]	1	0	NA	0	0	NA	1	0	2	0	0	0	NA
##	[34,]	1	0	0	2	1	2	NA	2	2	0	1	NA	NA
##	[35,]	2	2	2	0	1	NA	0	NA	0	0	2	0	NA
##	[36,]	0	NA	NA	2	1	NA	2	NA	NA	1	0	2	2
##	[37,]	0	NA	0	2	1	0	0	2	1	1	1	0	NA
##	[38,]	0	0	NA	2	0	NA	0	0	0	1	NA	0	2
##	[39,]	NA	0	NA	NA	2	NA	NA	NA	NA	1	0	1	NA
##	[40,]	NA	2	1	0	2	2	0	NA	0	NA	2	0	2
##	[41,]	0	0	0	NA	2	1	0	0	0	0	1	0	0
##	[42,]	0	0	2	2	1	0	0	0	0	NA	0	0	1
##	[43,]	NA	0	NA	1	1	NA	2	0	0	2	2	0	0
##	[44,]	0	NA	2	NA	NA	0	NA	NA	NA	1	1	0	0
##	[45,]	1	0	0	0	1	2	0	0	0	NA	2	0	1
##	[46,]	2	2	0	0	NA	1	NA	NA	NA	NA	0	NA	NA
##	[47,]	0	1	0	NA	0	0	1	1	0	0	0	1	0
##	[48,]	NA	0	NA	1	NA	2	0	0	NA	2	0	NA	1
##	[49,]	1	0	1	0	0	2	2	2	0	1	1	2	0
##	[50,]	NA	1	0	NA	0	2	2	0	NA	NA	0	NA	2
##		[,14]	[,15]	[,16]	[,17]	[,18]	[,19]	[,20]	[,21]	[,22]	[,23]	[,24]		
##	[1,]	0	0	2	1	0	NA	1	1	NA	NA	0		
##	[2,]	0	2	2	1	NA	NA	0	0	NA	NA	0		

##	[3,]	0	0	0	0	1	NA	0	2	1	0	1
##	[4,]	1	NA	0	NA	NA	0	NA	0	0	0	0
##	[5,]	1	2	0	0	0	1	NA	2	NA	0	0
##	[6,]	NA	NA	1	2	2	0	0	0	0	NA	1
##	[7,]	NA	2	0	0	0	0	1	0	NA	1	0
##	[8,]	0	NA	0	2	0	1	2	NA	2	2	2
##	[9,]	0	NA	2	0	1	0	0	2	1	NA	1
##	[10,]	NA	0	2	2	0	2	NA	1	0	2	NA
##	[11,]	0	0	NA	0	0	NA	0	NA	0	2	0
##	[12,]	NA	0	2	NA	NA	2	0	1	0	0	1
##	[13,]	0	2	1	2	NA	0	0	NA	2	NA	2
##	[14,]	2	0	0	NA	0	0	2	2	0	NA	0
##	[15,]	2	0	2	0	NA	NA	NA	0	NA	NA	NA
##	[16,]	NA	2	NA	0	0	NA	NA	NA	NA	NA	0
##	[17,]	2	NA	NA	0	2	0	NA	NA	NA	NA	2
##	[18,]	2	1	0	NA	0	2	0	NA	0	1	2
##	[19,]	NA	NA	1	1	1	1	0	0	2	2	NA
##	[20,]	2	NA	NA	0	2	1	NA	0	0	NA	1
##	[21,]	NA	2	0	NA	2	NA	1	0	1	NA	2
##	[22,]	NA	0	2	NA	0	2	0	1	0	1	1
##	[23,]	2	NA	NA	0	2	0	NA	0	NA	0	1
##	[24,]	1	NA	1	0	1	2	NA	2	1	0	NA
##	[25,]	2	1	NA	0	0	1	0	1	2	NA	NA
##	[26,]	NA	1	NA	1	0	NA	2	1	NA	NA	NA
##	[27,]	0	0	0	0	2	0	0	0	NA	1	NA
##	[28,]	1	0	0	NA	1	2	0	0	2	2	2
##	[29,]	0	2	2	0	2	NA	2	0	0	0	2
##	[30,]	0	0	0	0	2	NA	0	0	NA	NA	NA
##	[31,]	2	1	1	0	NA	2	0	NA	NA	NA	0
##	[32,]	0	0	0	1	2	NA	2	NA	NA	0	1
##	[33,]	0	1	0	0	0	0	0	NA	0	2	2
##	[34,]	2	0	0	NA	1	0	2	2	NA	NA	NA
##	[35,]	1	0	2	0	NA	0	NA	1	0	0	0
##	[36,]	2	NA	1	0	1	NA	1	2	1	0	0
##	[37,]	0	2	0	0	2	NA	NA	0	0	2	1
##	[38,]	0	NA	0	0	2	0	NA	0	1	0	1
##	[39,]	2	0	NA	1	2	2	NA	NA	0	0	NA
##	[40,]	NA	1	NA	2	0	NA	0	0	0	2	NA
##	[41,]	1	2	2	2	0	NA	0	2	0	NA	1
##	[42,]	0	0	0	0	0	NA	1	1	0	1	0
##	[43,]	0	0	NA	0	NA	2	NA	0	1	NA	0
##	[44,]	1	0	NA	2	NA	0	2	NA	2	NA	0
##	[45,]	0	0	0	1	NA	0	1	NA	0	0	2
##	[46,]	0	2	NA	2	NA	NA	2	2	0	NA	1
##	[47,]	1	2	NA	1	NA	1	0	NA	1	0	0
##	[48,]	2	NA	2	0	NA	1	NA	1	1	NA	2
##	[49,]	0	NA	0	0	0	NA	NA	2	0	NA	NA
##	[50,]	1	NA	2	NA	0	1	0	NA	NA	NA	0
##		[,25]	[,26]	[,27]	[,28]	[,29]	[,30]	[,31]	[,32]	[,33]	[,34]	[,35]
##	[1,]	1	1	0	0	NA	2	2	2	0	0	2
##	[2,]	NA	0	0	0	0	NA	NA	2	2	NA	1
##	[3,]	0	2	NA	0	NA	0	NA	0	1	2	2
##	[4,]	0	2	0	1	NA	0	NA	NA	0	NA	NA
##	[5,]	NA	NA	1	NA	0	NA	NA	1	NA	2	2

##	[6,]	NA	0	1	0	NA	0	1	0	1	1	0
##	[7,]	0	NA	0	1	NA	1	1	1	2	0	0
##	[8,]	2	1	NA	NA	0	NA	2	NA	0	2	0
##	[9,]	0	NA	2	0	0	NA	0	NA	1	NA	1
##	[10,]	NA	0	2	0	2	NA	2	0	NA	0	0
##	[11,]	NA	2	NA	2	NA	1	1	0	0	0	NA
##	[12,]	0	1	0	0	2	2	0	2	0	0	NA
##	[13,]	1	2	0	NA	2	0	0	0	NA	2	1
##	[14,]	2	2	NA	NA	2	1	NA	2	0	1	1
##	[15,]	1	NA	1	0	NA	2	2	NA	0	0	NA
##	[16,]	2	0	0	2	0	NA	NA	0	1	1	NA
##	[17,]	NA	2	2	NA	NA	2	2	2	0	1	2
##	[18,]	0	NA	0	0	0	1	NA	2	1	0	1
##	[19,]	0	2	1	0	2	2	0	0	2	1	NA
##	[20,]	NA	1	2	NA	0	1	NA	1	NA	1	0
##	[21,]	0	2	1	0	NA	NA	0	NA	NA	0	1
##	[22,]	1	NA	0	1	1	0	0	2	2	0	NA
##	[23,]	0	2	2	0	NA	0	NA	0	0	2	NA
##	[24,]	0	2	0	NA	0	2	1	NA	NA	NA	0
##	[25,]	NA	1	0	0	NA	1	0	0	1	0	0
##	[26,]	NA	0	1	1	0	NA	0	NA	2	NA	0
##	[27,]	0	NA	NA	2	0	2	0	0	1	NA	NA
##	[28,]	1	0	NA	NA	NA	2	0	2	NA	NA	NA
##	[29,]	2	NA	NA	1	0	0	0	0	0	NA	NA
##	[30,]	NA	NA	0	1	NA	0	1	NA	2	0	NA
##	[31,]	NA	NA	0	0	0	0	0	0	NA	1	1
##	[32,]	2	0	2	1	1	NA	NA	0	NA	NA	1
##	[33,]	0	0	2	2	NA	NA	1	NA	0	2	2
##	[34,]	2	0	2	0	2	0	1	NA	NA	NA	1
##	[35,]	NA	0	NA	0	2	2	NA	NA	0	NA	0
##	[36,]	1	2	1	NA	NA	NA	0	2	1	2	NA
##	[37,]	0	0	1	2	2	0	0	NA	NA	1	NA
##	[38,]	0	NA	0	2	1	0	2	0	NA	0	0
##	[39,]	NA	1	0	NA	NA	0	NA	0	NA	0	1
##	[40,]	0	NA	1	2	NA	2	NA	0	1	0	0
##	[41,]	NA	0	1	0	NA	NA	2	0	NA	NA	NA
##	[42,]	0	1	NA	2	2	NA	2	2	NA	NA	NA
##	[43,]	1	NA	0	2	NA	NA	NA	0	0	2	NA
##	[44,]	0	0	0	1	NA	1	NA	2	1	0	NA
##	[45,]	NA	0	1	NA	1	0	1	NA	NA	2	0
##	[46,]	NA	0	2	2	0	NA	1	NA	NA	0	2
##	[47,]	0	1	NA	NA	1	NA	NA	NA	0	0	0
##	[48,]	2	0	0	0	NA	1	1	NA	2	0	0
##	[49,]	NA	NA	1	2	0	1	0	NA	0	NA	NA
##	[50,]	0	NA	1	NA	NA	1	0	0	1	NA	2
##		[,36]	[,37]	[,38]	[,39]	[,40]	[,41]	[,42]	[,43]	[,44]	[,45]	[,46]
##	[1,]	NA	2	2	0	0	NA	0	0	NA	1	NA
##	[2,]	1	0	0	0	1	0	2	2	0	2	0
##	[3,]	0	2	1	NA	NA	0	1	1	1	2	0
##	[4,]	2	1	0	NA	2	NA	1	2	1	NA	NA
##	[5,]	0	0	0	0	0	NA	NA	0	1	1	NA
##	[6,]	0	0	1	0	NA	NA	0	NA	1	1	2
##	[7,]	NA	NA	1	1	0	1	NA	0	0	1	NA
##	[8,]	0	0	1	NA	NA	0	1	NA	0	0	0

##	[9,]	NA	0	NA	0	2	0	2	0	1	NA	0
##	[10,]	0	0	NA	0	1	1	1	NA	NA	0	2
##	[11,]	0	0	1	0	1	1	0	1	0	0	2
##	[12,]	0	1	0	0	1	NA	NA	1	NA	NA	1
##	[13,]	0	0	2	NA	NA	0	0	0	1	0	2
##	[14,]	NA	NA	1	2	0	2	2	0	0	NA	1
##	[15,]	0	0	2	0	2	NA	0	0	0	NA	NA
##	[16,]	NA	NA	2	NA	0	NA	NA	NA	0	NA	0
##	[17,]	0	0	1	1	0	NA	1	2	NA	1	NA
##	[18,]	2	1	NA	NA	1	1	0	0	NA	0	1
##	[19,]	2	2	2	1	NA	NA	0	NA	2	NA	NA
##	[20,]	2	2	NA	1	2	0	1	0	2	1	2
##	[21,]	0	0	NA	NA	0	NA	NA	0	1	NA	NA
##	[22,]	1	0	NA	0	NA	NA	0	0	1	0	0
##	[23,]	NA	NA	0	NA	0	2	1	0	0	0	0
##	[24,]	NA	1	1	2	2	0	0	0	2	NA	NA
##	[25,]	NA	1	1	2	1	NA	NA	2	2	0	1
##	[26,]	0	0	0	0	0	1	0	NA	NA	0	NA
##	[27,]	2	1	0	NA	1	0	0	NA	1	NA	0
##	[28,]	NA	1	0	0	2	NA	NA	0	NA	NA	NA
##	[29,]	NA	0	NA	0	2	2	2	2	2	0	2
##	[30,]	2	0	0	0	NA	2	NA	NA	2	NA	2
##	[31,]	1	1	NA	NA	1	0	0	0	0	NA	NA
##	[32,]	NA	NA	NA	1	1	1	NA	0	0	0	0
##	[33,]	0	1	2	1	NA	2	NA	2	0	NA	NA
##	[34,]	NA	1	1	NA	1	2	NA	NA	1	NA	0
##	[35,]	2	1	1	2	2	0	2	0	NA	2	NA
##	[36,]	0	0	NA	0	0	NA	NA	NA	2	0	1
##	[37,]	2	0	1	0	0	0	NA	0	NA	1	0
##	[38,]	1	0	NA	2	0	0	NA	1	2	0	2
##	[39,]	NA	NA	0	0	NA	0	0	NA	NA	1	NA
##	[40,]	2	0	0	2	NA	NA	NA	NA	1	0	NA
##	[41,]	0	NA	0	0	NA	NA	NA	2	1	1	NA
##	[42,]	NA	2	NA	2	1	NA	NA	NA	0	0	NA
##	[43,]	NA	2	1	2	1	1	0	1	2	NA	2
##	[44,]	0	0	0	2	0	2	NA	0	0	NA	0
##	[45,]	1	0	1	NA	NA	NA	NA	2	NA	NA	0
##	[46,]	0	NA	0	2	2	0	2	0	1	NA	NA
##	[47,]	NA	NA	2	1	0	1	1	0	1	0	NA
##	[48,]	2	0	NA	2	NA	0	2	0	0	NA	0
##	[49,]	0	0	0	NA	0	2	0	0	0	1	NA
##	[50,]	1	NA	1	0	0	NA	2	1	NA	0	0
##	[,47] [,48] [,49] [,50]											
##	[1,]	NA	NA	0	1							
##	[2,]	0	0	1	2							
##	[3,]	0	NA	0	0							
##	[4,]	NA	2	0	2							
##	[5,]	2	NA	2	0							
##	[6,]	NA	1	1	NA							
##	[7,]	0	0	NA	NA							
##	[8,]	0	2	0	2							
##	[9,]	0	0	NA	0							
##	[10,]	0	NA	NA	NA							
##	[11,]	NA	1	0	0							


```
## [12,] 0 NA 2 NA
## [13,] 2 NA NA 0
## [14,] NA 1 0 2
## [15,] 0 1 1 0
## [16,] NA NA NA 0
## [17,] 0 NA 0 NA
## [18,] 0 1 0 2
## [19,] NA NA 1 1
## [20,] 0 0 2 NA
## [21,] NA NA 2 0
## [22,] NA 2 2 NA
## [23,] 1 NA NA 2
## [24,] 2 1 1 0
## [25,] 1 0 1 NA
## [26,] 1 0 0 0
## [27,] 0 1 0 1
## [28,] 0 0 0 0
## [29,] NA 2 NA 0
## [30,] 2 1 NA 0
## [31,] NA 0 0 NA
## [32,] NA NA NA NA
## [33,] NA NA 1 0
## [34,] 2 1 NA NA
## [35,] 2 1 1 1
## [36,] NA NA 0 NA
## [37,] 1 2 0 2
## [38,] 0 1 2 0
## [39,] NA 2 2 0
## [40,] 1 NA 0 0
## [41,] 2 2 2 0
## [42,] 0 NA NA 0
## [43,] 2 0 NA 2
## [44,] 2 0 0 0
## [45,] 0 NA 0 NA
## [46,] 0 1 2 NA
## [47,] NA 0 NA NA
## [48,] 2 NA 1 2
## [49,] 0 1 0 NA
## [50,] 0 NA NA 2
```

- Sort the rows in matrix R by the largest row sum to lowest. Be careful about the NA's!

```
order(rowSums(R, na.rm=TRUE), decreasing=TRUE)
```

```
## [1] 34 35 14 19 20 22 25 43 48 17 46 8 24 36 37 1 12 4 10 13 40 2 11
## [24] 29 33 41 5 15 38 3 6 23 9 18 31 42 50 27 28 30 44 49 21 32 16 45
## [47] 7 39 47 26
```

- We will now learn the `apply` function. This is a handy function that saves writing for loops which should be eschewed in R. Use the `apply` function to compute a vector whose entries are the standard deviation of each row. Use the `apply` function to compute a vector whose entries are the standard deviation of each column. Be careful about the NA's! This should be one line.

```
apply(R, MARGIN = 1, sd, na.rm=TRUE)
```

```
## [1] 0.85215816722 0.83205029434 0.82807867121 0.87066900492 0.86711818075
## [6] 0.75996059566 0.61220087877 0.93455856719 0.84281592351 0.90558034297
## [11] 0.79979754523 0.85588532090 0.89459504822 0.90556993049 0.90243777591
## [16] 0.84418225411 0.89928422716 0.75833704583 0.79716520920 0.80622577483
## [21] 0.84077140277 0.82285973943 0.82381956682 0.77459666924 0.79046271831
## [26] 0.66219534414 0.75807647576 0.81649658093 0.95600222596 0.85901293693
## [31] 0.80950789391 0.74775650111 0.87078025831 0.82686886579 0.89410905932
## [36] 0.84002688129 0.83738850732 0.83300806660 0.83757892854 0.90354816528
## [41] 0.87423435890 0.83816526318 0.88687914726 0.84316331143 0.73906595605
## [46] 0.93094933625 0.61220087877 0.88687914726 0.79789512885 0.81517858720
```

```
apply(R, MARGIN = 2, sd, na.rm=TRUE)
```

```
## [1] 0.78072439645 0.82836355919 0.82182530102 0.85512059455 0.74907350181
## [6] 0.85301950530 0.90632696717 0.84492824744 0.76523564056 0.78363384079
## [11] 0.81066855082 0.71771928199 0.89201957643 0.87376286169 0.89796949077
## [16] 0.90792308282 0.80546911040 0.89794560320 0.85901293693 0.84861216259
## [21] 0.85588532090 0.77459666924 0.89155582824 0.82285973943 0.82227511432
## [26] 0.87926630988 0.80024034851 0.87679459896 0.90486632647 0.84440066184
## [31] 0.81867681600 0.93642615266 0.79176634141 0.85208592300 0.80455691406
## [36] 0.88963130018 0.74721705905 0.75037528148 0.88616323851 0.81096094706
## [41] 0.84497248158 0.84497248158 0.83198088451 0.78933142388 0.68144538746
## [46] 0.90155858477 0.89810654458 0.76341057035 0.84091786587 0.90551882884
```

- Use the `apply` function to compute a vector whose entries are the count of entries that are 1 or 2 in each column. This should be one line.

```
apply(R>0, MARGIN = 2, sum, na.rm=TRUE)
```

```
## [1] 18 15 17 21 19 23 16 18 10 21 21 11 18 21 17 18 16 19 17 14 19 15 13
## [24] 22 14 19 22 19 14 20 18 14 18 17 17 16 17 22 18 21 16 16 14 23 13 14
## [47] 15 20 18 14
```

- Use the `split` function to create a list whose keys are the column number and values are the vector of the columns. Look at the last example in the documentation `?split`.

```
split(R, col(R))
```

```
## $'1'
## [1] 1 NA NA 0 0 0 0 0 0 1 2 0 0 0 NA 2 0 0 1 0 NA 2 0
## [24] 1 2 1 NA NA NA 1 2 1 1 1 2 0 0 0 NA NA 0 0 NA 0 1 2
## [47] 0 NA 1 NA
##
## $'2'
## [1] 0 1 2 2 0 NA NA 2 NA 0 2 0 0 NA NA 0 2 NA 0 1 0 NA 0
## [24] 1 0 NA 0 1 0 NA NA 1 0 0 2 NA NA 0 0 2 0 0 0 NA 0 2
## [47] 1 0 0 1
##
```

```

## $'3'
## [1] 2 1 NA 2 2 0 0 0 NA 0 2 2 1 0 0 0 1 NA NA 1 0 NA 0
## [24] 0 1 0 NA 1 NA NA 1 0 NA 0 2 NA 0 NA NA 1 0 2 NA 2 0 0
## [47] 0 NA 1 0
##
## $'4'
## [1] 1 1 2 NA 0 2 1 NA NA 2 1 0 NA 0 2 2 0 0 NA NA NA 2 1
## [24] 1 1 0 0 1 0 0 0 NA 0 2 0 2 2 2 NA 0 NA 2 1 NA 0 0
## [47] NA 1 0 NA
##
## $'5'
## [1] 0 NA NA NA 2 2 NA 0 NA NA 0 NA 1 NA 0 0 0 0 NA 0 NA 1 1
## [24] 1 0 NA 1 1 0 0 2 0 0 1 1 1 1 0 2 2 2 1 1 NA 1 NA
## [47] 0 NA 0 0
##
## $'6'
## [1] 0 0 0 1 1 1 0 2 2 0 NA 1 0 1 0 1 0 NA 1 2 2 0 1
## [24] 0 0 2 0 0 0 2 0 1 NA 2 NA NA 0 NA NA 2 1 0 NA 0 2 1
## [47] 0 2 2 2
##
## $'7'
## [1] 0 2 0 0 NA NA 1 0 0 2 2 2 NA 0 2 2 NA NA 0 1 0 0 2
## [24] NA NA NA 0 NA 0 1 NA 0 1 NA 0 2 0 0 NA 0 0 0 2 NA 0 NA
## [47] 1 0 2 2
##
## $'8'
## [1] 0 0 0 1 2 NA 0 NA 0 NA NA 2 NA 2 1 1 2 NA NA 0 1 1 1
## [24] NA 0 1 1 NA 0 NA 2 2 0 2 NA NA 2 0 NA NA 0 0 0 NA 0 NA
## [47] 1 0 2 0
##
## $'9'
## [1] NA 0 NA NA 0 NA 1 0 2 NA 1 2 0 0 2 0 0 0 0 NA 0 2 0
## [24] 0 0 0 NA 0 0 0 1 0 2 2 0 NA 1 0 NA 0 0 0 0 NA 0 NA
## [47] 0 NA 0 NA
##
## $'10'
## [1] 2 1 NA 2 NA 0 NA NA 2 1 1 NA 0 NA 2 1 0 0 1 NA NA 1 1
## [24] NA 2 0 0 0 0 NA 2 NA 0 0 0 1 1 1 1 NA 0 NA 2 1 NA NA
## [47] 0 2 1 NA
##
## $'11'
## [1] 2 0 NA 1 NA NA 1 0 1 2 0 NA 2 0 0 0 NA 1 1 NA 2 1 0
## [24] 1 NA 0 2 NA NA 0 NA NA 0 1 2 0 1 NA 0 2 1 0 2 1 2 0
## [47] 0 0 1 0
##
## $'12'
## [1] NA 0 NA NA NA NA NA 0 NA NA 1 1 NA 0 NA 0 NA NA 0 0 NA NA 0
## [24] 0 2 1 1 0 0 1 2 0 0 NA 0 2 0 0 1 0 0 0 0 0 0 NA
## [47] 1 NA 2 NA
##
## $'13'
## [1] 0 NA 0 2 NA 2 NA 2 NA NA NA NA 0 2 NA 1 1 0 NA 1 1 2 NA
## [24] NA 2 0 2 0 0 0 NA NA NA NA NA 2 NA 2 NA 2 0 1 0 0 1 NA
## [47] 0 1 0 2

```

```

##
## $'14'
## [1] 0 0 0 1 1 NA NA 0 0 NA 0 NA 0 2 2 NA 2 2 NA 2 NA NA 2
## [24] 1 2 NA 0 1 0 0 2 0 0 2 1 2 0 0 2 NA 1 0 0 1 0 0
## [47] 1 2 0 1
##
## $'15'
## [1] 0 2 0 NA 2 NA 2 NA NA 0 0 0 2 0 0 2 NA 1 NA NA 2 0 NA
## [24] NA 1 1 0 0 2 0 1 0 1 0 0 NA 2 NA 0 1 2 0 0 0 0 2
## [47] 2 NA NA NA
##
## $'16'
## [1] 2 2 0 0 0 1 0 0 2 2 NA 2 1 0 2 NA NA 0 1 NA 0 2 NA
## [24] 1 NA NA 0 0 2 0 1 0 0 0 2 1 0 0 NA NA 2 0 NA NA 0 NA
## [47] NA 2 0 2
##
## $'17'
## [1] 1 1 0 NA 0 2 0 2 0 2 0 NA 2 NA 0 0 0 NA 1 0 NA NA 0
## [24] 0 0 1 0 NA 0 0 0 1 0 NA 0 0 0 0 1 2 2 0 0 2 1 2
## [47] 1 0 0 NA
##
## $'18'
## [1] 0 NA 1 NA 0 2 0 0 1 0 0 NA NA 0 NA 0 2 0 1 2 2 0 2
## [24] 1 0 0 2 1 2 2 NA 2 0 1 NA 1 2 2 2 0 0 0 NA NA NA NA
## [47] NA NA 0 0
##
## $'19'
## [1] NA NA NA 0 1 0 0 1 0 2 NA 2 0 0 NA NA 0 2 1 1 NA 2 0
## [24] 2 1 NA 0 2 NA NA 2 NA 0 0 0 NA NA 0 2 NA NA NA 2 0 0 NA
## [47] 1 1 NA 1
##
## $'20'
## [1] 1 0 0 NA NA 0 1 2 0 NA 0 0 0 2 NA NA NA 0 0 NA 1 0 NA
## [24] NA 0 2 0 0 2 0 0 2 0 2 NA 1 NA NA NA 0 0 1 NA 2 1 2
## [47] 0 NA NA 0
##
## $'21'
## [1] 1 0 2 0 2 0 0 NA 2 1 NA 1 NA 2 0 NA NA NA 0 0 0 1 0
## [24] 2 1 1 0 0 0 0 NA NA NA 2 1 2 0 0 NA 0 2 1 0 NA NA 2
## [47] NA 1 2 NA
##
## $'22'
## [1] NA NA 1 0 NA 0 NA 2 1 0 0 0 2 0 NA NA NA 0 2 0 1 0 NA
## [24] 1 2 NA NA 2 0 NA NA NA 0 NA 0 1 0 1 0 0 0 0 1 2 0 0
## [47] 1 1 0 NA
##
## $'23'
## [1] NA NA 0 0 0 NA 1 2 NA 2 2 0 NA NA NA NA NA 1 2 NA NA 1 0
## [24] 0 NA NA 1 2 0 NA NA 0 2 NA 0 0 2 0 0 2 NA 1 NA NA 0 NA
## [47] 0 NA NA NA
##
## $'24'
## [1] 0 0 1 0 0 1 0 2 1 NA 0 1 2 0 NA 0 2 2 NA 1 2 1 1
## [24] NA NA NA NA 2 2 NA 0 1 2 NA 0 0 1 1 NA NA 1 0 0 0 2 1

```

```

## [47] 0 2 NA 0
##
## $'25'
## [1] 1 NA 0 0 NA NA 0 2 0 NA NA 0 1 2 1 2 NA 0 0 NA 0 1 0
## [24] 0 NA NA 0 1 2 NA NA 2 0 2 NA 1 0 0 NA 0 NA 0 1 0 NA NA
## [47] 0 2 NA 0
##
## $'26'
## [1] 1 0 2 2 NA 0 NA 1 NA 0 2 1 2 2 NA 0 2 NA 2 1 2 NA 2
## [24] 2 1 0 NA 0 NA NA NA 0 0 0 0 2 0 NA 1 NA 0 1 NA 0 0 0
## [47] 1 0 NA NA
##
## $'27'
## [1] 0 0 NA 0 1 1 0 NA 2 2 NA 0 0 NA 1 0 2 0 1 2 1 0 2
## [24] 0 0 1 NA NA NA 0 0 2 2 2 NA 1 1 0 0 1 1 NA 0 0 1 2
## [47] NA 0 1 1
##
## $'28'
## [1] 0 0 0 1 NA 0 1 NA 0 0 2 0 NA NA 0 2 NA 0 0 NA 0 1 0
## [24] NA 0 1 2 NA 1 1 0 1 2 0 0 NA 2 2 NA 2 0 2 2 1 NA 2
## [47] NA 0 2 NA
##
## $'29'
## [1] NA 0 NA NA 0 NA NA 0 0 2 NA 2 2 2 NA 0 NA 0 2 0 NA 1 NA
## [24] 0 NA 0 0 NA 0 NA 0 1 NA 2 2 NA 2 1 NA NA NA 2 NA NA 1 0
## [47] 1 NA 0 NA
##
## $'30'
## [1] 2 NA 0 0 NA 0 1 NA NA NA 1 2 0 1 2 NA 2 1 2 1 NA 0 0
## [24] 2 1 NA 2 2 0 0 0 NA NA 0 2 NA 0 0 0 2 NA NA NA 1 0 NA
## [47] NA 1 1 1
##
## $'31'
## [1] 2 NA NA NA NA 1 1 2 0 2 1 0 0 NA 2 NA 2 NA 0 NA 0 0 NA
## [24] 1 0 0 0 0 0 1 0 NA 1 1 NA 0 0 2 NA NA 2 2 NA NA 1 1
## [47] NA 1 0 0
##
## $'32'
## [1] 2 2 0 NA 1 0 1 NA NA 0 0 2 0 2 NA 0 2 2 0 1 NA 2 0
## [24] NA 0 NA 0 2 0 NA 0 0 NA NA NA 2 NA 0 0 0 0 2 0 2 NA NA
## [47] NA NA NA 0
##
## $'33'
## [1] 0 2 1 0 NA 1 2 0 1 NA 0 0 NA 0 0 1 0 1 2 NA NA 2 0
## [24] NA 1 2 1 NA 0 2 NA NA 0 NA 0 1 NA NA NA 1 NA NA 0 1 NA NA
## [47] 0 2 0 1
##
## $'34'
## [1] 0 NA 2 NA 2 1 0 2 NA 0 0 0 2 1 0 1 1 0 1 1 0 0 2
## [24] NA 0 NA NA NA NA 0 1 NA 2 NA NA 2 1 0 0 0 NA NA 2 0 2 0
## [47] 0 0 NA NA
##
## $'35'
## [1] 2 1 2 NA 2 0 0 0 1 0 NA NA 1 1 NA NA 2 1 NA 0 1 NA NA

```

```

## [24] 0 0 0 NA NA NA NA 1 1 2 1 0 NA NA 0 1 0 NA NA NA NA 0 2
## [47] 0 0 NA 2
##
## $'36'
## [1] NA 1 0 2 0 0 NA 0 NA 0 0 0 0 NA 0 NA 0 2 2 2 0 1 NA
## [24] NA NA 0 2 NA NA 2 1 NA 0 NA 2 0 2 1 NA 2 0 NA NA 0 1 0
## [47] NA 2 0 1
##
## $'37'
## [1] 2 0 2 1 0 0 NA 0 0 0 0 1 0 NA 0 NA 0 1 2 2 0 0 NA
## [24] 1 1 0 1 1 0 0 1 NA 1 1 1 0 0 0 NA 0 NA 2 2 0 0 NA
## [47] NA 0 0 NA
##
## $'38'
## [1] 2 0 1 0 0 1 1 1 NA NA 1 0 2 1 2 2 1 NA 2 NA NA NA 0
## [24] 1 1 0 0 0 NA 0 NA NA 2 1 1 NA 1 NA 0 0 0 NA 1 0 1 0
## [47] 2 NA 0 1
##
## $'39'
## [1] 0 0 NA NA 0 0 1 NA 0 0 0 0 NA 2 0 NA 1 NA 1 1 NA 0 NA
## [24] 2 2 0 NA 0 0 0 NA 1 1 NA 2 0 0 2 0 2 0 2 2 NA 2
## [47] 1 2 NA 0
##
## $'40'
## [1] 0 1 NA 2 0 NA 0 NA 2 1 1 1 NA 0 2 0 0 1 NA 2 0 NA 0
## [24] 2 1 0 1 2 2 NA 1 1 NA 1 2 0 0 0 NA NA NA 1 1 0 NA 2
## [47] 0 NA 0 0
##
## $'41'
## [1] NA 0 0 NA NA NA 1 0 0 1 1 NA 0 2 NA NA NA 1 NA 0 NA NA 2
## [24] 0 NA 1 0 NA 2 2 0 1 2 2 0 NA 0 0 0 NA NA NA 1 2 NA 0
## [47] 1 0 2 NA
##
## $'42'
## [1] 0 2 1 1 NA 0 NA 1 2 1 0 NA 0 2 0 NA 1 0 0 1 NA 0 1
## [24] 0 NA 0 0 NA 2 NA 0 NA NA NA 2 NA NA NA 0 NA NA NA 0 NA NA 2
## [47] 1 2 0 2
##
## $'43'
## [1] 0 2 1 2 0 NA 0 NA 0 NA 1 1 0 0 0 NA 2 0 NA 0 0 0 0
## [24] 0 2 NA NA 0 2 NA 0 0 2 NA 0 NA 0 1 NA NA 2 NA 1 0 2 0
## [47] 0 0 0 1
##
## $'44'
## [1] NA 0 1 1 1 1 0 0 1 NA 0 NA 1 0 0 0 NA NA 2 2 1 1 0
## [24] 2 2 NA 1 NA 2 2 0 0 0 1 NA 2 NA 2 NA 1 1 0 2 0 NA 1
## [47] 1 0 0 NA
##
## $'45'
## [1] 1 2 2 NA 1 1 1 0 NA 0 0 NA 0 NA NA NA 1 0 NA 1 NA 0 0
## [24] NA 0 0 NA NA 0 NA NA 0 NA NA 2 0 1 0 1 0 1 0 NA NA NA NA
## [47] 0 NA 1 0
##
## $'46'

```

```
## [1] NA 0 0 NA NA 2 NA 0 0 2 2 1 2 1 NA 0 NA 1 NA 2 NA 0 0
## [24] NA 1 NA 0 NA 2 2 NA 0 NA 0 NA 1 0 2 NA NA NA NA 2 0 0 NA
## [47] NA 0 NA 0
##
## $'47'
## [1] NA 0 0 NA 2 NA 0 0 0 0 NA 0 2 NA 0 NA 0 0 NA 0 NA NA 1
## [24] 2 1 1 0 0 NA 2 NA NA NA 2 2 NA 1 0 NA 1 2 0 2 2 0 0
## [47] NA 2 0 0
##
## $'48'
## [1] NA 0 NA 2 NA 1 0 2 0 NA 1 NA NA 1 1 NA NA 1 NA 0 NA 2 NA
## [24] 1 0 0 1 0 2 1 0 NA NA 1 1 NA 2 1 2 NA 2 NA 0 0 NA 1
## [47] 0 NA 1 NA
##
## $'49'
## [1] 0 1 0 0 2 1 NA 0 NA NA 0 2 NA 0 1 NA 0 0 1 2 2 2 NA
## [24] 1 1 0 0 0 NA NA 0 NA 1 NA 1 0 0 2 2 0 2 NA NA 0 0 2
## [47] NA 1 0 NA
##
## $'50'
## [1] 1 2 0 2 0 NA NA 2 0 NA 0 NA 0 2 0 0 NA 2 1 NA 0 NA 2
## [24] 0 NA 0 1 0 0 0 NA NA 0 NA 1 NA 2 0 0 0 0 0 2 0 NA NA
## [47] NA 2 NA 2
```

- In one statement, use the `lapply` function to create a list whose keys are the column number and values are themselves a list with keys: “min” whose value is the minimum of the column, “max” whose value is the maximum of the column, “pct_missing” is the proportion of missingness in the column and “first_NA” whose value is the row number of the first time the NA appears.

```
lapply(split(R, col(R)), function(x){ c(min = min(x, na.rm = TRUE),
                                         max = max(x, na.rm=TRUE),
                                         pct_missing = sum(is.na(x))/length(x),
                                         first_NA = which.min(is.na(x)))})
```

```
## $'1'
##      min      max pct_missing first_NA
##    0.00    2.00      0.24      1.00
##
## $'2'
##      min      max pct_missing first_NA
##    0.00    2.00      0.26      1.00
##
## $'3'
##      min      max pct_missing first_NA
##    0.00    2.00      0.28      1.00
##
## $'4'
##      min      max pct_missing first_NA
##    0.00    2.00      0.26      1.00
##
## $'5'
##      min      max pct_missing first_NA
##    0.00    2.00      0.28      1.00
```

```

##
## $'6'
##      min      max pct_missing  first_NA
##      0.00      2.00      0.16      1.00
##
## $'7'
##      min      max pct_missing  first_NA
##      0.00      2.00      0.28      1.00
##
## $'8'
##      min      max pct_missing  first_NA
##      0.00      2.00      0.32      1.00
##
## $'9'
##      min      max pct_missing  first_NA
##      0.00      2.00      0.26      2.00
##
## $'10'
##      min      max pct_missing  first_NA
##      0.00      2.00      0.32      1.00
##
## $'11'
##      min      max pct_missing  first_NA
##      0.00      2.00      0.24      1.00
##
## $'12'
##      min      max pct_missing  first_NA
##      0.00      2.00      0.36      2.00
##
## $'13'
##      min      max pct_missing  first_NA
##      0.00      2.00      0.38      1.00
##
## $'14'
##      min      max pct_missing  first_NA
##      0.0      2.0      0.2      1.0
##
## $'15'
##      min      max pct_missing  first_NA
##      0.00      2.00      0.28      1.00
##
## $'16'
##      min      max pct_missing  first_NA
##      0.00      2.00      0.26      1.00
##
## $'17'
##      min      max pct_missing  first_NA
##      0.00      2.00      0.18      1.00
##
## $'18'
##      min      max pct_missing  first_NA
##      0.00      2.00      0.26      1.00
##
## $'19'

```


##	min	max	pct_missing	first_NA
##	0.00	2.00	0.36	4.00
##				
##	\$'20'			
##	min	max	pct_missing	first_NA
##	0.00	2.00	0.32	1.00
##				
##	\$'21'			
##	min	max	pct_missing	first_NA
##	0.00	2.00	0.28	1.00
##				
##	\$'22'			
##	min	max	pct_missing	first_NA
##	0.0	2.0	0.3	3.0
##				
##	\$'23'			
##	min	max	pct_missing	first_NA
##	0.00	2.00	0.46	3.00
##				
##	\$'24'			
##	min	max	pct_missing	first_NA
##	0.00	2.00	0.24	1.00
##				
##	\$'25'			
##	min	max	pct_missing	first_NA
##	0.00	2.00	0.34	1.00
##				
##	\$'26'			
##	min	max	pct_missing	first_NA
##	0.0	2.0	0.3	1.0
##				
##	\$'27'			
##	min	max	pct_missing	first_NA
##	0.0	2.0	0.2	1.0
##				
##	\$'28'			
##	min	max	pct_missing	first_NA
##	0.00	2.00	0.26	1.00
##				
##	\$'29'			
##	min	max	pct_missing	first_NA
##	0.00	2.00	0.44	2.00
##				
##	\$'30'			
##	min	max	pct_missing	first_NA
##	0.00	2.00	0.32	1.00
##				
##	\$'31'			
##	min	max	pct_missing	first_NA
##	0.00	2.00	0.32	1.00
##				
##	\$'32'			
##	min	max	pct_missing	first_NA
##	0.00	2.00	0.34	1.00

```

##
## $'33'
##      min      max pct_missing  first_NA
##      0.00      2.00      0.34      1.00
##
## $'34'
##      min      max pct_missing  first_NA
##      0.0      2.0      0.3      1.0
##
## $'35'
##      min      max pct_missing  first_NA
##      0.00      2.00      0.38      1.00
##
## $'36'
##      min      max pct_missing  first_NA
##      0.00      2.00      0.32      2.00
##
## $'37'
##      min      max pct_missing  first_NA
##      0.0      2.0      0.2      1.0
##
## $'38'
##      min      max pct_missing  first_NA
##      0.00      2.00      0.26      1.00
##
## $'39'
##      min      max pct_missing  first_NA
##      0.00      2.00      0.26      1.00
##
## $'40'
##      min      max pct_missing  first_NA
##      0.00      2.00      0.26      1.00
##
## $'41'
##      min      max pct_missing  first_NA
##      0.00      2.00      0.38      2.00
##
## $'42'
##      min      max pct_missing  first_NA
##      0.00      2.00      0.38      1.00
##
## $'43'
##      min      max pct_missing  first_NA
##      0.00      2.00      0.26      1.00
##
## $'44'
##      min      max pct_missing  first_NA
##      0.00      2.00      0.24      2.00
##
## $'45'
##      min      max pct_missing  first_NA
##      0.0      2.0      0.4      1.0
##
## $'46'

```

```
##           min           max pct_missing first_NA
##          0.00           2.00         0.42      2.00
##
## $'47'
##           min           max pct_missing first_NA
##          0.00           2.00         0.32      2.00
##
## $'48'
##           min           max pct_missing first_NA
##          0.00           2.00         0.38      2.00
##
## $'49'
##           min           max pct_missing first_NA
##          0.00           2.00         0.28      1.00
##
## $'50'
##           min           max pct_missing first_NA
##          0.00           2.00         0.32      1.00
```

- Set a seed and then create a vector `v` consisting of a sample of 1,000 iid normal realizations with mean -10 and variance 100.

```
set.seed(11)
v=rnorm(1000, mean=-10, sd=10)
head(v)
```

```
## [1] -15.9103110258 -9.7340563098 -25.1655309708 -23.6265334930
## [5]  1.7848915603 -19.3415131967
```

- Repeat this exercise by resetting the seed to ensure you obtain the same results.

```
set.seed(11)
v=rnorm(1000, mean=-10, sd=10)
head(v)
```

```
## [1] -15.9103110258 -9.7340563098 -25.1655309708 -23.6265334930
## [5]  1.7848915603 -19.3415131967
```

- Find the average of `v` and the standard error of `v`.

```
mean(v)
```

```
## [1] -9.9120889267
```

```
SE = sd(v)/sqrt(1000)
SE
```

```
## [1] 0.3150349092
```

- Find the 5%ile of `v` and use the `qnorm` function to compute what it theoretically should be. Is the estimate about what is expected by theory?

Yes this is to be expected because v is made up of realization from $\text{norm}(\text{mean}=-10, \text{sd}=10)$, and should be close to the 5th percentile of $\text{norm}(\text{mean}=-10, \text{sd}=10)$.

```
quantile(v, probs = 0.05)
```

```
##           5%  
## -26.366352937
```

```
qnorm(0.05, mean=-10, sd=10)
```

```
## [1] -26.44853627
```

- What is the percentile of v that corresponds to the value 0? What should it be theoretically? Is the estimate about what is expected by theory?

This is also to be expected by theory because v is made up of realizations from $\text{norm}(\text{mean}=-10, \text{sd}=10)$, and so should have approximately the same CDF as $\text{norm}(\text{mean}=-10, \text{sd}=10)$.

```
ecdf(v)(0)
```

```
## [1] 0.845
```

```
pnorm(0, mean=-10, sd=10)
```

```
## [1] 0.84134474607
```