

Roulette Simulation

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Introduction

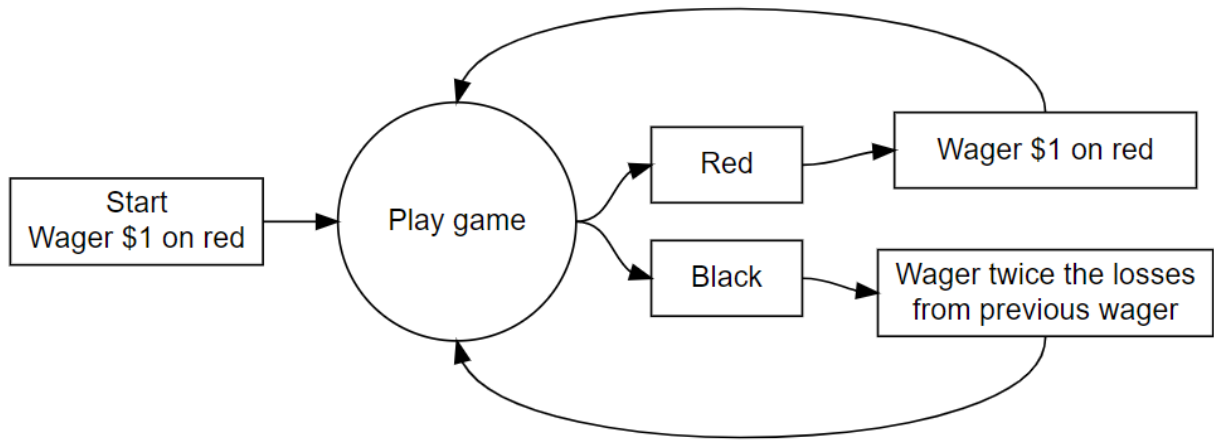
In this blog post, we explore Martingale's strategy and its viability through simulation and based on certain constraints applicable in a real world scenario(albeit with different parameter values depending on the setting).

Background

A brief blurb about Roulette in General :

A roulette table composed of 38 evenly sized pockets on a wheel. The pockets are colored red, black, or green. The pockets are also numbered. Roulette is a game of chance in which a pocket is randomly selected. Gamblers may wager on several aspects of the outcome. For example, one may place a wager that the randomly selected pocket will be red or odd numbered or will be a specific number. In our scenario, all one needs to know is that there are 38 pockets of which 2 are green, 18 are red, and 18 are black. The payout for a bet on black (or red) is \$1 for each \$1 wagered. This means that if a gambler bets \$1 on black and the randomly selected pocket is black, then the gambler will get the original \$1 wager and an additional \$1 as winnings.

Here's a pictorial representation of Martingale's strategy. The Martingale strategy appears to always end in positive earnings, regardless of how unlucky a string of spins may be.



Constraints for the Simulation :

Stopping rule

A player will use the above strategy and play until

- The player has W(Winning Threshold for Stopping) dollars
- The player goes bankrupt
- The player completes L wagers (or plays)

In this scenario W is 300\$ (Starting Budget + Winnings) and L is 1000 plays.

Budget

The player starts with B(Starting Budget B - 200\$) dollars. The player cannot wager more money than he/she has.

Maximum wager

Some casinos have a maximum bet. Call this parameter M. If the strategy directs the player to wager more than M dollars, then the player will only wager M dollars(100\$ in this scenario).

Methods

The below chunks of code show the modular processes involved in simulation of the roulette game based on the above constraints. `one_series()` function is a culmination of the entire roulette spin which utilizes all the modular functions.

This chunk of code simulates a single spin of the roulette

```

#Simulates Single Spin
single_spin <- function(){
  possible_outcomes <- c(rep("red",18), rep("black",18), rep("green",2))
  sample(possible_outcomes, 1)
}
  
```

This chunk of code calculates the wager for the upcoming turn based on the previous wager as well as the previous outcome.

```
#Simulates martingale wager
martingale_wager <- function(
  previous_wager
  , previous_outcome
  , max_wager
  , current_budget
){
  if(previous_outcome == "red") return(1)
  min(2*previous_wager, max_wager, current_budget)
}
```

This chunk of code simulates one turn of the roulette outputting specific parameters to be used as part of a series of spins

```
#Simulates a single play/turn in a roulette game.
one_play <- function(previous_ledger_entry, max_wager){
  # Create a copy of the input object that will become the output object
  out <- previous_ledger_entry
  out[1, "game_index"] <- previous_ledger_entry[1, "game_index"] + 1
  out[1, "starting_budget"] <- previous_ledger_entry[1, "ending_budget"]
  out[1, "wager"] <- martingale_wager(
    previous_wager = previous_ledger_entry[1, "wager"]
    , previous_outcome = previous_ledger_entry[1, "outcome"]
    , max_wager = max_wager
    , current_budget = out[1, "starting_budget"]
  )
  out[1, "outcome"] <- single_spin()
  out[1, "ending_budget"] <- out[1, "starting_budget"] +
    ifelse(out[1, "outcome"] == "red", +1, -1)*out[1, "wager"]
  return(out)
}
```

This chunk of code establishes the stopping condition based on factors discussed above

```
#Stopping Condition
stopping_rule <- function(
  ledger_entry
  , winning_threshold
){
  ending_budget <- ledger_entry[1, "ending_budget"]
  if(ending_budget <= 0) return(TRUE)
  if(ending_budget >= winning_threshold) return(TRUE)
  FALSE
}
```

This chunk of code simulates the entire series of roulette spins until a certain stopping condition is achieved

```
#Simulation of the entire roulette game.
one_series <- function(
  max_games, starting_budget, winning_threshold, max_wager
){
  # Initialize ledger
  ledger <- data.frame(
```

```

    game_index = 0:max_games
    , starting_budget = NA_integer_
    , wager = NA_integer_
    , outcome = NA_character_
    , ending_budget = NA_integer_
  )
  ledger[1, "wager"] <- 1
  ledger[1, "outcome"] <- "red"
  ledger[1, "ending_budget"] <- starting_budget
  for(i in 2:nrow(ledger)){
    ledger[i,] <- one_play(ledger[i-1,], max_wager)
    if(stopping_rule(ledger[i,], winning_threshold)) break
  }
  # Return non-empty portion of ledger
  ledger[2:i, ]
}

```

Results

The below code chunk gives the profit over a full series of roulette spins which is a property of interest

```

#Profit function given the entire series of plays..
profit <- function(ledger){
  n <- nrow(ledger)
  profit <- ledger[n, "ending_budget"] - ledger[1, "starting_budget"]
  return(profit)
}

```

```

#All of the below function inputs are based on constraints described above
a <- one_series(1000,200,300,100)
a

```

##	game_index	starting_budget	wager	outcome	ending_budget
## 2	1	200	1	red	201
## 3	2	201	1	black	200
## 4	3	200	2	black	198
## 5	4	198	4	black	194
## 6	5	194	8	red	202
## 7	6	202	1	red	203
## 8	7	203	1	red	204
## 9	8	204	1	red	205
## 10	9	205	1	red	206
## 11	10	206	1	red	207
## 12	11	207	1	black	206
## 13	12	206	2	red	208
## 14	13	208	1	red	209
## 15	14	209	1	red	210
## 16	15	210	1	black	209
## 17	16	209	2	black	207
## 18	17	207	4	red	211
## 19	18	211	1	black	210

## 20	19	210	2	red	212
## 21	20	212	1	red	213
## 22	21	213	1	black	212
## 23	22	212	2	black	210
## 24	23	210	4	red	214
## 25	24	214	1	black	213
## 26	25	213	2	red	215
## 27	26	215	1	green	214
## 28	27	214	2	red	216
## 29	28	216	1	red	217
## 30	29	217	1	red	218
## 31	30	218	1	black	217
## 32	31	217	2	red	219
## 33	32	219	1	red	220
## 34	33	220	1	green	219
## 35	34	219	2	black	217
## 36	35	217	4	black	213
## 37	36	213	8	red	221
## 38	37	221	1	black	220
## 39	38	220	2	red	222
## 40	39	222	1	red	223
## 41	40	223	1	black	222
## 42	41	222	2	red	224
## 43	42	224	1	red	225
## 44	43	225	1	black	224
## 45	44	224	2	red	226
## 46	45	226	1	red	227
## 47	46	227	1	red	228
## 48	47	228	1	red	229
## 49	48	229	1	red	230
## 50	49	230	1	red	231
## 51	50	231	1	red	232
## 52	51	232	1	red	233
## 53	52	233	1	red	234
## 54	53	234	1	red	235
## 55	54	235	1	black	234
## 56	55	234	2	red	236
## 57	56	236	1	red	237
## 58	57	237	1	black	236
## 59	58	236	2	red	238
## 60	59	238	1	red	239
## 61	60	239	1	black	238
## 62	61	238	2	red	240
## 63	62	240	1	red	241
## 64	63	241	1	black	240
## 65	64	240	2	red	242
## 66	65	242	1	black	241
## 67	66	241	2	red	243
## 68	67	243	1	black	242
## 69	68	242	2	red	244
## 70	69	244	1	black	243
## 71	70	243	2	black	241
## 72	71	241	4	red	245
## 73	72	245	1	red	246

## 74	73	246	1	black	245
## 75	74	245	2	red	247
## 76	75	247	1	black	246
## 77	76	246	2	black	244
## 78	77	244	4	black	240
## 79	78	240	8	black	232
## 80	79	232	16	red	248
## 81	80	248	1	black	247
## 82	81	247	2	black	245
## 83	82	245	4	black	241
## 84	83	241	8	red	249
## 85	84	249	1	black	248
## 86	85	248	2	red	250
## 87	86	250	1	black	249
## 88	87	249	2	red	251
## 89	88	251	1	green	250
## 90	89	250	2	black	248
## 91	90	248	4	black	244
## 92	91	244	8	red	252
## 93	92	252	1	black	251
## 94	93	251	2	red	253
## 95	94	253	1	red	254
## 96	95	254	1	red	255
## 97	96	255	1	black	254
## 98	97	254	2	black	252
## 99	98	252	4	black	248
## 100	99	248	8	red	256
## 101	100	256	1	red	257
## 102	101	257	1	black	256
## 103	102	256	2	red	258
## 104	103	258	1	red	259
## 105	104	259	1	black	258
## 106	105	258	2	black	256
## 107	106	256	4	red	260
## 108	107	260	1	black	259
## 109	108	259	2	black	257
## 110	109	257	4	black	253
## 111	110	253	8	black	245
## 112	111	245	16	black	229
## 113	112	229	32	red	261
## 114	113	261	1	black	260
## 115	114	260	2	red	262
## 116	115	262	1	black	261
## 117	116	261	2	red	263
## 118	117	263	1	black	262
## 119	118	262	2	black	260
## 120	119	260	4	black	256
## 121	120	256	8	red	264
## 122	121	264	1	red	265
## 123	122	265	1	red	266
## 124	123	266	1	red	267
## 125	124	267	1	red	268
## 126	125	268	1	black	267
## 127	126	267	2	black	265

## 128	127	265	4	red	269
## 129	128	269	1	black	268
## 130	129	268	2	black	266
## 131	130	266	4	red	270
## 132	131	270	1	red	271
## 133	132	271	1	red	272
## 134	133	272	1	black	271
## 135	134	271	2	black	269
## 136	135	269	4	red	273
## 137	136	273	1	black	272
## 138	137	272	2	red	274
## 139	138	274	1	red	275
## 140	139	275	1	black	274
## 141	140	274	2	red	276
## 142	141	276	1	black	275
## 143	142	275	2	black	273
## 144	143	273	4	red	277
## 145	144	277	1	red	278
## 146	145	278	1	red	279
## 147	146	279	1	black	278
## 148	147	278	2	red	280
## 149	148	280	1	red	281
## 150	149	281	1	red	282
## 151	150	282	1	red	283
## 152	151	283	1	black	282
## 153	152	282	2	red	284
## 154	153	284	1	black	283
## 155	154	283	2	red	285
## 156	155	285	1	black	284
## 157	156	284	2	red	286
## 158	157	286	1	black	285
## 159	158	285	2	black	283
## 160	159	283	4	black	279
## 161	160	279	8	red	287
## 162	161	287	1	green	286
## 163	162	286	2	black	284
## 164	163	284	4	red	288
## 165	164	288	1	green	287
## 166	165	287	2	black	285
## 167	166	285	4	red	289
## 168	167	289	1	black	288
## 169	168	288	2	black	286
## 170	169	286	4	red	290
## 171	170	290	1	red	291
## 172	171	291	1	green	290
## 173	172	290	2	black	288
## 174	173	288	4	red	292
## 175	174	292	1	black	291
## 176	175	291	2	red	293
## 177	176	293	1	black	292
## 178	177	292	2	red	294
## 179	178	294	1	black	293
## 180	179	293	2	red	295
## 181	180	295	1	black	294

```
## 182      181      294      2    black      292
## 183      182      292      4     red       296
## 184      183      296      1     red       297
## 185      184      297      1     red       298
## 186      185      298      1     red       299
## 187      186      299      1     red       300
```

```
profit(a)
```

```
## [1] 100
```

The results are not always positive in the sense that the chances of winning and losing seem to be random and the Martingale strategy **under these constraints is not** a guaranteed success. A specific observation is that a vast majority of profit results tend to be either -200 or 100. This is due to the either one of the stopping conditions being achieved - 1000 turns (extremely low chance of happening due to the small difference between starting amount and the winning threshold and the starting amount itself being low enough that 8-9 continuous losses would cause the player to lose the entire initial capital) or achieving the winning threshold or losing the initial capital/budget.

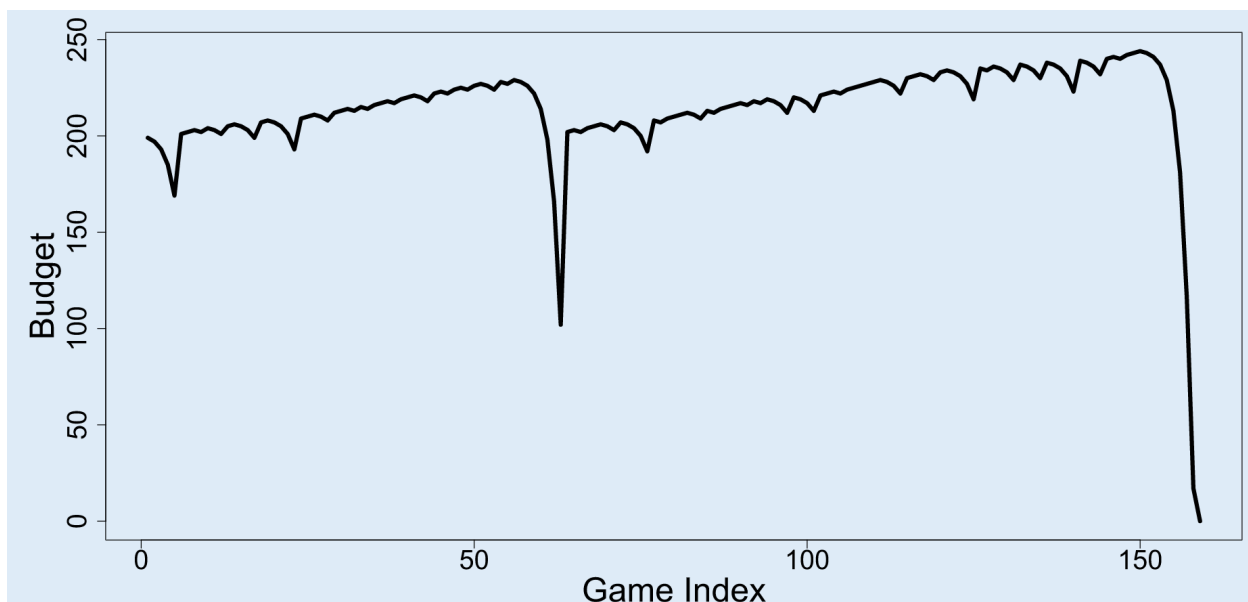
```
library(magrittr)
```

```
svg(filename = "pattern.svg", width=16, height =8.25)
par(cex.axis=2.0, cex.lab = 2.5, mar = c(8,8,2,2), bg = rgb(222, 235, 247, max = 255))
#set.seed(1)
#Constraints chosen based on above stipulations.
ledger <- one_series(1000,200,300,100)
plot(ledger[,c(1,5)], type = "l", lwd = 5, xlab = "Game Index", ylab = "Budget")
dev.off()
```

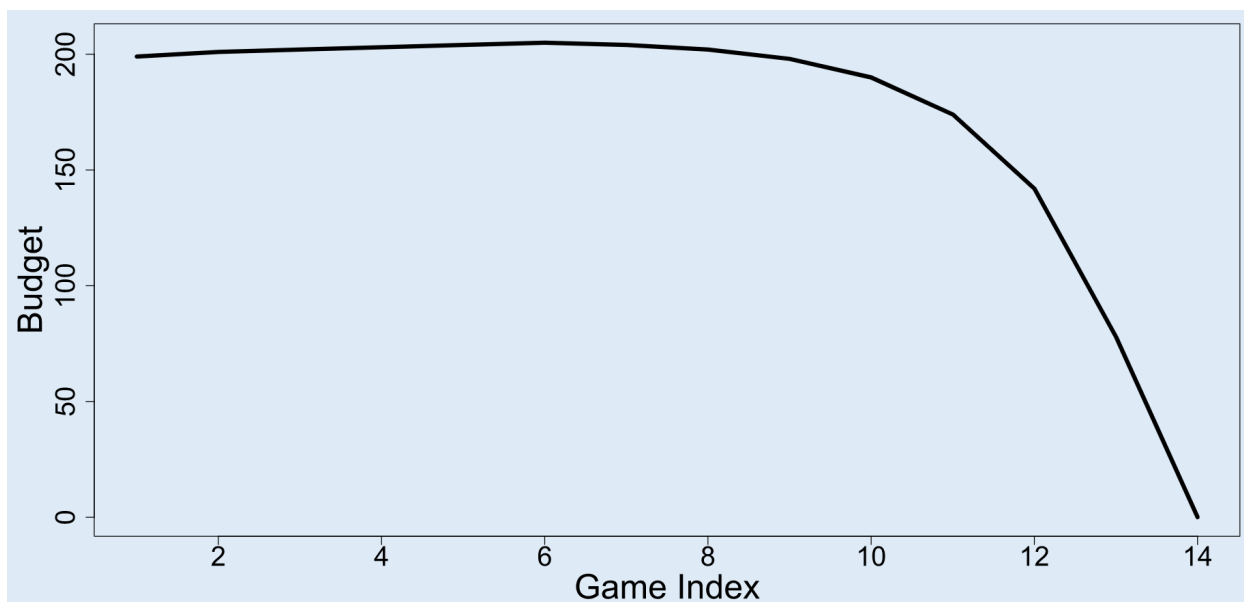
```
## pdf
## 2
```

The below figures show both gamblers winning and losing over the series of wagers based on above constraints.

Lost Wager



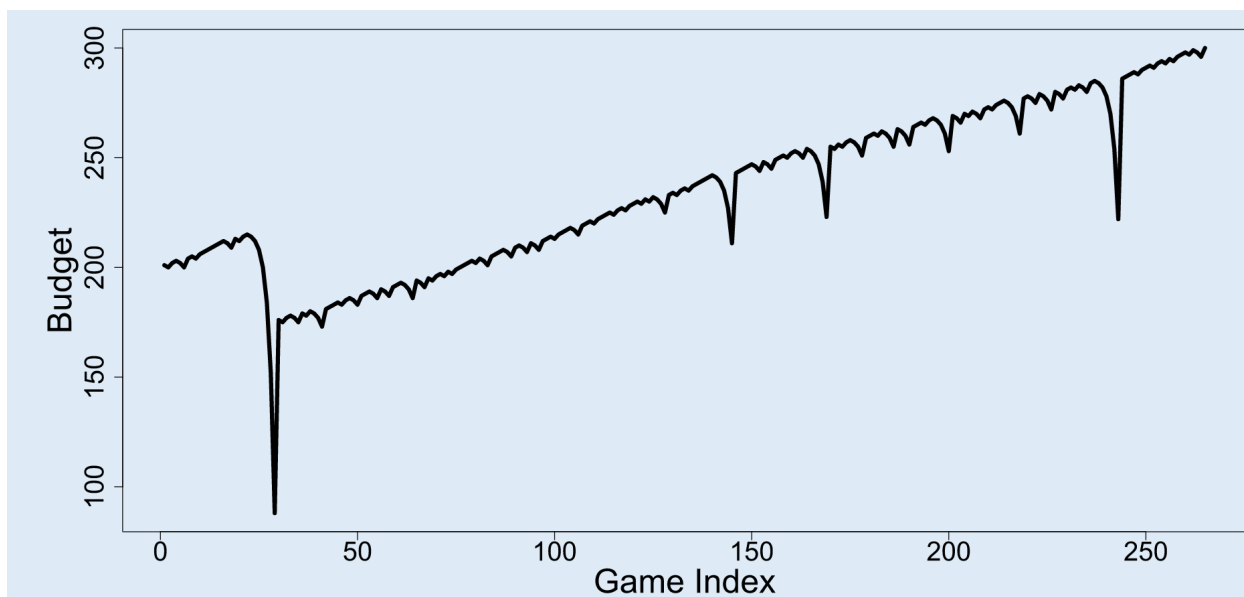
Here we see an example of a player get half way to his win condition but drops off sharply at his 250th turn.



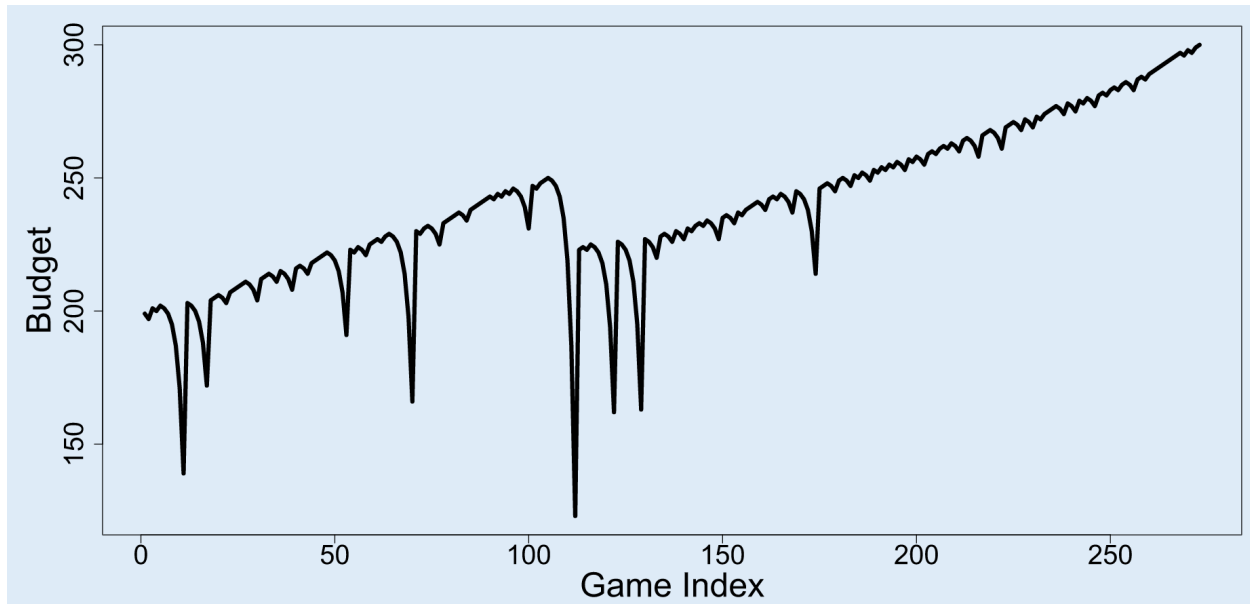
Here we see an example of a very swift loss right at turn 14.

Here we see a

Won Wager



Here is an example of a player almost losing at around turn 30 but managing to use the strategy and reach his win condition at turn 270 or so.



Here is another example of a player almost losing at many points in the game but managing to eek out the win eventually.

Conclusion and Discussion

A massive limitation with this simulation is the chosen constraint value. It can be concluded that the Martingale's strategy cannot guarantee a win with the above constraints. But, modification of said constraints can increase the probability of positive results. It can also be effectively concluded that the Martingale's strategy **guarantees** positive net earnings when there are no constraints and that the gambler has full control as to how many turns he gets to play, has an infinite (or a very large capital to spend) and there isn't a maximum bet restriction. Modifying some of these constraints can shift the odds in the gambler's favor though.

For instance,

- The budget of the gambler is directly proportional to the chance of a positive outcome.
- The number of chances to play is also directly proportional to the chance of a positive outcome. (In our scenario this played a very minor role due to how the stopping conditions were set up.)
- The maximum bet amount is inversely (although not in any linear fashion) proportional to the chance of a positive outcome.

All of the above listed points are highly situational and only applicable in a specific scenario to successfully execute the Martingale strategy as will be discussed in the conclusion.

The function below returns the total profits/losses of many replicates of roulette games using different constraints.

```
PResult <- function(r, max_games, starting_budget, winning_threshold, max_wager)
{
  rep <- c(replicate(r, profit(one_series(max_games, starting_budget, winning_threshold, max_wager))))
  return (rep)
}
```

```
library(tidyverse)
```

```
## -- Attaching packages ----- tidyverse 1.3.1 --
```

```
## v ggplot2 3.3.5      v purrr  0.3.4
## v tibble  3.1.4      v dplyr  1.0.7
## v tidyr   1.1.3      v stringr 1.4.0
## v readr   2.0.1      v forcats 0.5.1
```

```
## -- Conflicts ----- tidyverse_conflicts() --
```

```
## x tidyr::extract() masks magrittr::extract()
## x dplyr::filter()  masks stats::filter()
## x dplyr::lag()      masks stats::lag()
## x purrr::set_names() masks magrittr::set_names()
```

```
library(ggplot2)
```

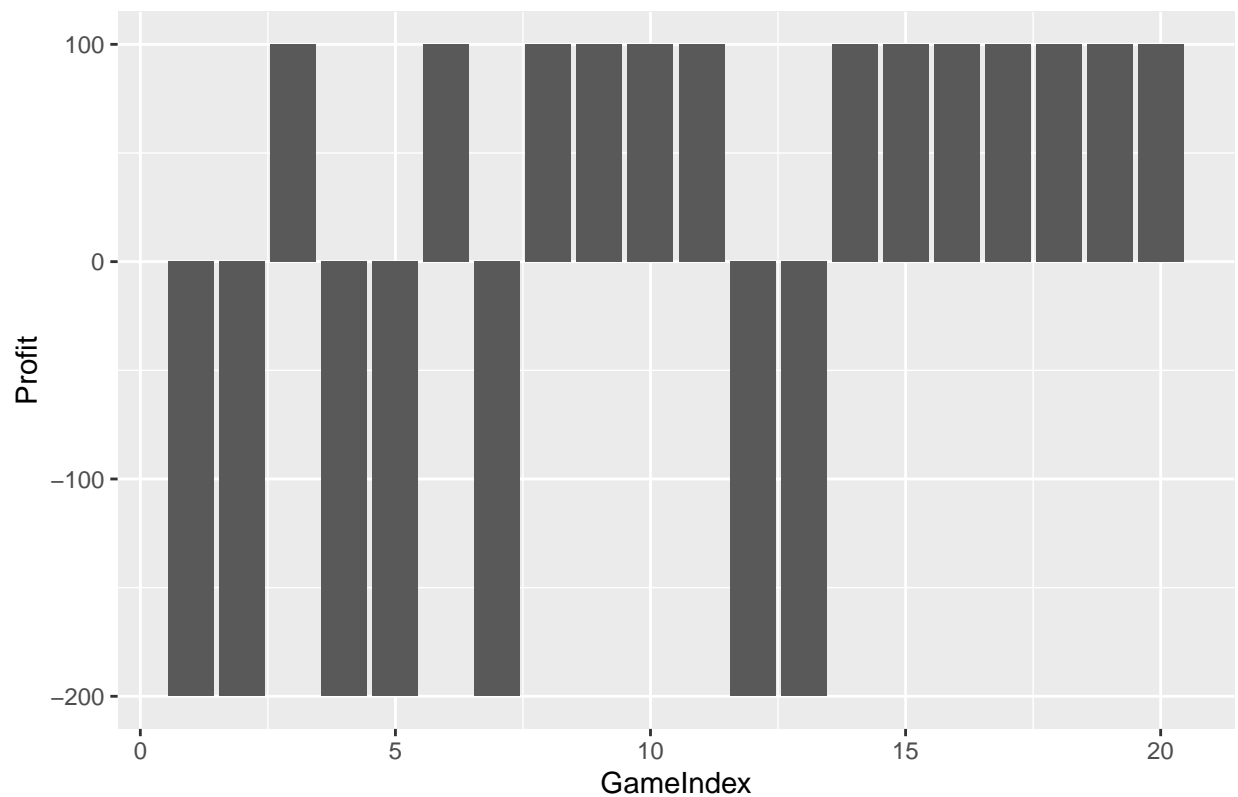
This chunk of code shows how changing parameters based on the discussion above affects the outcome of the series of wagers i.e, gives us the distribution of earnings over games as well as average earnings over a series of games.

In order to make the histogram viewable, it's shown for 20 repetitions but the distribution can be extended upto any number of required repetitions. The means/average earnings shown will be for 1000 repetitions.

```
##.
set.seed(1)
a <- PResult(20, 1000, 200, 300, 100)
a1 <- as.data.frame(a)
a1 <- mutate(a1, GameIndex = 1:length(a))
colnames(a1) <- c("Profit", "GameIndex")
ggplot(data = a1, aes(x = GameIndex, y = Profit)) + geom_histogram(stat = "identity") + labs(title = "A
```

```
## Warning: Ignoring unknown parameters: binwidth, bins, pad
```

Average Earnings Simulation (20 Repetitions)



```
a2 <- PResult(1000, 1000, 200, 300, 100)
a2 <- as.data.frame(a2)
a2 <- mutate(a2, GameIndex = 1:length(a2))
colnames(a2) <- c("Profit", "GameIndex")
mean(a2$Profit)
```

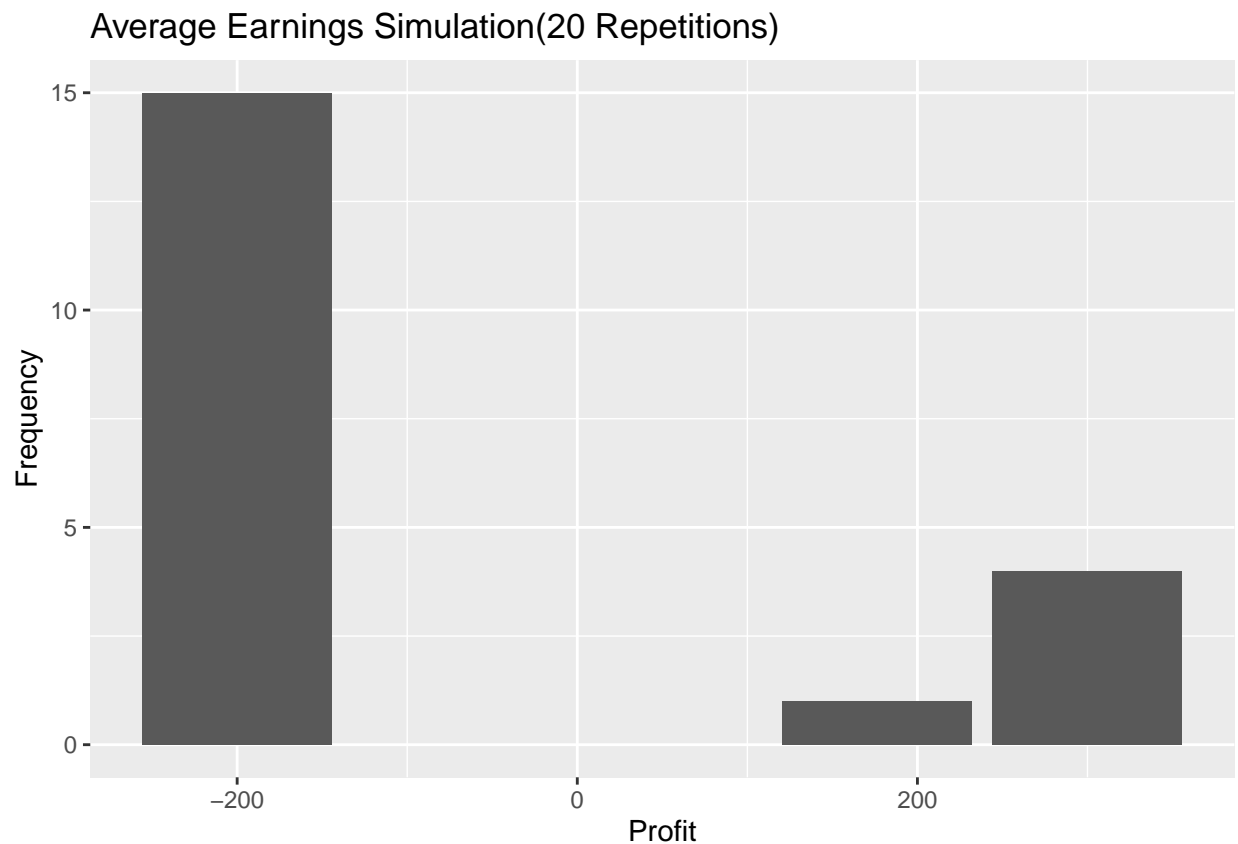
```
## [1] -48.316
```

```
# get_last <- function(x) x[length(x)]
#
#
# # Simulation
# walk_out_money <- rep(NA, 50)
# for(j in seq_along(walk_out_money)){
#   walk_out_money[j] <- one_series(starting_budget = 200, winning_threshold = 300, max_games = 1000, m
# }
#
# lapply(walk_out_money, function(x) x[length(x)] - 200)
#
# # Walk out money distribution
# #hist(as.numeric(walk_out_money), breaks = 100)
#
# d1 <- as.data.frame(lapply(walk_out_money, function(x) x[length(x)] - 200))
# index(d1, col = 1:length(d1))
# dim(d1)
```

With these set of parameters, we have a mean earnings of -48.316.

```
set.seed(1)
b<- PResult(20, 1000, 200, 500, 100)
b1 <- as.data.frame(b)
b1 <- mutate(b1, GameIndex = 1:length(b))
colnames(b1) <- c("Profit", "GameIndex")
b2 <- b1 %>% group_by(Profit) %>% count()
colnames(b2) <- c("Profit", "Frequency")
ggplot(data = b2, aes(x = Profit, y = Frequency)) + geom_col(stat = "identity") + labs(title = "Average
```

```
## Warning: Ignoring unknown parameters: stat
```



```
b2<- PResult(1000, 1000, 200, 500, 100)
b2 <- as.data.frame(b2)
b2 <- mutate(b2, GameIndex = 1:length(b2))
colnames(b2) <- c("Profit", "GameIndex")
mean(b2$Profit)
```

```
## [1] -96.324
```

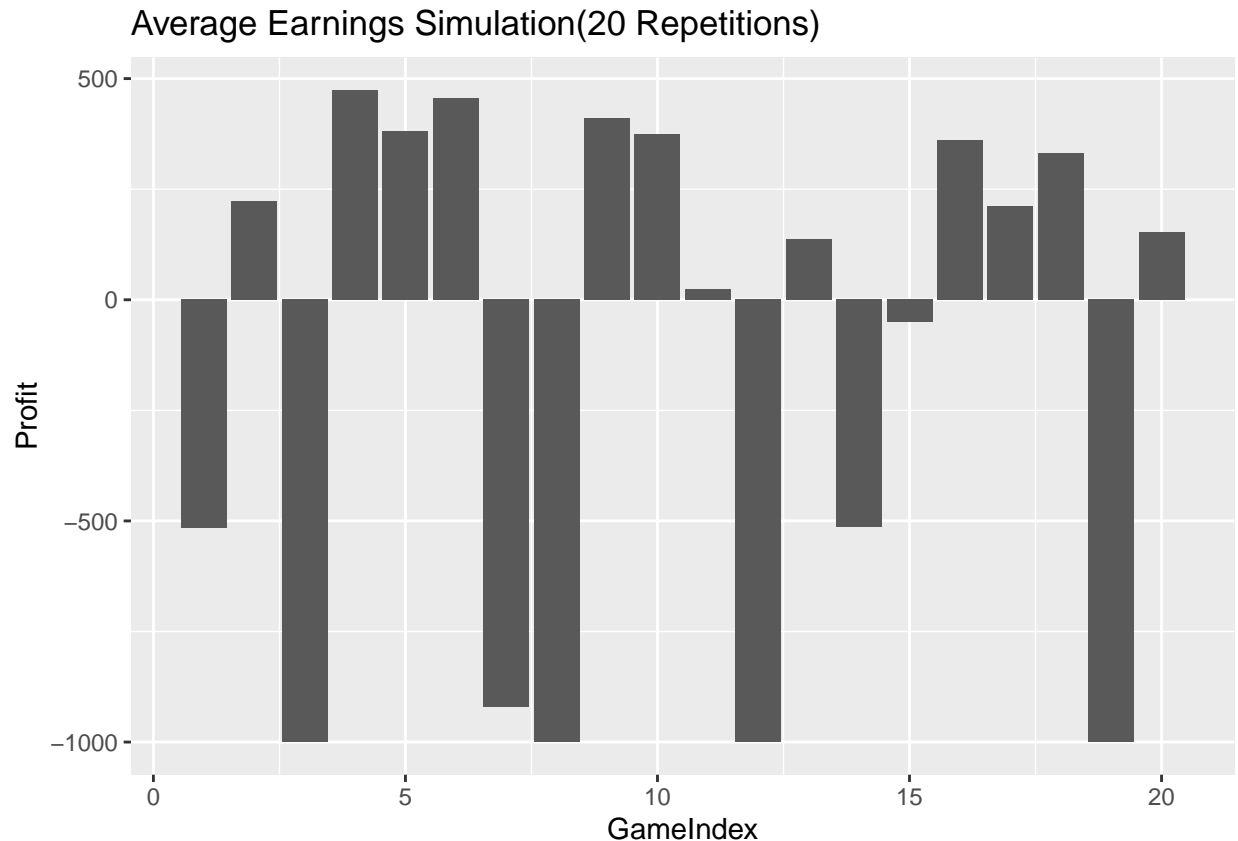
With these set of parameters, the mean earnings is -96.324. We had increased the winning threshold which proves to be problematic for the most part as the player would have to risk more to reach the higher winning threshold(to attain stopping condition) while having only 200\$ to play around with.

```

set.seed(1)
c <- PResult(20, 1000, 1000, 2000, 200)
c1 <- as.data.frame(c)
c1 <- mutate(c1, GameIndex = 1:length(c))
colnames(c1) <- c("Profit", "GameIndex")
ggplot(data = c1, aes(x = GameIndex, y = Profit)) + geom_histogram(stat = "identity") + labs(title = "A

```

```
## Warning: Ignoring unknown parameters: binwidth, bins, pad
```



```

c2<- PResult(1000, 1000, 200, 500, 100)
c2 <- as.data.frame(c2)
c2 <- mutate(c2, GameIndex = 1:length(c2))
colnames(c2) <- c("Profit", "GameIndex")
mean(c2$Profit)

```

```
## [1] -97.798
```

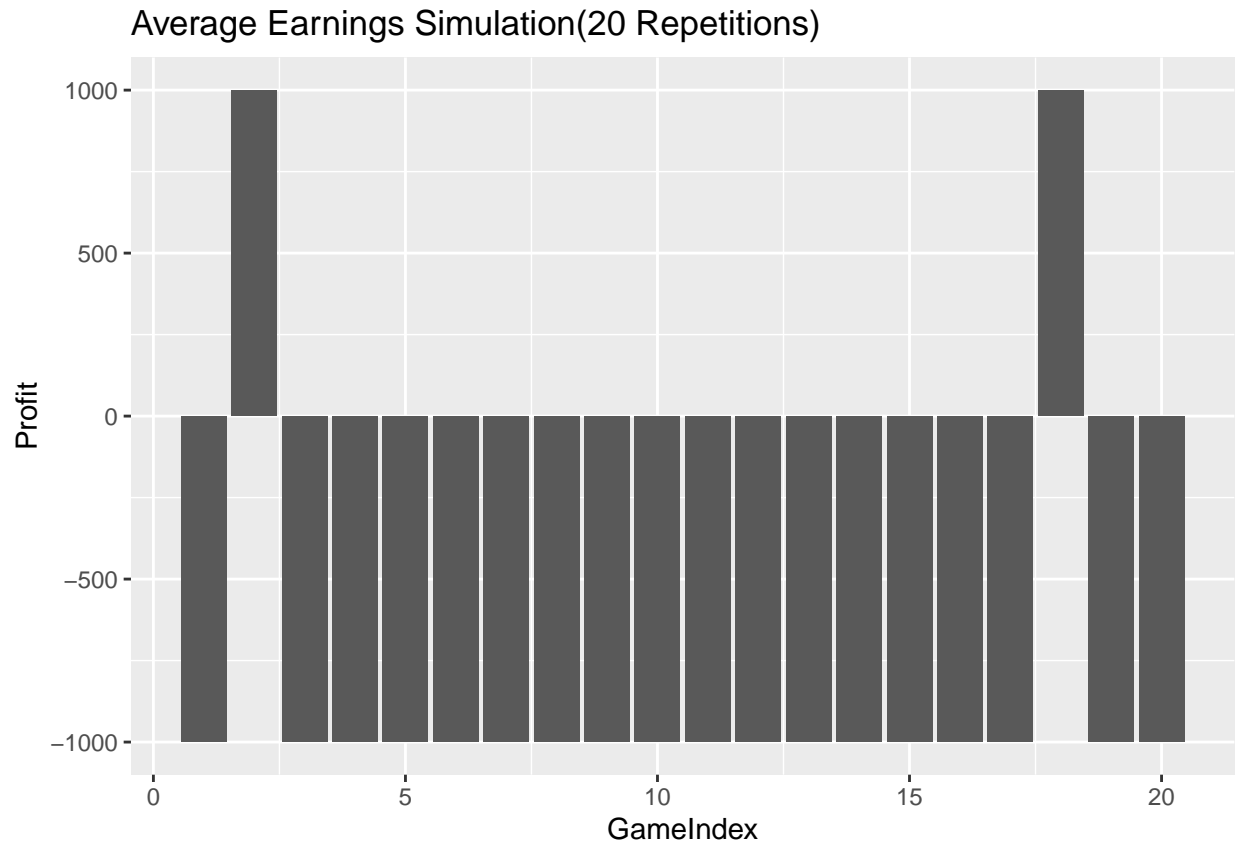
With these set of parameters, the mean earnings is -97.798 (We seem to have some sorely unlucky series of games so far!). We had increased the starting threshold and adjusted the winning threshold as well and the results don't seem too promising in this case either. An interesting observation is that the winning threshold is never reached (it would be reached with much higher repetitions but clearly, the probability is low. And the 1000 turn limit was actually used up in case of positive profits.). Perhaps more turns could deliver a full 1000\$ profit? Let's explore that below.

```

set.seed(2)
d <- PResult(20, 10000, 1000, 2000, 200)
d1 <- as.data.frame(d)
d1 <- mutate(d1, GameIndex = 1:length(d))
colnames(d1) <- c("Profit", "GameIndex")
ggplot(data = d1, aes(x = GameIndex, y = Profit)) + geom_histogram(stat = "identity") + labs(title = "A

```

```
## Warning: Ignoring unknown parameters: binwidth, bins, pad
```



```

d2<- PResult(1000, 1000, 200, 500, 100)
d2 <- as.data.frame(d2)
d2 <- mutate(d2, GameIndex = 1:length(d2))
colnames(d2) <- c("Profit", "GameIndex")
mean(d2$Profit)

```

```
## [1] -100.776
```

As expected, with a higher number of chances, we are able to reach the full 1000\$ profit(1000\$ being the difference between the winning_threshold and the starting_budget) but we still end up with negative mean earnings of -100.76.

The below chunk of code calculates the average number of plays before a stopping condition is achieved.

```

NumberofPlays <- function(ledger){
  n <- nrow(ledger)
  return(n)
}

AverageNumberofPlays <- function(r, max_games, starting_budget, winning_threshold, max_wager)
{
  rep <- c(replicate(r, NumberofPlays(one_series(max_games, starting_budget, winning_threshold, max_wager)
  return(mean(rep))
}

```

```

AverageNumberofPlays(1000, 1000, 200, 300, 100)

```

```
## [1] 202.771
```

Here we find that the average number of plays using the initial given parameters is 208.807

#Conclusion

Using the above observations, we can clearly see that Martingale's strategy while not a slam dunk in every scenario. This can be extended to state that based on a set of static conditions, Martingale's strategy is not a definite win strategy. It can however be considered a definite win strategy if there are no constraints(i.e, no budget limitations, no betting limitations etc) or if the constraints the dynamic (i.e, the starting budget can be completely controlled by the player as the game progresses - this alone could greatly increase the winning chances if the number of plays or the betting limit is high enough.). The more dynamic constraints, the faster/less number of turns it'll take to gain a positive outcome. Dynamically changing the winning threshold based on previous turns by the player also would guarantee a success provided the dynamic starting budget and reasonably high max number of turns.

““