UFC RESEARCH QUESTION 1

Due 8th December, 2024

```
ufc <- read.csv("ufc-master.csv")

# View the data
head(ufc)

# data cleaning cell</pre>
```

data cleaning cell
colSums(is.na(ufc))

BlueFighter
0
BlueOdds
219
${ t Blue Expected Value}$
219
Location
0
Winner
0
WeightClass
0
NumberOfRounds
0
${\tt BlueCurrentWinStreak}$
0
${ t Blue Avg Sig Str Landed}$
930
BlueAvgSubAtt
832
${ t Blue Avg TDPct}$
842

${\tt BlueLongestWinStreak}$	BlueLosses
0	0
${\tt BlueTotalRoundsFought}$	${\tt BlueTotalTitleBouts}$
0	0
BlueWinsByDecisionMajority	BlueWinsByDecisionSplit
0	0
BlueWinsByDecisionUnanimous	BlueWinsByKO
0	O
BluewinsbySubmission O	BlueWinsByTKODoctorStoppage 0
BlueWins	BlueStance
Didewins 0	Didestance 0
BlueHeightCms	BlueReachCms
0	0
BlueWeightLbs	RedCurrentLoseStreak
0	0
${\tt RedCurrentWinStreak}$	RedDraws
0	0
${\tt RedAvgSigStrLanded}$	${\tt RedAvgSigStrPct}$
455	357
RedAvgSubAtt	${\tt RedAvgTDLanded}$
357	357
RedAvgTDPct	RedLongestWinStreak
367	0
RedLosses	RedTotalRoundsFought
0 RedTotalTitleBouts	0 PodMingPyDocidionMajority
redictaliftlebouts	RedWinsByDecisionMajority
RedWinsByDecisionSplit	RedWinsByDecisionUnanimous
0	0
RedWinsByKO	RedWinsBySubmission
0	0
${\tt RedWinsByTKODoctorStoppage}$	RedWins
0	0
RedStance	${\tt RedHeightCms}$
0	0
RedReachCms	${\tt RedWeightLbs}$
0	0
RedAge	BlueAge
0	UinCtmaal-Dif
LoseStreakDif O	WinStreakDif 0
LongestWinStreakDif	WinDif
rongeschingclegkhil	WINDII

0	0
LossDif	${\tt TotalRoundDif}$
0	0
${ t TotalTitleBoutDif}$	KODif
0	0
SubDif	${\tt HeightDif}$
0	0
ReachDif	AgeDif
0	0
${ t SigStrDif}$	${ t AvgSubAttDif}$
0	0
AvgTDDif	EmptyArena
0	1436
BMatchWCRank	RMatchWCRank
5289	4716
RWFlyweightRank	RWFeatherweightRank
6382	6469
RWStrawweightRank	RWBantamweightRank 6324
6334 RHeavyweightRank	
6295	RLightHeavyweightRank 6296
RMiddleweightRank	RWelterweightRank
6296	6290
RLightweightRank	RFeatherweightRank
6295	6303
RBantamweightRank	RFlyweightRank
6299	6292
RPFPRank	BWFlyweightRank
6228	6406
${\tt BWFeatherweightRank}$	${\tt BWStrawweightRank}$
6477	6380
${\tt BWBantamweightRank}$	${\tt BHeavyweightRank}$
6371	6332
${\tt BLightHeavyweightRank}$	${\tt BMiddleweightRank}$
6360	6341
${\tt BWelterweightRank}$	${ t BLightweightRank}$
6360	6359
BFeatherweightRank	BBantamweightRank
6355	6360
BFlyweightRank	BPFPRank
6348	6411
BetterRank	Finish
0	0

```
FinishRoundTime
                                     TotalFightTimeSecs
                 RedDecOdds
                                            BlueDecOdds
                       1077
                                                    1107
                   RSub0dds
                                                BSub0dds
                       1326
                                                    1350
                    RK00dds
                                                 BK00dds
                       1324
                                                    1351
# removing the data which has way too many missing values
ufc = subset(ufc, select = -c(BMatchWCRank, RMatchWCRank, RWFlyweightRank,
                              RWFeatherweightRank, RWStrawweightRank, RWBantamweightRank,
                              RHeavyweightRank, RLightHeavyweightRank, RMiddleweightRank,
                              RWelterweightRank, RLightweightRank, RFeatherweightRank,
                              RBantamweightRank, RFlyweightRank, RPFPRank, BWFlyweightRank,
                              BWFeatherweightRank, BWStrawweightRank, BWBantamweightRank,
                              BHeavyweightRank, BLightHeavyweightRank, BMiddleweightRank,
                              BWelterweightRank, BLightweightRank, BFeatherweightRank,
                              BBantamweightRank, BFlyweightRank, BPFPRank))
library(dplyr)
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
# removing all missing value rows from the columns of interest
ufc_clean <- ufc %>%
  filter(
    !is.na(RedAvgSubAtt),
    !is.na(BlueAvgSubAtt),
```

FinishRound

622

FinishDetails

```
!is.na(BlueReachCms),
!is.na(RedReachCms),
!is.na(BlueAvgSigStrLanded),
!is.na(RedAvgSigStrLanded),
!is.na(TotalFightTimeSecs),
!is.na(WeightClass)
)
nrow(ufc_clean)
```

[1] 4895

```
filtered_ufc_blue <- ufc_clean[c("BlueReachCms", "BlueAvgSigStrLanded",
                                  "WeightClass", "BlueHeightCms",
                                  "BlueCurrentWinStreak")]
colnames(filtered_ufc_blue) <- c("ReachCms", "AvgSigStrLanded", "WeightClass",
                                  "Height", "WinStreak")
filtered_ufc_red <- ufc_clean[c("RedReachCms", "RedAvgSigStrLanded",
                                 "WeightClass", "RedHeightCms", "RedCurrentWinStreak")]
colnames(filtered_ufc_red) <- c("ReachCms", "AvgSigStrLanded",</pre>
                                 "WeightClass", "Height", "WinStreak")
# appending the two data sets
ufc_q1 <- rbind(filtered_ufc_blue, filtered_ufc_red)</pre>
# exclude outlier(one observation with 0 cm reach)
ufc_q1 <- ufc_q1[ufc_q1$ReachCms > 0,]
ufc_q1 <- ufc_q1[ufc_q1$AvgSigStrLanded > 0, ]
# check missing value: no missing
colSums(is.na(ufc_q1))
```

```
ReachCms AvgSigStrLanded WeightClass Height WinStreak
0 0 0 0 0
```

```
# Log-transform the variables
ufc_q1$LogAvgSigStrLanded <- log(ufc_q1$AvgSigStrLanded)
ufc_q1$LogReachCms <- log(ufc_q1$ReachCms)
model_q1 <- lm(LogAvgSigStrLanded ~ LogReachCms * WeightClass + Height + WinStreak, data = usummary(model_q1)</pre>
```

Call:

Residuals:

Min 1Q Median 3Q Max -4.2411 -1.0121 0.1562 0.9527 3.0385

Coefficients:

Odeilicienus.			_
		Std. Error	
(Intercept)	20.299645	5.245888	3.870
LogReachCms	-3.366757	1.035627	-3.251
WeightClassCatch Weight	-9.847861	12.912028	-0.763
WeightClassFeatherweight	16.755691	7.550510	2.219
WeightClassFlyweight	57.696594	9.120653	6.326
WeightClassHeavyweight	-13.523345	7.987021	-1.693
WeightClassLight Heavyweight	-14.172206	8.268269	-1.714
WeightClassLightweight	-0.404508	6.949375	-0.058
WeightClassMiddleweight	-5.138435	7.523066	-0.683
WeightClassWelterweight	-1.955368	6.981548	-0.280
WeightClassWomen's Bantamweight	-10.914377	11.779043	-0.927
WeightClassWomen's Featherweight	-1.750844	31.478949	-0.056
WeightClassWomen's Flyweight	-23.948120	10.169350	-2.355
WeightClassWomen's Strawweight	-42.488577	9.154564	-4.641
Height	-0.003163	0.003207	-0.986
WinStreak	0.041084	0.006774	6.065
LogReachCms:WeightClassCatch Weight	1.781285	2.487705	0.716
LogReachCms:WeightClassFeatherweight	-3.188399	1.458418	-2.186
LogReachCms:WeightClassFlyweight	-11.267528	1.772449	-6.357
LogReachCms:WeightClassHeavyweight	2.646695	1.527269	1.733
LogReachCms:WeightClassLight Heavyweight	2.794438	1.582216	1.766
LogReachCms:WeightClassLightweight	0.154980	1.341615	0.116
LogReachCms:WeightClassMiddleweight	1.050072	1.444678	0.727
LogReachCms:WeightClassWelterweight	0.474180	1.344581	0.353
LogReachCms:WeightClassWomen's Bantamweight	2.093198	2.290004	0.914
LogReachCms:WeightClassWomen's Featherweight	0.143693	6.098032	0.024
LogReachCms:WeightClassWomen's Flyweight	4.511495	1.979945	2.279
LogReachCms:WeightClassWomen's Strawweight	8.272935	1.790830	4.620
	Pr(> t)		
(Intercept)	0.00011 **	*	

 (Intercept)
 0.00011 ***

 LogReachCms
 0.00115 **

 WeightClassCatch Weight
 0.44567

```
WeightClassFeatherweight
                                              0.02650 *
                                             2.63e-10 ***
WeightClassFlyweight
WeightClassHeavyweight
                                              0.09046 .
WeightClassLight Heavyweight
                                              0.08655 .
WeightClassLightweight
                                              0.95358
WeightClassMiddleweight
                                              0.49461
WeightClassWelterweight
                                              0.77942
WeightClassWomen's Bantamweight
                                              0.35416
WeightClassWomen's Featherweight
                                              0.95565
WeightClassWomen's Flyweight
                                              0.01855 *
WeightClassWomen's Strawweight
                                             3.51e-06 ***
Height
                                              0.32408
                                             1.37e-09 ***
WinStreak
LogReachCms:WeightClassCatch Weight
                                              0.47399
LogReachCms:WeightClassFeatherweight
                                              0.02882 *
LogReachCms:WeightClassFlyweight
                                             2.15e-10 ***
LogReachCms:WeightClassHeavyweight
                                              0.08313 .
LogReachCms:WeightClassLight Heavyweight
                                              0.07740 .
LogReachCms:WeightClassLightweight
                                              0.90804
LogReachCms:WeightClassMiddleweight
                                              0.46733
LogReachCms:WeightClassWelterweight
                                              0.72435
LogReachCms:WeightClassWomen's Bantamweight
                                              0.36071
LogReachCms:WeightClassWomen's Featherweight 0.98120
LogReachCms:WeightClassWomen's Flyweight
                                              0.02271 *
LogReachCms:WeightClassWomen's Strawweight
                                             3.89e-06 ***
Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
Residual standard error: 1.108 on 9707 degrees of freedom
Multiple R-squared: 0.05948,
                                Adjusted R-squared: 0.05686
F-statistic: 22.74 on 27 and 9707 DF, p-value: < 2.2e-16
```

```
# Load necessary libraries
library(car) # For VIF
```

Loading required package: carData

Attaching package: 'car'

The following object is masked from 'package:dplyr':

recode

```
library(ggplot2)  # For residual plots

# 1. Check Variance Inflation Factor (VIF) for collinearity
vif_values <- vif(model_q1)</pre>
```

there are higher-order terms (interactions) in this model consider setting type = 'predictor'; see ?vif

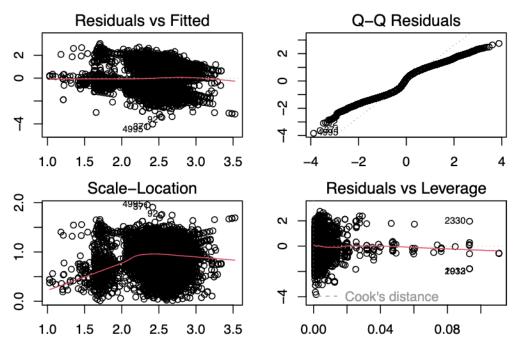
```
print("Variance Inflation Factor (VIF):")
```

[1] "Variance Inflation Factor (VIF):"

print(vif_values)

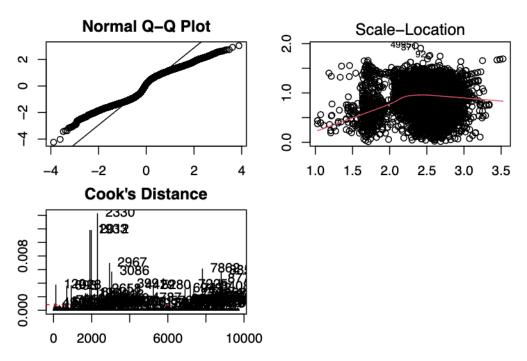
```
GVIF Df GVIF^(1/(2*Df))
LogReachCms 3.128002e+01 1 5.592854
WeightClass 2.148248e+52 12 151.531708
Height 6.742370e+00 1 2.596607
WinStreak 1.010645e+00 1 1.005308
LogReachCms:WeightClass 2.184107e+52 12 151.636265
```

```
# 2. Residuals vs Fitted Plot for Linearity
par(mfrow = c(2, 2), mar = c(2, 2, 2)) # Set plotting layout
plot(model_q1)
```



```
# 3. Normal Q-Q Plot for Normality of Residuals
qqnorm(residuals(model_q1))
qqline(residuals(model_q1))
# 4. Scale-Location Plot for Homoscedasticity
plot(model_q1, which = 3)
# 5. Check for influential points using Cook's Distance
cooksd <- cooks.distance(model_q1)</pre>
plot(cooksd, type = "h", main = "Cook's Distance", ylab = "Cook's Distance")
# Highlight observations with Cook's Distance > threshold
threshold <- 4 / nrow(ufc_clean)</pre>
influential <- which(cooksd > threshold)
abline(h = threshold, col = "red", lty = 2)
text(x = influential, y = cooksd[influential], labels = names(cooksd[influential]), pos = 4)
# 6. R-squared value
r_squared <- summary(model_q1)$r.squared
cat("R-squared:", r_squared, "\n")
```

R-squared: 0.05947709



```
# Load necessary library
library(knitr)

# Create a summary of the model
model_summary <- summary(model_q1)

# Extract coefficients and format into a data frame
coef_table <- as.data.frame(model_summary$coefficients)
colnames(coef_table) <- c("Estimate", "Std. Error", "t value", "Pr(>|t|)")

# Round to 3 decimal places
coef_table <- round(coef_table, 3)

# Create a kable table
kable(coef_table, caption = "Regression Coefficients for model_log", format = "markdown")</pre>
```

Table 1: Regression Coefficients for model_log

		Std.		
	Estimate	Error	t value	$\Pr(> t)$
(Intercept)	20.300	5.246	3.870	0.000
LogReachCms	-3.367	1.036	-3.251	0.001

	Std.			
	Estimate	Error	t value	$\Pr(> t)$
WeightClassCatch Weight	-9.848	12.912	-0.763	0.446
WeightClassFeatherweight	16.756	7.551	2.219	0.026
WeightClassFlyweight	57.697	9.121	6.326	0.000
WeightClassHeavyweight	-13.523	7.987	-1.693	0.090
WeightClassLight Heavyweight	-14.172	8.268	-1.714	0.087
WeightClassLightweight	-0.405	6.949	-0.058	0.954
Weight Class Middle weight	-5.138	7.523	-0.683	0.495
WeightClassWelterweight	-1.955	6.982	-0.280	0.779
WeightClassWomen's Bantamweight	-10.914	11.779	-0.927	0.354
WeightClassWomen's Featherweight	-1.751	31.479	-0.056	0.956
WeightClassWomen's Flyweight	-23.948	10.169	-2.355	0.019
WeightClassWomen's Strawweight	-42.489	9.155	-4.641	0.000
Height	-0.003	0.003	-0.986	0.324
WinStreak	0.041	0.007	6.065	0.000
LogReachCms:WeightClassCatch	1.781	2.488	0.716	0.474
Weight				
LogReach Cms: Weight Class Feather weight	-3.188	1.458	-2.186	0.029
LogReachCms:WeightClassFlyweight	-11.268	1.772	-6.357	0.000
LogReachCms:WeightClassHeavyweight	2.647	1.527	1.733	0.083
LogReachCms:WeightClassLight	2.794	1.582	1.766	0.077
Heavyweight				
LogReachCms:WeightClassLightweight	0.155	1.342	0.116	0.908
LogReachCms:WeightClassMiddleweight	1.050	1.445	0.727	0.467
LogReachCms:WeightClassWelterweight	0.474	1.345	0.353	0.724
LogReachCms: WeightClassWomen's	2.093	2.290	0.914	0.361
Bantamweight				
LogReachCms: WeightClassWomen's	0.144	6.098	0.024	0.981
Featherweight				
LogReachCms: WeightClassWomen's	4.511	1.980	2.279	0.023
Flyweight				
LogReachCms: WeightClassWomen's Strawweight	8.273	1.791	4.620	0.000