

Foundations of Probability in R

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Calculating probabilities

You're in charge of the sales team, and it's time for performance reviews, starting with Amir. As part of the review, you want to randomly select a few of the deals that he's worked on over the past year so that you can look at them more deeply. Before you start selecting deals, you'll first figure out what the chances are of selecting certain deals.

Instructions

- Count the number of deals Amir worked on for each product type.
- Create a new column called prob by dividing n by the total number of deals Amir worked on.
- QUESTION 1: If you randomly select one of Amir's deals, what's the probability that the deal will involve Product C?

```
library(dplyr)
```

```
## Warning: package 'dplyr' was built under R version 4.3.3
```

```
##
```

```
## Attaching package: 'dplyr'
```

```
## The following objects are masked from 'package:stats':
```

```
##
```

```
##      filter, lag
```

```
## The following objects are masked from 'package:base':
```

```
##
```

```
##      intersect, setdiff, setequal, union
```

```
# amir_deals <- read.csv(file.choose(), header = T)
```

```
amir_deals <- read.csv("amir_deals.csv")
```

```
head(amir_deals, 10)
```

```
##      X  product client status  amount num_users
## 1    1 Product F Current   Won 7389.52        19
## 2    2 Product C      New   Won 4493.01        43
## 3    3 Product B      New   Won 5738.09        87
## 4    4 Product I Current   Won 2591.24        83
## 5    5 Product E Current   Won 6622.97        17
## 6    6 Product B      New   Won 5496.27         2
## 7    7 Product C Current   Won 3043.13        29
## 8    8 Product N Current   Won 7340.64        13
## 9    9 Product F Current   Won 6780.85        80
## 10 10 Product B Current   Won 5237.24        23
```

```
# Count the deals for each product
amir_deals %>%
  count(product)
```

```
##      product  n
## 1 Product A 23
## 2 Product B 62
## 3 Product C 15
## 4 Product D 40
## 5 Product E  5
## 6 Product F 11
## 7 Product G  2
## 8 Product H  8
## 9 Product I  7
## 10 Product J  2
## 11 Product N  3
```

```
# Calculate probability of picking a deal with each product
amir_deals %>%
  count(product) %>%
  mutate(prob = n/sum(n))
```

```
##      product  n      prob
## 1 Product A 23 0.12921348
## 2 Product B 62 0.34831461
## 3 Product C 15 0.08426966
## 4 Product D 40 0.22471910
## 5 Product E  5 0.02808989
## 6 Product F 11 0.06179775
## 7 Product G  2 0.01123596
## 8 Product H  8 0.04494382
## 9 Product I  7 0.03932584
## 10 Product J  2 0.01123596
## 11 Product N  3 0.01685393
```

Sampling deals

In the previous exercise, you counted the deals Amir worked on. Now it's time to randomly pick five deals so that you can reach out to each customer and ask if they were satisfied with the service they received. You'll try doing this both with and without replacement.

Additionally, you want to make sure this is done randomly and that it can be reproduced in case you get asked how you chose the deals, so you'll need to set the random seed before sampling from the deals.

dplyr is loaded and amir_deals is available.

Instructions

- Set the random seed to 31
- Take a sample of 5 deals without replacement
- Take a sample of 5 deals with replacement.
- QUESTION 2: What type of sampling is better to use for this situation?

```
# Set random seed to 31
set.seed(31)
```

```
# Sample 5 deals without replacement
amir_deals %>%
  sample_n(5)

##      X  product client status  amount num_users
## 1 173 Product D Current   Lost 3086.88         55
## 2  49 Product C Current   Lost 3727.66         19
## 3  75 Product D Current   Lost 4274.80          9
## 4 168 Product B Current    Won 4965.08          9
## 5  43 Product A Current    Won 5827.35         50

# Set random seed to 31
set.seed(31)

# Sample 5 deals with replacement
amir_deals %>%
  sample_n(5, replace = TRUE)

##      X  product client status  amount num_users
## 1 173 Product D Current   Lost 3086.88         55
## 2  49 Product C Current   Lost 3727.66         19
## 3  75 Product D Current   Lost 4274.80          9
## 4 168 Product B Current    Won 4965.08          9
## 5  43 Product A Current    Won 5827.35         50

#sample_n(5)
```

Creating a probability distribution

A new restaurant opened a few months ago, and the restaurant's management wants to optimize its seating space based on the size of the groups that come most often. On one night, there are 10 groups of people waiting to be seated at the restaurant, but instead of being called in the order they arrived, they will be called randomly. In this exercise, you'll investigate the probability of groups of different sizes getting picked first. Data on each of the ten groups is contained in the `restaurant_groups` data frame.

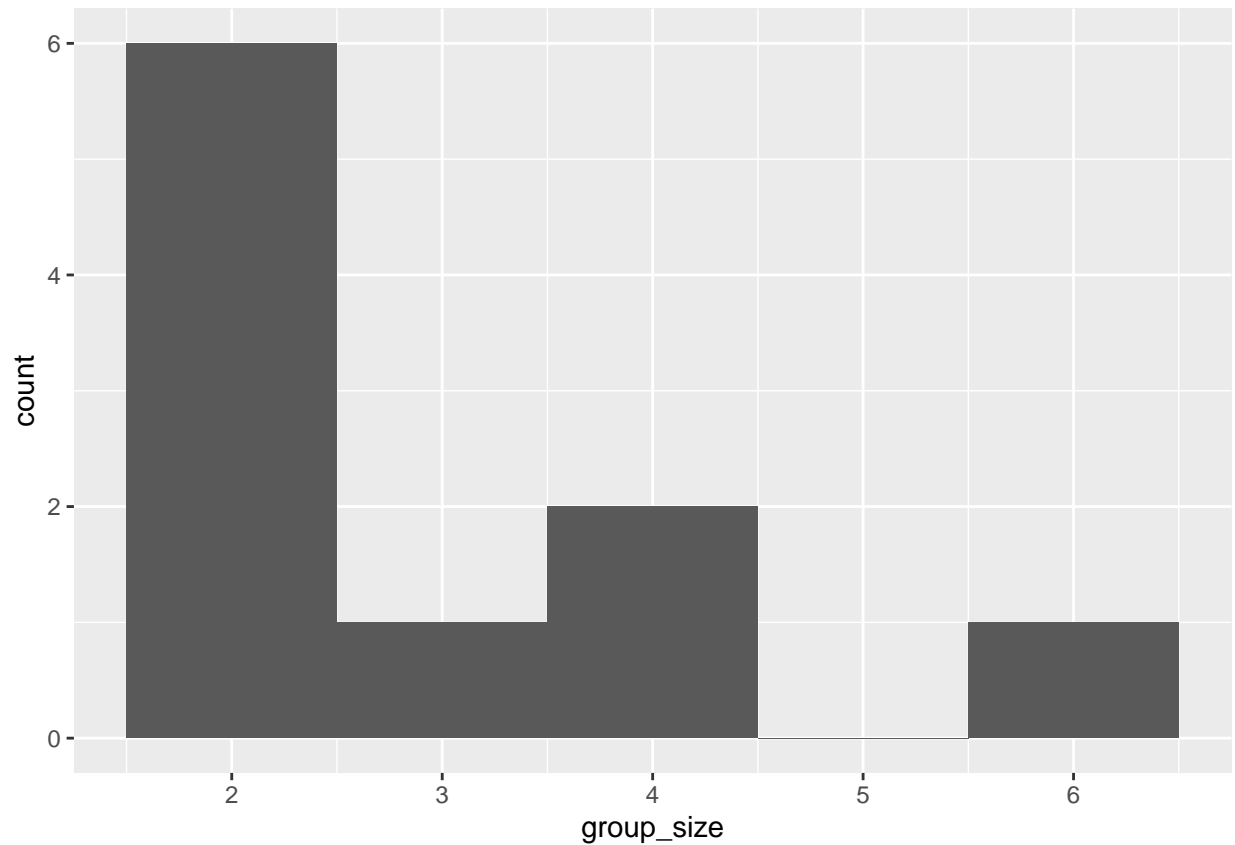
Remember that expected value can be calculated by multiplying each possible outcome with its corresponding probability and taking the sum. The `restaurant_groups` data is available and `dplyr` and `ggplot2` are loaded.

Instructions

- Create a histogram of the `group_size` column of `restaurant_groups`, setting the number of bins to 5.
- Count the number of each `group_size` in `restaurant_groups`, then add a column called `probability` that contains the probability of randomly selecting a group of each size. Store this in a new data frame called `size_distribution`.
- Calculate the expected value of the `size_distribution`, which represents the expected group size.
- Calculate the probability of randomly picking a group of 4 or more people by filtering and summarizing.

```
library(ggplot2)

## Warning: package 'ggplot2' was built under R version 4.3.3
restaurant_groups <- read.csv("restaurant_groups.csv")
# Create a histogram of restaurant_groups
ggplot(restaurant_groups, aes(group_size)) +
  geom_histogram(bins = 5)
```



```
# Create probability distribution
size_distribution <- restaurant_groups %>%
  # Count number of each group size
  count(group_size) %>%
  # Calculate probability
  mutate(probability = n / sum(n))
```

```
size_distribution
```

```
##   group_size n probability
## 1          2 6          0.6
## 2          3 1          0.1
## 3          4 2          0.2
## 4          6 1          0.1
```

```
# Create probability distribution
size_distribution <- restaurant_groups %>%
  count(group_size) %>%
  mutate(probability = n / sum(n))
```

```
# Calculate expected group size
expected_val <- sum(size_distribution$group_size *
  size_distribution$probability)
expected_val
```

```
## [1] 2.9
```

```

# Create probability distribution
size_distribution <- restaurant_groups %>%
  count(group_size) %>%
  mutate(probability = n / sum(n))

# Calculate probability of picking group of 4 or more
size_distribution %>%
  # Filter for groups of 4 or larger
  filter(group_size >= 4) %>%
  # Calculate prob_4_or_more by taking sum of probabilities
  summarise(prob_4_or_above = sum(probability))

```

```

##   prob_4_or_above
## 1              0.3

```

```

restaurant_groups%>%
  count(group_size)%>%
  mutate(prob = n/sum(n)) %>%
  select(-n)

```

```

##   group_size prob
## 1          2  0.6
## 2          3  0.1
## 3          4  0.2
## 4          6  0.1

```

```

size_distribution <- restaurant_groups %>%
  count(group_size) %>%
  mutate(probability = n / sum(n)) %>%
  select(-n)
size_distribution

```

```

##   group_size probability
## 1          2          0.6
## 2          3          0.1
## 3          4          0.2
## 4          6          0.1

```

```

# Calculate probability of picking group of 4 or more
x=size_distribution %>%
  # Filter for groups of 4 or larger
  filter(group_size >= 4)
  # Calculate prob_4_or_more by taking sum of probabilities
  sum(x$probability)

```

```

## [1] 0.3

```

Data back-ups

The sales software used at your company is set to automatically back itself up, but no one knows exactly what time the back-ups happen. It is known, however, that back-ups happen exactly every 30 minutes. Amir comes back from sales meetings at random times to update the data on the client he just met with. He wants to know how long he'll have to wait for his newly-entered data to get backed up. Use your new knowledge of continuous uniform distributions to model this situation and answer Amir's questions.

Instructions

- To model how long Amir will wait for a back-up using a continuous uniform distribution, save his lowest possible wait time as min and his longest possible wait time as max. Remember that back-ups happen every 30 minutes.

```
# Min and max wait times for back-up that happens every 30 min
min <- 0
max <- 30
```

- Calculate the probability that Amir has to wait less than 5 minutes, and store in a variable called prob_less_than_5.

```
# Min and max wait times for back-up that happens every 30 min
min <- 0
max <- 30

# Calculate probability of waiting less than 5 mins
prob_less_than_5 <- punif(5, min, max)
prob_less_than_5

## [1] 0.1666667
```

```
?punif()
```

```
## starting httpd help server ... done
```

- Calculate the probability that Amir has to wait more than 5 minutes, and store in a variable called prob_greater_than_5.

```
# Min and max wait times for back-up that happens every 30 min
min <- 0
max <- 30

# Calculate probability of waiting more than 5 mins
prob_greater_than_5 <- punif(5, min, max, lower.tail = FALSE)
prob_greater_than_5

## [1] 0.8333333
```

- Calculate the probability that Amir has to wait between 10 and 20 minutes, and store in a variable called prob_between_10_and_20.

```
# Min and max wait times for back-up that happens every 30 min
min <- 0
max <- 30

# Calculate probability of waiting 10-20 mins
prob_between_10_and_20 <- punif(20, min, max) - punif(10, min, max)
prob_between_10_and_20

## [1] 0.3333333
```

Simulating wait times

To give Amir a better idea of how long he'll have to wait, you'll simulate Amir waiting 1000 times and create a histogram to show him what he should expect. Recall from the last exercise that his minimum wait time is 0 minutes and his maximum wait time is 30 minutes.

A data frame called wait_times is available and dplyr and ggplot2 are loaded

- Set the random seed to 334

```
# Set random seed to 334
set.seed(334)
```

- Generate 1000 wait times from the continuous uniform distribution that models Amir's wait time. Add this as a new column called time in the wait_times data frame.

```
# Set random seed to 334
set.seed(334)

# Generate 1000 wait times between 0 and 30 mins, save in time column
wait_times <- read.csv("wait_times.csv")
wait_times %>%
  mutate(time = runif(1000, min = 0, max = 30))
```

##	simulation_nb	time
## 1	1	29.45315459
## 2	2	16.29111785
## 3	3	0.36924658
## 4	4	24.18677361
## 5	5	23.42601313
## 6	6	15.92492170
## 7	7	17.54207142
## 8	8	2.81082236
## 9	9	1.22953709
## 10	10	18.81465174
## 11	11	15.84175237
## 12	12	9.41775125
## 13	13	8.83677828
## 14	14	23.96499244
## 15	15	13.07967382
## 16	16	4.01695817
## 17	17	7.99449634
## 18	18	4.58148719
## 19	19	14.70777321
## 20	20	25.78698850
## 21	21	29.99619469
## 22	22	18.16970525
## 23	23	14.09443719
## 24	24	12.61347899
## 25	25	12.29471693
## 26	26	1.71292025
## 27	27	16.78696537
## 28	28	17.82353542
## 29	29	5.68281747
## 30	30	22.49724553
## 31	31	11.25188137
## 32	32	19.08385771
## 33	33	0.05211500
## 34	34	0.29523587
## 35	35	24.78207979
## 36	36	24.27452489
## 37	37	12.58902050
## 38	38	7.31610345
## 39	39	21.98134186

## 40	40 1.44244587
## 41	41 16.48523526
## 42	42 18.47448285
## 43	43 28.68268560
## 44	44 16.51950021
## 45	45 26.10806020
## 46	46 25.44633436
## 47	47 27.31116007
## 48	48 16.78878267
## 49	49 4.59194899
## 50	50 25.07935484
## 51	51 19.51173451
## 52	52 23.75916684
## 53	53 22.00560742
## 54	54 23.85031574
## 55	55 5.17543838
## 56	56 5.22998160
## 57	57 15.10675023
## 58	58 19.09363108
## 59	59 16.93706591
## 60	60 22.65432542
## 61	61 23.89705532
## 62	62 8.72819969
## 63	63 8.44010386
## 64	64 28.33278771
## 65	65 21.83944697
## 66	66 10.93783317
## 67	67 27.76403802
## 68	68 2.56605274
## 69	69 1.22772286
## 70	70 26.77751457
## 71	71 16.04902673
## 72	72 11.37484696
## 73	73 17.39148053
## 74	74 21.88181105
## 75	75 2.09128249
## 76	76 2.31436525
## 77	77 8.61378860
## 78	78 18.21443749
## 79	79 14.50121775
## 80	80 7.34539799
## 81	81 25.34504854
## 82	82 22.62102517
## 83	83 17.53644251
## 84	84 21.02231852
## 85	85 26.09942971
## 86	86 13.48281613
## 87	87 21.73122370
## 88	88 24.38144228
## 89	89 29.51306323
## 90	90 26.85805804
## 91	91 7.09091901
## 92	92 13.55704284
## 93	93 4.64813914

## 94	94 11.72299788
## 95	95 29.16724531
## 96	96 2.56408873
## 97	97 19.96485271
## 98	98 26.84800347
## 99	99 1.98970109
## 100	100 27.80477462
## 101	101 4.68890270
## 102	102 29.75399736
## 103	103 28.86156492
## 104	104 1.62335933
## 105	105 26.58146717
## 106	106 6.95498544
## 107	107 24.37387391
## 108	108 20.63972661
## 109	109 3.31054327
## 110	110 15.24809941
## 111	111 16.66182927
## 112	112 21.08719027
## 113	113 2.13087585
## 114	114 5.50382916
## 115	115 2.18534266
## 116	116 16.72306163
## 117	117 15.27638286
## 118	118 21.93987240
## 119	119 8.14819436
## 120	120 13.22942961
## 121	121 4.81524780
## 122	122 4.63081524
## 123	123 3.90025639
## 124	124 12.27260683
## 125	125 13.50092378
## 126	126 8.49478288
## 127	127 1.31813536
## 128	128 18.35409017
## 129	129 19.33882381
## 130	130 3.67552059
## 131	131 10.96036061
## 132	132 17.36368593
## 133	133 9.11544772
## 134	134 20.69915994
## 135	135 27.80066973
## 136	136 9.96377338
## 137	137 24.10210552
## 138	138 8.26637429
## 139	139 22.07137189
## 140	140 15.21410348
## 141	141 7.70470032
## 142	142 24.57876420
## 143	143 5.11461786
## 144	144 20.73345027
## 145	145 8.30975461
## 146	146 16.99033177
## 147	147 26.57024723

## 148	148 18.91592598
## 149	149 20.11793115
## 150	150 25.66186169
## 151	151 3.30988693
## 152	152 12.44513940
## 153	153 3.90224391
## 154	154 13.82439813
## 155	155 14.13531585
## 156	156 13.10707968
## 157	157 21.15397943
## 158	158 10.74559600
## 159	159 22.55146616
## 160	160 21.56216869
## 161	161 1.36846875
## 162	162 4.44459092
## 163	163 0.47738479
## 164	164 15.44300649
## 165	165 13.61435918
## 166	166 14.59740393
## 167	167 18.13859917
## 168	168 4.90525077
## 169	169 24.78090435
## 170	170 18.75689236
## 171	171 4.70107029
## 172	172 25.96421457
## 173	173 21.05015198
## 174	174 20.60453370
## 175	175 28.19999088
## 176	176 27.77331149
## 177	177 22.88689508
## 178	178 23.77027513
## 179	179 19.82064425
## 180	180 13.67029054
## 181	181 19.86780181
## 182	182 5.09980166
## 183	183 20.09009174
## 184	184 9.19944863
## 185	185 3.47489387
## 186	186 7.87616526
## 187	187 15.94566443
## 188	188 10.17268332
## 189	189 17.05031343
## 190	190 27.05733098
## 191	191 27.02909490
## 192	192 2.83697048
## 193	193 7.85712880
## 194	194 18.33081177
## 195	195 10.23665127
## 196	196 26.60579928
## 197	197 23.83012579
## 198	198 9.70762531
## 199	199 3.94041099
## 200	200 22.12165447
## 201	201 10.68154077

## 202	202 0.76626179
## 203	203 7.09016587
## 204	204 4.05329900
## 205	205 0.83210992
## 206	206 18.85119532
## 207	207 4.35439821
## 208	208 0.64977399
## 209	209 25.84636275
## 210	210 15.84230244
## 211	211 25.86542755
## 212	212 26.07684359
## 213	213 6.18999966
## 214	214 8.43593902
## 215	215 24.18478795
## 216	216 10.62778520
## 217	217 17.19132731
## 218	218 11.31612540
## 219	219 27.80950904
## 220	220 7.08353750
## 221	221 22.54112792
## 222	222 22.89749492
## 223	223 14.78477028
## 224	224 14.04549201
## 225	225 14.93056498
## 226	226 15.61904760
## 227	227 27.90506173
## 228	228 29.03126386
## 229	229 17.95046012
## 230	230 9.50867706
## 231	231 28.96867423
## 232	232 17.23006020
## 233	233 16.48326953
## 234	234 29.62941040
## 235	235 29.53264363
## 236	236 13.06570023
## 237	237 5.51182835
## 238	238 8.97947283
## 239	239 6.95682119
## 240	240 13.94301978
## 241	241 5.88415202
## 242	242 29.99165510
## 243	243 27.53866999
## 244	244 17.52703829
## 245	245 28.10205803
## 246	246 7.26896494
## 247	247 5.85534172
## 248	248 19.06698948
## 249	249 25.44289668
## 250	250 13.32176088
## 251	251 6.08293406
## 252	252 10.95717730
## 253	253 15.72299918
## 254	254 6.05257920
## 255	255 10.13583530

## 256	256 16.32800652
## 257	257 0.50360160
## 258	258 24.69565599
## 259	259 19.79090953
## 260	260 6.41763006
## 261	261 21.39276288
## 262	262 15.25108414
## 263	263 16.11735773
## 264	264 24.61424236
## 265	265 24.08914387
## 266	266 14.65908080
## 267	267 1.53653435
## 268	268 21.94335671
## 269	269 7.29698789
## 270	270 25.79171011
## 271	271 13.81756713
## 272	272 1.52692696
## 273	273 28.68397699
## 274	274 24.44759800
## 275	275 18.47182235
## 276	276 9.82464750
## 277	277 7.37684553
## 278	278 27.34689845
## 279	279 11.81990997
## 280	280 18.56640825
## 281	281 28.41001900
## 282	282 3.69871438
## 283	283 21.62252435
## 284	284 29.01304065
## 285	285 21.17122008
## 286	286 9.23375448
## 287	287 9.28087743
## 288	288 9.11876739
## 289	289 14.28881688
## 290	290 27.64225040
## 291	291 29.12166760
## 292	292 14.81106031
## 293	293 19.26581445
## 294	294 2.18942036
## 295	295 13.09335056
## 296	296 19.36590429
## 297	297 20.83139085
## 298	298 14.47464931
## 299	299 22.82562831
## 300	300 27.00556817
## 301	301 6.12020946
## 302	302 26.50418573
## 303	303 2.33169044
## 304	304 18.21444613
## 305	305 17.73878159
## 306	306 0.88235375
## 307	307 3.94441641
## 308	308 7.16087096
## 309	309 19.83257173

## 310	310 21.29177334
## 311	311 14.23450629
## 312	312 15.29842467
## 313	313 2.49413843
## 314	314 19.13876231
## 315	315 23.24597251
## 316	316 17.85439791
## 317	317 16.83771461
## 318	318 10.20920923
## 319	319 10.30384726
## 320	320 2.80929160
## 321	321 26.55758000
## 322	322 29.83452890
## 323	323 10.94769653
## 324	324 13.64371708
## 325	325 27.14646269
## 326	326 11.73016779
## 327	327 23.42090358
## 328	328 18.68195778
## 329	329 6.77623506
## 330	330 4.32711476
## 331	331 26.58448611
## 332	332 20.53492705
## 333	333 21.78986641
## 334	334 19.78927551
## 335	335 9.16987331
## 336	336 25.49647434
## 337	337 12.95142716
## 338	338 9.21591724
## 339	339 3.71928454
## 340	340 10.18400854
## 341	341 13.81445620
## 342	342 18.75217533
## 343	343 3.04595439
## 344	344 10.79786948
## 345	345 29.31001355
## 346	346 1.21589832
## 347	347 16.71420172
## 348	348 7.01302179
## 349	349 2.50405112
## 350	350 14.31464099
## 351	351 23.06294401
## 352	352 14.79304570
## 353	353 17.88536563
## 354	354 10.09426212
## 355	355 28.05442481
## 356	356 21.08625339
## 357	357 17.73074598
## 358	358 22.54238734
## 359	359 18.68309828
## 360	360 9.12667959
## 361	361 6.78268570
## 362	362 12.87106037
## 363	363 13.27456597

## 364	364 11.96363262
## 365	365 7.82005662
## 366	366 14.76887760
## 367	367 25.22041113
## 368	368 28.75726938
## 369	369 4.62506805
## 370	370 8.21708702
## 371	371 14.27739504
## 372	372 5.19841835
## 373	373 29.14449930
## 374	374 7.17904563
## 375	375 21.33901682
## 376	376 9.15599429
## 377	377 29.91129147
## 378	378 7.24296600
## 379	379 0.56037860
## 380	380 2.84814615
## 381	381 1.77340397
## 382	382 21.86930791
## 383	383 2.43868958
## 384	384 13.71335877
## 385	385 22.65005260
## 386	386 2.35276919
## 387	387 6.99185851
## 388	388 16.07555050
## 389	389 1.66261677
## 390	390 18.19484275
## 391	391 19.43370478
## 392	392 8.45190485
## 393	393 5.64412612
## 394	394 27.98906704
## 395	395 22.63233759
## 396	396 25.22347289
## 397	397 13.63546245
## 398	398 7.64288333
## 399	399 15.19147471
## 400	400 2.91592826
## 401	401 7.02979135
## 402	402 17.54522154
## 403	403 9.56778450
## 404	404 2.31630396
## 405	405 9.02047624
## 406	406 9.19811845
## 407	407 29.74029722
## 408	408 27.20223204
## 409	409 5.07001682
## 410	410 6.42362910
## 411	411 10.36749214
## 412	412 17.06248441
## 413	413 14.09376311
## 414	414 20.99937669
## 415	415 19.15727428
## 416	416 6.19942550
## 417	417 18.63208443

## 418	418 11.16741173
## 419	419 12.83131198
## 420	420 14.85977447
## 421	421 13.66304567
## 422	422 19.83423738
## 423	423 4.01754856
## 424	424 27.47191604
## 425	425 23.92742818
## 426	426 18.87272294
## 427	427 26.66857070
## 428	428 0.15995986
## 429	429 16.22308373
## 430	430 13.42008427
## 431	431 16.17911689
## 432	432 10.01260702
## 433	433 28.56098204
## 434	434 25.75032657
## 435	435 21.39014875
## 436	436 17.99363716
## 437	437 19.80622008
## 438	438 28.30116014
## 439	439 21.51915263
## 440	440 18.16048590
## 441	441 20.55970985
## 442	442 29.42974746
## 443	443 21.53584968
## 444	444 14.64190344
## 445	445 17.29999723
## 446	446 27.82951849
## 447	447 18.69110554
## 448	448 21.12538267
## 449	449 15.30695465
## 450	450 3.55855007
## 451	451 25.98504815
## 452	452 18.26334755
## 453	453 22.07158551
## 454	454 29.04561422
## 455	455 11.31640266
## 456	456 21.28965942
## 457	457 22.34975370
## 458	458 0.52093525
## 459	459 20.34933524
## 460	460 1.42289677
## 461	461 27.59932949
## 462	462 2.89372426
## 463	463 18.81056723
## 464	464 6.40705688
## 465	465 21.20666586
## 466	466 15.50679774
## 467	467 6.70022594
## 468	468 27.89876780
## 469	469 19.11103416
## 470	470 14.01322254
## 471	471 16.22470100

## 472	472 27.87636896
## 473	473 24.34938096
## 474	474 25.39768851
## 475	475 4.45197798
## 476	476 1.53750490
## 477	477 22.63620186
## 478	478 14.25112401
## 479	479 24.80635514
## 480	480 15.37004618
## 481	481 21.88684908
## 482	482 10.51672163
## 483	483 24.50996733
## 484	484 11.72664362
## 485	485 18.45670694
## 486	486 4.31498132
## 487	487 12.07116565
## 488	488 5.01115491
## 489	489 22.04236976
## 490	490 4.63524594
## 491	491 22.42581264
## 492	492 2.26363320
## 493	493 6.90867981
## 494	494 26.44155943
## 495	495 0.55509018
## 496	496 0.14361411
## 497	497 15.38088849
## 498	498 8.41076879
## 499	499 28.90797283
## 500	500 23.83094005
## 501	501 18.23616864
## 502	502 2.37112971
## 503	503 27.59622060
## 504	504 2.43101197
## 505	505 20.11683017
## 506	506 1.15869150
## 507	507 21.94247100
## 508	508 21.83020435
## 509	509 9.59816972
## 510	510 18.86794155
## 511	511 26.48611553
## 512	512 2.29030104
## 513	513 15.46201328
## 514	514 7.95034215
## 515	515 17.05755436
## 516	516 6.58767968
## 517	517 10.07531458
## 518	518 1.78737030
## 519	519 14.13113301
## 520	520 5.33891652
## 521	521 7.99536390
## 522	522 17.49613540
## 523	523 29.61214882
## 524	524 22.00217134
## 525	525 10.62230572

## 526	526 5.39191182
## 527	527 28.28376311
## 528	528 4.09335926
## 529	529 3.32336883
## 530	530 13.09906285
## 531	531 12.29163133
## 532	532 1.02804122
## 533	533 3.63867698
## 534	534 28.89265722
## 535	535 1.05562723
## 536	536 27.72558298
## 537	537 24.31650326
## 538	538 15.21048615
## 539	539 23.74477396
## 540	540 10.95966802
## 541	541 14.82571477
## 542	542 9.69711231
## 543	543 13.49656616
## 544	544 29.05379775
## 545	545 26.99015575
## 546	546 11.70335405
## 547	547 24.72241084
## 548	548 20.80310749
## 549	549 21.29865146
## 550	550 4.27487598
## 551	551 0.18877497
## 552	552 15.52615040
## 553	553 8.03915869
## 554	554 7.31296097
## 555	555 15.91223379
## 556	556 1.51910824
## 557	557 23.18906535
## 558	558 15.88347173
## 559	559 16.51760637
## 560	560 23.22923348
## 561	561 15.25847444
## 562	562 25.09333366
## 563	563 13.93844577
## 564	564 15.25182184
## 565	565 18.29192387
## 566	566 16.18723450
## 567	567 15.78160366
## 568	568 18.97948554
## 569	569 27.56204371
## 570	570 6.90978096
## 571	571 2.81436247
## 572	572 22.30922569
## 573	573 4.22921974
## 574	574 25.49643823
## 575	575 25.86424883
## 576	576 15.46743162
## 577	577 9.50067313
## 578	578 28.23881332
## 579	579 17.41533428

## 580	580 26.07017294
## 581	581 11.27339201
## 582	582 0.44793117
## 583	583 17.17957113
## 584	584 0.57937812
## 585	585 4.70337945
## 586	586 15.36253812
## 587	587 24.09799757
## 588	588 9.98313663
## 589	589 24.29376919
## 590	590 20.03532447
## 591	591 16.86236385
## 592	592 28.49393641
## 593	593 24.80527276
## 594	594 10.29042019
## 595	595 17.28378473
## 596	596 4.29394705
## 597	597 4.13894035
## 598	598 27.10331543
## 599	599 19.01784361
## 600	600 4.13087844
## 601	601 7.61913192
## 602	602 0.22471917
## 603	603 7.58358095
## 604	604 0.70820356
## 605	605 26.15456688
## 606	606 25.00787396
## 607	607 12.13710360
## 608	608 29.51772553
## 609	609 23.29129434
## 610	610 0.79500558
## 611	611 8.64801754
## 612	612 8.92626902
## 613	613 3.55413855
## 614	614 20.95414013
## 615	615 13.97414454
## 616	616 12.05990767
## 617	617 25.30801366
## 618	618 21.94305107
## 619	619 27.24628801
## 620	620 26.99409339
## 621	621 21.48838394
## 622	622 25.64111639
## 623	623 16.02956112
## 624	624 24.97976664
## 625	625 19.43177693
## 626	626 4.07977849
## 627	627 22.07207988
## 628	628 18.20536372
## 629	629 8.25589764
## 630	630 25.94543448
## 631	631 26.68614750
## 632	632 13.17340614
## 633	633 4.13971917

## 634	634 18.69442663
## 635	635 21.15617341
## 636	636 26.62828154
## 637	637 1.58231570
## 638	638 6.39004376
## 639	639 1.36808490
## 640	640 7.59057867
## 641	641 12.24693146
## 642	642 5.26981595
## 643	643 5.19349064
## 644	644 16.23022058
## 645	645 19.43334169
## 646	646 8.32330740
## 647	647 21.81385769
## 648	648 26.36377704
## 649	649 16.52527308
## 650	650 21.36186962
## 651	651 21.76063207
## 652	652 9.43167117
## 653	653 28.81018798
## 654	654 1.97403420
## 655	655 20.11645230
## 656	656 20.45099184
## 657	657 8.20116245
## 658	658 7.61681224
## 659	659 22.39749865
## 660	660 20.45608649
## 661	661 23.24401018
## 662	662 16.03424410
## 663	663 24.69163685
## 664	664 7.58833783
## 665	665 19.79045543
## 666	666 4.83634152
## 667	667 22.71104205
## 668	668 22.46168953
## 669	669 21.28329621
## 670	670 14.84826955
## 671	671 17.71113721
## 672	672 24.99863414
## 673	673 1.11363730
## 674	674 5.18335772
## 675	675 17.93925157
## 676	676 15.98093777
## 677	677 9.71065524
## 678	678 28.08909436
## 679	679 12.12228267
## 680	680 21.41512543
## 681	681 27.59098297
## 682	682 10.43861453
## 683	683 17.17619657
## 684	684 15.64023849
## 685	685 7.81212623
## 686	686 7.94228212
## 687	687 27.67277035

## 688	688 11.51521985
## 689	689 7.23150658
## 690	690 3.21479284
## 691	691 19.54473338
## 692	692 15.11418222
## 693	693 29.11112360
## 694	694 18.85301092
## 695	695 24.07615190
## 696	696 4.43075927
## 697	697 8.07530223
## 698	698 7.10870020
## 699	699 16.42089552
## 700	700 5.14947769
## 701	701 26.45798509
## 702	702 15.92709096
## 703	703 22.54949688
## 704	704 23.11841410
## 705	705 23.20727685
## 706	706 24.02396576
## 707	707 28.11650863
## 708	708 23.53483649
## 709	709 10.83651507
## 710	710 14.52483727
## 711	711 19.27092228
## 712	712 4.96947936
## 713	713 6.37777295
## 714	714 19.66820788
## 715	715 10.33001991
## 716	716 2.76706099
## 717	717 2.65794648
## 718	718 13.18265599
## 719	719 15.87154296
## 720	720 26.15205297
## 721	721 3.47143965
## 722	722 5.68069562
## 723	723 21.81475559
## 724	724 18.99552750
## 725	725 4.07545613
## 726	726 22.48334356
## 727	727 25.15314901
## 728	728 4.14790113
## 729	729 26.80710363
## 730	730 12.71870700
## 731	731 0.23671982
## 732	732 18.27752029
## 733	733 20.38129024
## 734	734 25.49995270
## 735	735 19.34074980
## 736	736 27.88514764
## 737	737 23.62582308
## 738	738 4.57260252
## 739	739 4.84854108
## 740	740 13.21212999
## 741	741 27.19329578

## 742	742 1.45556869
## 743	743 24.08456709
## 744	744 16.55591665
## 745	745 10.49539158
## 746	746 2.12426508
## 747	747 3.30015942
## 748	748 27.21315860
## 749	749 13.60083407
## 750	750 6.42074840
## 751	751 15.80302568
## 752	752 7.64344939
## 753	753 3.61414381
## 754	754 3.44661766
## 755	755 4.44318358
## 756	756 2.85897752
## 757	757 18.31471199
## 758	758 9.24812254
## 759	759 15.63217551
## 760	760 6.64337234
## 761	761 25.32337808
## 762	762 10.41814572
## 763	763 19.13094874
## 764	764 22.86960758
## 765	765 22.85894301
## 766	766 5.86332334
## 767	767 0.05504268
## 768	768 27.66123282
## 769	769 25.70949992
## 770	770 29.95611676
## 771	771 15.84321597
## 772	772 10.35282277
## 773	773 12.69412339
## 774	774 3.80673125
## 775	775 6.16406827
## 776	776 2.99790168
## 777	777 24.26350489
## 778	778 16.80755850
## 779	779 13.83899124
## 780	780 10.87304362
## 781	781 10.24308386
## 782	782 10.72382789
## 783	783 9.30063335
## 784	784 10.87484150
## 785	785 10.25127639
## 786	786 19.50456381
## 787	787 1.03514625
## 788	788 24.25549979
## 789	789 13.27121684
## 790	790 25.87039498
## 791	791 11.24287691
## 792	792 15.81979402
## 793	793 3.38889360
## 794	794 7.21797730
## 795	795 21.59932912

## 796	796 1.33471450
## 797	797 17.68906805
## 798	798 0.33489309
## 799	799 13.29441077
## 800	800 25.58799972
## 801	801 8.34362524
## 802	802 5.24244705
## 803	803 29.86772894
## 804	804 11.27868217
## 805	805 1.77041067
## 806	806 12.26556163
## 807	807 4.08045416
## 808	808 9.23194574
## 809	809 12.71412474
## 810	810 21.38715362
## 811	811 12.41157442
## 812	812 3.27262945
## 813	813 17.68681702
## 814	814 21.29433260
## 815	815 12.17109138
## 816	816 10.58917779
## 817	817 16.98192174
## 818	818 2.54616702
## 819	819 25.22683462
## 820	820 16.03544157
## 821	821 15.04064024
## 822	822 3.77273770
## 823	823 18.32934281
## 824	824 5.93192280
## 825	825 10.68586400
## 826	826 2.71822324
## 827	827 14.42678049
## 828	828 11.41946887
## 829	829 3.00407069
## 830	830 10.19740361
## 831	831 21.18003317
## 832	832 26.14440167
## 833	833 17.06804041
## 834	834 23.59274075
## 835	835 11.64061781
## 836	836 2.32737881
## 837	837 20.44001259
## 838	838 4.76478560
## 839	839 24.71379337
## 840	840 24.77519453
## 841	841 17.99014271
## 842	842 5.75588196
## 843	843 2.16911993
## 844	844 1.22090104
## 845	845 3.82330611
## 846	846 20.15392916
## 847	847 9.60326056
## 848	848 5.16540379
## 849	849 12.45379922

## 850	850 13.28796059
## 851	851 11.86287006
## 852	852 20.29859643
## 853	853 13.84376860
## 854	854 4.95336537
## 855	855 9.14834554
## 856	856 6.23437267
## 857	857 12.74852165
## 858	858 18.42469239
## 859	859 2.50559419
## 860	860 4.44630169
## 861	861 27.09661371
## 862	862 21.36234795
## 863	863 3.00574969
## 864	864 0.83713321
## 865	865 25.36174874
## 866	866 28.87985375
## 867	867 13.74831506
## 868	868 25.35327182
## 869	869 12.70325953
## 870	870 5.17307359
## 871	871 13.63094065
## 872	872 14.55916693
## 873	873 6.27789009
## 874	874 2.80767876
## 875	875 22.85234640
## 876	876 19.30655676
## 877	877 0.27821728
## 878	878 16.68406505
## 879	879 6.40513663
## 880	880 7.72135927
## 881	881 0.79103180
## 882	882 27.04417687
## 883	883 1.79142114
## 884	884 5.56745985
## 885	885 2.18831601
## 886	886 2.21925403
## 887	887 17.70047102
## 888	888 6.21666149
## 889	889 1.36030485
## 890	890 21.21758044
## 891	891 23.19494686
## 892	892 14.47518775
## 893	893 8.84783751
## 894	894 5.32492983
## 895	895 22.48245830
## 896	896 19.28372380
## 897	897 27.48034459
## 898	898 24.70487057
## 899	899 8.00545029
## 900	900 8.45565304
## 901	901 7.80314541
## 902	902 25.80687525
## 903	903 28.20051021

## 904	904 13.05773278
## 905	905 19.17772947
## 906	906 29.72308018
## 907	907 20.27046272
## 908	908 17.97622324
## 909	909 21.67749231
## 910	910 1.12684628
## 911	911 22.94989524
## 912	912 24.13175539
## 913	913 28.41362476
## 914	914 25.11089469
## 915	915 0.67646472
## 916	916 16.55111630
## 917	917 15.80507174
## 918	918 21.53329839
## 919	919 27.35162175
## 920	920 4.24038327
## 921	921 15.61486362
## 922	922 11.89807023
## 923	923 21.03012824
## 924	924 8.61335917
## 925	925 2.95336662
## 926	926 22.10096980
## 927	927 21.42984963
## 928	928 26.57457779
## 929	929 0.32530567
## 930	930 11.67946577
## 931	931 24.84896766
## 932	932 28.30590534
## 933	933 19.27874641
## 934	934 22.59183099
## 935	935 5.54906694
## 936	936 15.92738783
## 937	937 23.82532625
## 938	938 27.78281898
## 939	939 11.12884755
## 940	940 17.37371000
## 941	941 2.08964465
## 942	942 14.29634497
## 943	943 17.93836294
## 944	944 0.02836663
## 945	945 23.52984892
## 946	946 12.08360466
## 947	947 4.83382977
## 948	948 22.58295289
## 949	949 24.41841301
## 950	950 4.35457614
## 951	951 18.55395347
## 952	952 9.81627590
## 953	953 12.85618000
## 954	954 3.75257013
## 955	955 14.58650635
## 956	956 28.88595521
## 957	957 12.81251761


```
## 958          958 17.77156874
## 959          959  2.36495243
## 960          960 10.50472292
## 961          961 10.88613254
## 962          962 18.03848800
## 963          963  6.85033504
## 964          964 29.35935741
## 965          965 14.77169513
## 966          966 15.44966381
## 967          967 11.82772621
## 968          968 28.18201558
## 969          969 28.28515080
## 970          970 25.84118795
## 971          971  2.32863119
## 972          972  6.74184497
## 973          973 15.22274302
## 974          974 13.61504895
## 975          975 26.69289316
## 976          976  2.79583029
## 977          977 20.07001006
## 978          978 28.03155079
## 979          979  5.06785383
## 980          980  2.62971973
## 981          981 28.59565169
## 982          982 11.50206714
## 983          983 11.71660302
## 984          984 15.06892989
## 985          985 15.64412930
## 986          986 23.11437869
## 987          987 12.91659115
## 988          988 21.31045686
## 989          989  8.83304000
## 990          990 28.03850106
## 991          991 24.60741165
## 992          992  1.44645144
## 993          993 26.57894109
## 994          994 29.12689109
## 995          995 23.14159813
## 996          996 24.09883305
## 997          997 26.35455905
## 998          998  9.82991366
## 999          999 20.18595314
## 1000         1000 22.39149915
```

```
head(wait_times, 10)
```

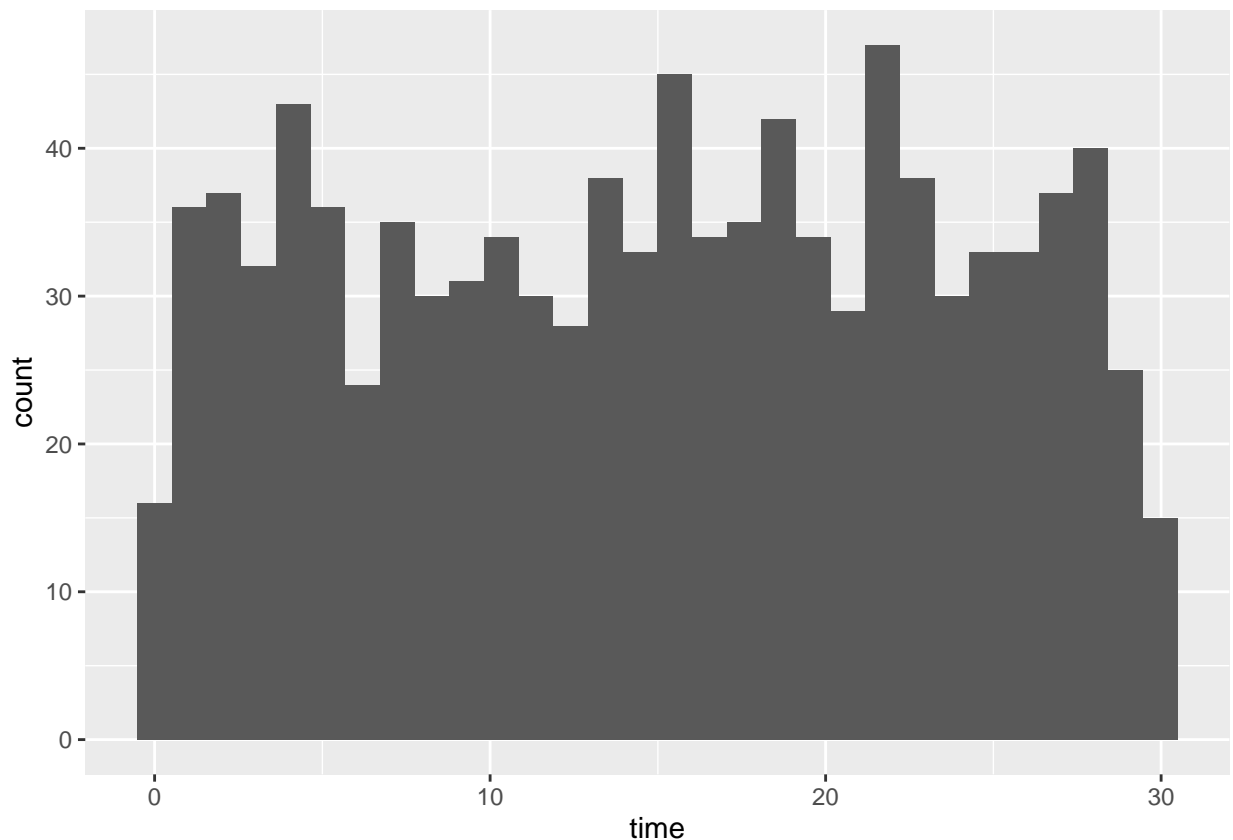
```
##      simulation_nb
## 1                1
## 2                2
## 3                3
## 4                4
## 5                5
## 6                6
## 7                7
## 8                8
```

```
## 9          9
## 10         10
```

- Create a histogram of the simulated wait times with 30 bins.

```
# Set random seed to 334
set.seed(334)

# Generate 1000 wait times between 0 and 30 mins, save in time column
wait_times %>%
  mutate(time = runif(1000, min = 0, max = 30)) %>%
  # Create a histogram of simulated times
  ggplot(aes(time)) +
  geom_histogram(bins = 30)
```



Simulating sales deals

Assume that Amir usually works on 3 deals per week, and overall, he wins 30% of deals he works on. Each deal has a binary outcome: it's either lost, or won, so you can model his sales deals with a binomial distribution. In this exercise, you'll help Amir simulate a year's worth of his deals so he can better understand his performance.

Instructions

- Set the random seed to 10 and simulate a single deal.

```
# Set random seed to 10
set.seed(10)
```

```
# Simulate a single deal
rbinom(1, 1, 0.3)
```

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

- Simulate a typical week of Amir's deals, or one week of 3 deals.

```
# Set random seed to 10
set.seed(10)
```

```
# Simulate 1 week of 3 deals
rbinom(1, 3, 0.3)
```

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

- Simulate a year's worth of Amir's deals, or 52 weeks of 3 deals each, and store in deals. Calculate the mean number of deals he won per week.

```
# Set random seed to 10
set.seed(10)
```

```
# Simulate 52 weeks of 3 deals
deals <- rbinom(52, 3, 0.3)
```

```
# Calculate mean deals won per week
mean(deals)
```

```
## [1] 0.8076923
```

```
deals
```

```
## [1] 1 0 1 1 0 0 0 0 1 1 1 1 0 1 1 1 0 0 1 2 2 1 1 1 1 2 0 1 1 1 0 0 2 1 1 2 2
## [39] 1 1 0 0 0 1 0 0 0 1 0 2 1 2
```

Calculating binomial probabilities

Just as in the last exercise, assume that Amir wins 30% of deals. He wants to get an idea of how likely he is to close a certain number of deals each week. In this exercise, you'll calculate what the chances are of him closing different numbers of deals using the binomial distribution.

Instructions

- What's the probability that Amir closes all 3 deals in a week?

```
# Probability of closing 3 out of 3 deals
dbinom(3, 3, 0.3)
```

```
## [1] 0.027
```

- What's the probability that Amir closes 1 or fewer deals in a week?

```
# Probability of closing <= 1 deal out of 3 deals
pbinom(1, 3, 0.3)
```

```
## [1] 0.784
```

- What's the probability that Amir closes more than 1 deal?

```
# Probability of closing > 1 deal out of 3 deals
pbinom(1, 3, 0.3, lower.tail = FALSE)
```

```
## [1] 0.216
```

How many sales will be won?

Now Amir wants to know how many deals he can expect to close each week if his win rate changes. Luckily, you can use your binomial distribution knowledge to help him calculate the expected value in different situations. Recall from the video that the expected value of a binomial distribution can be calculated by $n \times p$

- Calculate the expected number of sales out of the 3 he works on that Amir will win each week if he maintains his 30% win rate.
- Calculate the expected number of sales out of the 3 he works on that he'll win if his win rate drops to 25%.
- Calculate the expected number of sales out of the 3 he works on that he'll win if his win rate rises to 35%.

```
# Expected number won with 30% win rate  
won_30pct <- 3 * 0.3  
won_30pct
```

```
## [1] 0.9
```

```
# Expected number won with 25% win rate  
won_25pct <- 3 * 0.25  
won_25pct
```

```
## [1] 0.75
```

```
# Expected number won with 35% win rate  
won_35pct <- 3 * 0.35  
won_35pct
```

```
## [1] 1.05
```