

The Relationship Between Exercise Time and Weight Loss

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```
library(tidyr)
```

```
## Warning: package 'tidyr' was built under R version 3.6.2
```

```
library(ggplot2)
```

```
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
```

```
library(reshape2)
```

```
##
```

```
## Attaching package: 'reshape2'
```

```
## The following object is masked from 'package:tidyr':
```

```
##
```

```
##      smiths
```

Abstract

The purpose of this data analysis is to apply statistical models to health data for a call center. While there were many data points missing for weight gain and the total metabolic minutes, many of these could be found using the other metrics available. The health data was processed to provide as many viable data points as possible without compromising the integrity of the analysis. Two models were constructed: a linear model to examine the relationship between total metabolic minutes and weight gain and a logistic regression model to examine the relationship between shift time and weight gain.

Introduction

We have been given several health measurements of employees working at a call center. Over eight months, the metrics have tracked information including work shift time, exercise time, weight gained (binary yes/no measurement), amount of weight gained, total metabolic minutes, and more. We hope to figure out whether certain factors, specifically total metabolic minutes and shift time, play a role in employees' weight gain.

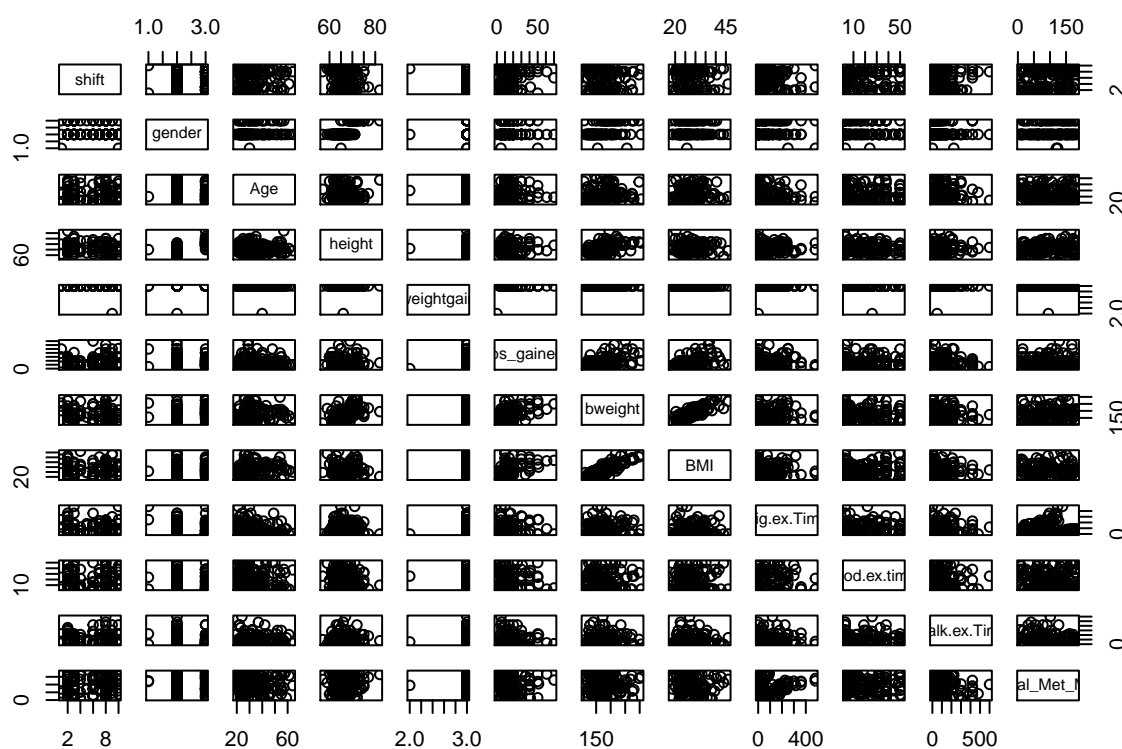
The Data

The data used in this analysis was provided by the call center. The data consists of 392 observations and 83 features, with variables such as member information, body measurements, exercise time, and weight situation. There are thirteen variables that we will use for this analysis, including shift time, gender, age, height, weight gain, pounds gained, body weight, body mass index, vigorous exercise time, moderate exercise time, walking exercise time, and total metabolic minutes. It is important to note that there were many missing values for total metabolic minutes, but these could be filled in using the other variables available. There were also missing values for pounds gained, but no features were available to account for this so these observations were removed.

We load and subset the raw data into a new dataset that contains the columns that we will use for our analysis. After the initial examination of the data, we renamed several variables by eliminating white spaces in order to improve the easiness of analysis. Using the formula provided, $\text{Total_met_min} = 8 * \text{Vig_ex_time} + 4 * \text{Mod_ex_time} + 3.3 * \text{Walk_ex_time}$, we filled in the missing values of variable `Total_Met_Min`.

```
### Already a numeric variable? No need to change
dat1$Total_Met_Min = as.numeric(dat1$Total_Met_Min)
```

```
pairs(dat1)
```

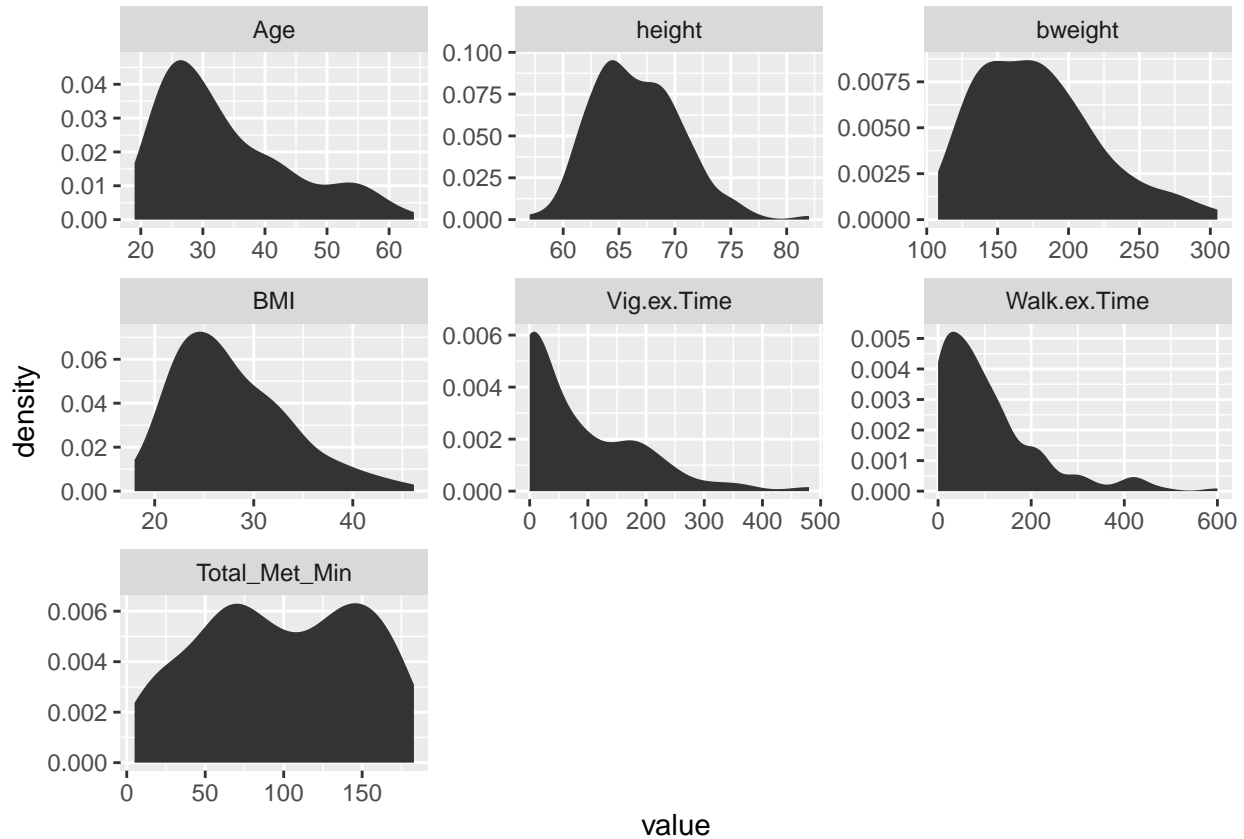


```
dat1.try <- dat1[c(3:4, 7:12)]
dat1.try <- melt(dat1.try)
```

```
## Using Mod.ex.time as id variables
```

```
ggplot(data = dat1.try, aes(x = value)) + stat_density() + facet_wrap(~variable, scales = 'free')
```

```
## Warning: Removed 82 rows containing non-finite values (stat_density).
```



We see that Age and BMI are right-skewed with nonzero values, so we will do a log transformation on these variables.

```
dat1$Age <- log(dat1$Age)
dat1$BMI <- log(dat1$BMI)
```

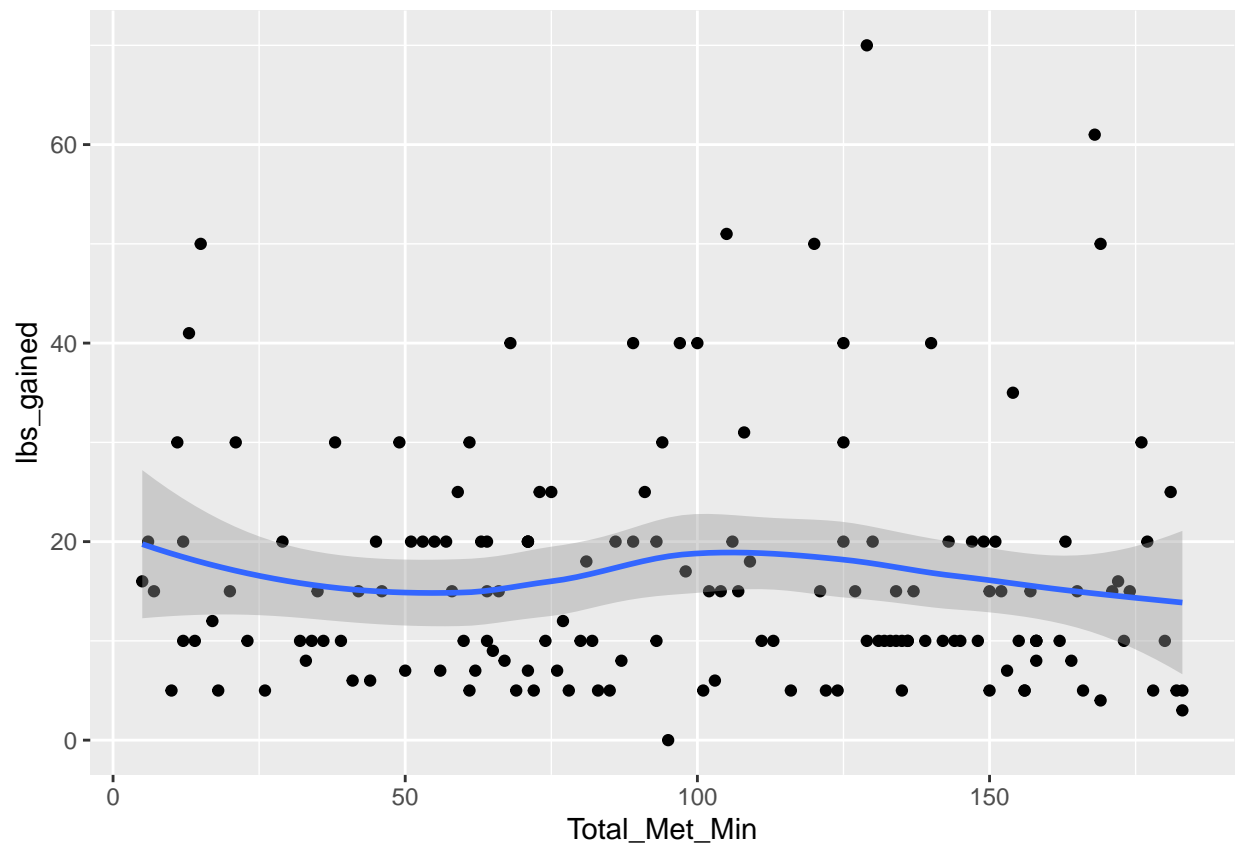
```
lm = lm(lbs_gained ~ Total_Met_Min, data = dat1)
summary(lm)
```

```
##
## Call:
## lm(formula = lbs_gained ~ Total_Met_Min, data = dat1)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -16.406  -7.288  -1.913   3.543  53.824
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
```

```
## (Intercept) 17.048084 2.115141 8.060 2.03e-13 ***
## Total_Met_Min -0.006761 0.019050 -0.355 0.723
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 12.05 on 153 degrees of freedom
## Multiple R-squared: 0.0008225, Adjusted R-squared: -0.005708
## F-statistic: 0.126 on 1 and 153 DF, p-value: 0.7232
```

```
ggplot(data = dat1, aes(x = Total_Met_Min, y = lbs_gained)) + geom_point() + geom_smooth()
```

```
## `geom_smooth()` using method = 'loess' and formula 'y ~ x'
```



```
i = 1
for (obs in dat1$lbs_gained) {
  if (obs > 45) {
    dat1 = dat1[-i,]
  } else {
    i = i + 1
  }
}
```

```
i = 1
for (obs in dat1$Total_Met_Min) {
```

```

if (obs >= 5000) {
  dat1 = dat1[-i,]
} else {
  i = i + 1
}
}

```

```
lm = lm(lbs_gained ~ Total_Met_Min, data = dat1)
```

```
summary(lm)
```

```

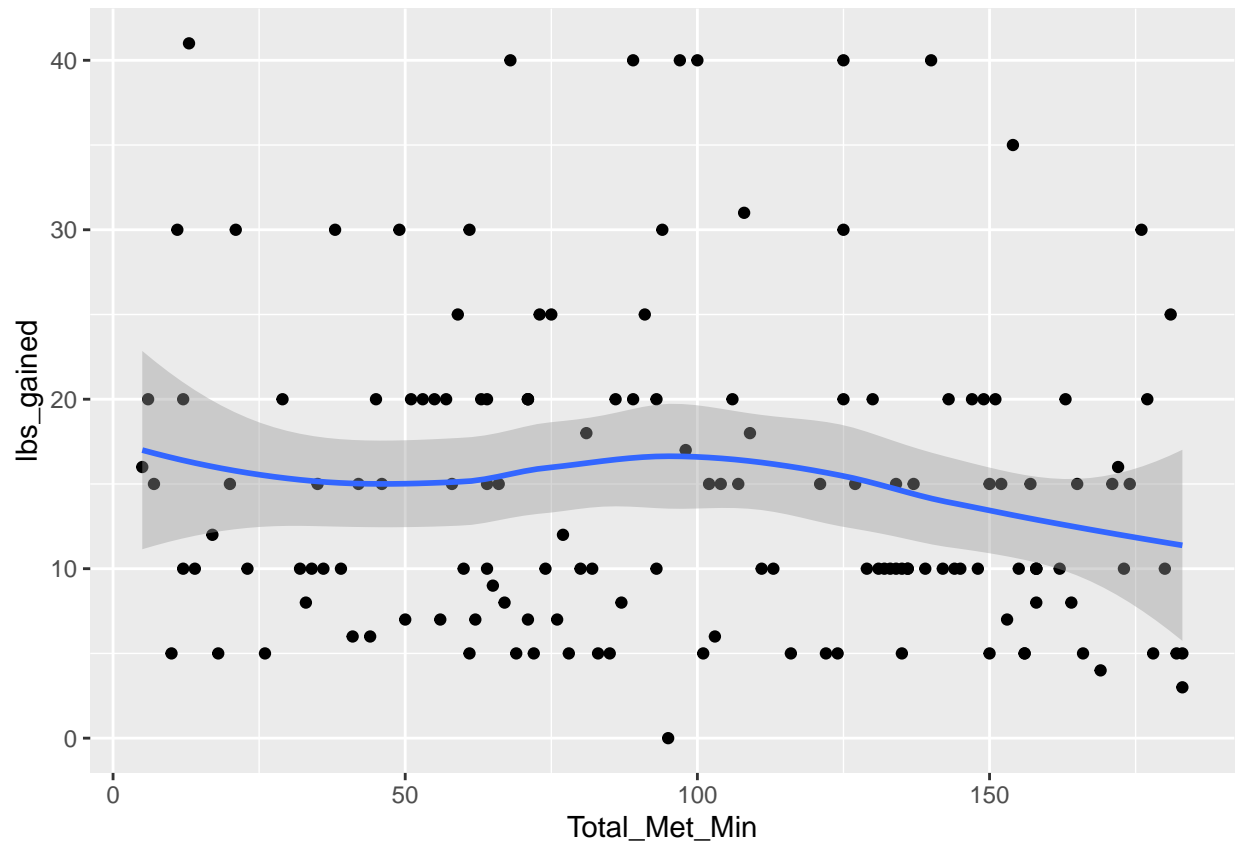
##
## Call:
## lm(formula = lbs_gained ~ Total_Met_Min, data = dat1)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -14.875  -6.588  -3.225   4.470  26.077
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  16.88414    1.62851  10.368  <2e-16 ***
## Total_Met_Min -0.02115    0.01477  -1.432   0.154
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 9.127 on 147 degrees of freedom
## Multiple R-squared:  0.01376,    Adjusted R-squared:  0.007054
## F-statistic: 2.051 on 1 and 147 DF,  p-value: 0.1542

```

P-value got bigger

```
ggplot(data = dat1, aes(x = Total_Met_Min, y = lbs_gained)) + geom_point() + geom_smooth()
```

```
## `geom_smooth()` using method = 'loess' and formula 'y ~ x'
```



Risheng: replace lbs_gained value with 0 if weightgain = 0 and drop the rows that both weightgain and lbs_gained are missing. Also calculate the beginning weight by subtracting lbs_gain from body weight. Calculate the change in BMI using beginning weight and height.

```
dat1$lbs_gained[dat1$weightgain == "No"] <- 0

dat1$beg_weight <- dat1$bweight - dat1$lbs_gained

dat1$Beg_BMI <- (dat1$beg_weight/(dat1$height^2))*703
dat1$BMChange <- dat1$BMI - dat1$Beg_BMI

dat1 <- dat1 %>%
  filter_at(vars(weightgain, lbs_gained), any_vars(complete.cases(.)))
```

By turning weightgain into a dummy variable and running a logistic regression, we can measure if other variables have influence on weightgain. First, we factored the two categorical variables and built a model with variables include shift, gender, Age, height and Beg_BMI and Total_Met_Min. Because the change in BMI is very minimal, we decided to use the beginning BMI as one of the regressors. The half normal plot shows that there is no obvious outliers in the model:

```
dat1$weightgain.b <- ifelse(dat1$weightgain == 'Yes', 1, 0)

dat1
```

##	shift	gender	Age	height	weightgain	lbs_gained	bweight	BMI
## 1	8am	Female	3.258097	62	Yes	18	124	3.121483
## 2	9am	Female	3.295837	65	Yes	6	151	3.223664
## 3	9am	Male	3.688879	69	Yes	10	180	3.280159
## 4	7am	Male	3.951244	72	Yes	20	190	3.249211
## 5	7am	Female	4.060443	62	Yes	20	NA	NA
## 6	other	Female	3.135494	NA	Yes	5	NA	NA
## 7	8am	Female	3.496508	62	Yes	15	155	3.344627
## 8	8am	Male	3.401197	75	Yes	8	160	2.995732
## 9	8am	Female	4.077537	63	Yes	10	150	3.279783
## 10	8am	Female	3.713572	68	Yes	10	205	3.439456
## 11	8am	Female	3.135494	64	Yes	10	NA	NA
## 12	8am	Male	3.091042	72	Yes	20	235	3.461665
## 13	7am	Male	3.988984	72	Yes	40	298	3.699077
## 14	8am	Male	3.988984	69	Yes	20	185	3.307619
## 15	11am	Female	3.737670	66	Yes	20	209	3.518388
## 16	7am	Male	3.713572	70	Yes	10	204	3.376563
## 17	other	Female	3.737670	63	Yes	5	129	3.128951
## 18	8am	Female	3.737670	57	Yes	10	113	3.196630
## 19	8am	Female	NA	NA	Yes	3	NA	NA
## 20	8am	Female	3.828641	64	Yes	15	140	3.179303
## 21	9am	Female	3.688879	66	No	0	NA	NA
## 22	8am	Male	3.401197	75	Yes	40	225	3.336481
## 23	9am	Female	3.218876	60	Yes	30	NA	NA
## 24	9am	Female	3.401197	61	Yes	15	135	3.239071
## 25	9am	Female	3.828641	62	Yes	20	120	3.088767
## 26	8am	Female	3.295837	64	Yes	7	NA	NA
## 27	10am	Female	3.258097	62	Yes	31	170	3.436886
## 28	10am	Male	3.135494	69	Yes	5	170	3.222868
## 29	11am	Male	4.007333	65	Yes	5	108	2.888704
## 30	9am	Female	3.091042	64	Yes	5	120	3.025291
## 31	8am	Male	3.218876	71	Yes	20	NA	NA
## 32	8am	Female	3.091042	64	Yes	7	132	3.120601
## 33	10am	Male	3.850148	68	Yes	25	200	3.414772
## 34	10am	Female	3.044522	64	Yes	5	130	3.105035
## 35	8am	Female	3.178054	67	Yes	5	173	3.299165
## 36		Female	3.526361	61	Yes	30	202	3.641788
## 37	other	Female	3.332205	65	Yes	5	135	3.111736
## 38	9am	Female	3.178054	63	Yes	10	NA	NA
## 39	12pm	Female	3.401197	64	Yes	10	269	3.832330
## 40	11am	Female	3.258097	63	Yes	8	132	3.151881
## 41	11am	Male	3.178054	69	Yes	8	150	3.097837
## 42	2pm	Female	3.044522	63	Yes	15	118	3.039749
## 43	1pm	Female	3.135494	67	Yes	5	155	3.189241
## 44	11am	Female	3.688879	61	Yes	41	NA	NA
## 45	2pm	Female	3.401197	66	Yes	10	140	3.117507
## 46	2pm	Female	3.465736	66	Yes	10	NA	NA
## 47	other	Female	3.332205	66	Yes	10	NA	NA
## 48	2pm	Male	NA	NA	Yes	10	173	NA
## 49	9am	Male	NA	NA	Yes	25	180	NA
## 50	8am	Female	3.332205	69	Yes	15	180	3.280159
## 51	2pm	Male	NA	72	Yes	30	250	3.523415
## 52	8am	Female	3.806662	62	Yes	18	167	3.419037
## 53	11am	Female	3.218876	67	Yes	20	155	3.189241

## 54	9am Female	3.295837	59	Yes	10	217	3.780090
## 55	other Female	4.007333	60	Yes	30	170	3.502550
## 56	8am Female	3.433987	63	Yes	7	NA	NA
## 57	8am Female	3.496508	64	Yes	10	NA	NA
## 58	8am Female	3.295837	67	Yes	20	145	3.122805
## 59	2pm Female	3.178054	65	Yes	17	NA	NA
## 60	8am Female	3.295837	69	Yes	15	NA	NA
## 61	1pm Female	3.688879	62	Yes	5	NA	NA
## 62	11am Female	3.583519	61	Yes	10	NA	NA
## 63	9am Female	3.401197	61	Yes	5	139	3.268047
## 64	10am Female	3.367296	63	Yes	30	235	3.728581
## 65	7am Female	3.218876	66	Yes	10	128	3.028199
## 66	8am Male	3.258097	69	Yes	20	220	3.480625
## 67	8am Male	3.433987	72	Yes	40	250	3.523415
## 68	12pm Male	3.401197	73	Yes	5	200	3.272606
## 69	8am Male	NA	NA	Yes	8	194	NA
## 70	8am Female	3.806662	68	Yes	20	200	3.414772
## 71	2pm Female	NA	61	Yes	20	135	3.239071
## 72	10am Female	3.367296	69	Yes	8	125	2.915606
## 73	10am Male	3.526361	70	Yes	15	275	3.675034
## 74	other Female	3.931826	68	Yes	10	207	3.449035
## 75	11am Male	3.258097	69	Yes	5	150	3.097837
## 76	1pm Male	3.218876	73	Yes	10	185	3.194993
## 77	12pm Female	4.110874	66	Yes	10	165	3.282038
## 78	11am Female	NA	71	Yes	20	NA	NA
## 79	12pm Female	3.178054	66	Yes	10	170	3.312002
## 80	8am Female	NA	68	Yes	20	NA	NA
## 81	Male	3.610918	70	Yes	15	235	3.518091
## 82	10am Male	4.043051	82	Yes	15	195	3.015045
## 83	8am Female	4.043051	67	Yes	10	188	3.382354
## 84	10am Female	4.007333	62	Yes	12	185	3.521348
## 85	9am Female	3.806662	68	Yes	10	165	3.222469
## 86	9am Female	3.610918	64	Yes	4	150	3.248046
## 87	12pm Female	3.637586	64	Yes	20	180	3.430433
## 88	10am Female	3.218876	67	Yes	10	178	3.327910
## 89	9am Female	3.465736	NA	Yes	5	135	NA
## 90	8am Male	3.465736	71	Yes	15	187	3.261169
## 91	7am Male	3.496508	68	Yes	20	180	3.309448
## 92	11am Female	3.761200	69	Yes	10	238	3.559340
## 93	8am Female	3.555348	64	Yes	30	NA	NA
## 94	10am Female	3.295837	62	Yes	20	NA	NA
## 95	8am Male	3.367296	70	Yes	20	188	3.294725
## 96	10am Female	3.218876	65	Yes	5	115	2.951258
## 97	8am Female	4.007333	67	Yes	30	NA	NA
## 98	11am Female	3.178054	68	Yes	10	140	3.057768
## 99	7am Female	3.688879	68	Yes	25	NA	NA
## 100	7am Female	3.091042	64	Yes	15	162	3.325036
## 101	7am Female	3.258097	62	Yes	15	138	3.228430
## 102	2pm Female	2.995732	67	Yes	7	120	2.933325
## 103	2pm Female	3.912023	68	Yes	20	NA	NA
## 104	10am Female	3.951244	64	Yes	12	208	3.575151
## 105	10am Female	3.091042	62	Yes	20	150	3.311637
## 106	10am Female	3.091042	64	Yes	20	170	3.373484
## 107	8am Female	3.610918	66	Yes	6	140	3.117507

## 108	11am	Male	3.465736	75	Yes	16	185	3.140698
## 109	8am	Male	3.218876	70	Yes	15	170	3.194173
## 110	12pm	Female	3.951244	63	Yes	16	185	3.489513
## 111	10am	Male	4.007333	72	Yes	10	210	3.349202
## 112	10am	Female	3.332205	67	Yes	5	NA	NA
## 113	8am	Female	NA	64	Yes	7	145	3.214466
## 114	8am	Female	3.555348	70	Yes	20	240	3.538928
## 115	11am	Male	3.688879	70	Yes	5	215	3.429137
## 116	8am	Female	3.465736	65	Yes	10	157	3.262701
## 117	7am	Female	3.737670	67	Yes	15	201	3.449352
## 118	8am	Female	4.158883	64	Yes	10	145	3.214466
## 119	8am	Female	3.091042	66	Yes	9	142	3.132010
## 120	8am	Female	3.178054	68	Yes	40	170	3.252311
## 121	11am	Male	3.891820	66	Yes	15	138	3.103240
## 122	8am	Female	3.713572	62	Yes	15	187	3.532226
## 123	8am	Female	3.688879	64	Yes	35	130	3.105035
## 124	10am	Female	3.465736	69	Yes	15	195	3.360028
## 125	10am	Female	3.135494	65	Yes	6	NA	NA
## 126	10am	Male	3.555348	70	Yes	10	273	3.667911
## 127	11am	Female	3.135494	68	Yes	5	145	3.092859
## 128	7am	Male	3.737670	69	Yes	10	210	3.434310
## 129	8am	Female	3.218876	64	Yes	20	133	3.128075
## 130	other	Male	3.218876	75	Yes	15	175	3.085116
## 131	9am	Female	3.433987	70	Yes	40	215	3.429137
## 132	8am	Female	2.944439	69	Yes	15	145	3.063858
## 133	10am	Female	3.433987	NA	Yes	5	185	NA
## 134	9am	Female	3.496508	65	Yes	10	165	3.312366
## 135	8am	Female	3.295837	63	Yes	30	183	3.478467
## 136	11am	Female	3.178054	66	Yes	10	140	3.117507
## 137	9am	Female	3.258097	64	Yes	7	125	3.065725
## 138	2pm	Male	3.555348	74	Yes	25	230	3.385407
## 139	10am	Female	3.135494	70	Yes	20	158	3.121042
## 140	8am	Female	3.526361	61	Yes	5	123	3.145875
## 141	9am	Female	3.258097	71	Yes	20	305	3.750210
## 142	8am	Male	3.258097	72	Yes	40	280	3.636796
## 143	10am	Male	3.091042	71	Yes	25	215	3.400530
## 144	11am	Male	3.044522	71	Yes	10	168	3.154017
## 145	8am	Female	3.332205	65	Yes	10	NA	NA
## 146	10am	Female	3.465736	67	Yes	15	165	3.251924
## 147	11am	Female	3.295837	66	Yes	15	NA	NA
## 148	other		3.401197	65	Yes	5	155	3.249987
## 149	11am	Male	3.555348	77	Yes	10	220	3.261552
##	Vig.ex.Time	Mod.ex.time	Walk.ex.Time	Total_Met_Min	beg_weight	Beg_BMI		
## 1	180	160	100	81	106	19.38554		
## 2	40	0	0	103	145	24.12663		
## 3	40	20	10	129	170	25.10187		
## 4	180	180	0	71	170	23.05363		
## 5	90	40	0	177	NA	NA		
## 6	0	150	20	156	NA	NA		
## 7	60	30	80	174	140	25.60354		
## 8	0	40	30	87	152	18.99662		
## 9	90	0	140	14	140	24.79718		
## 10	50	160	125	36	195	29.64641		
## 11	0	75	45	132	NA	NA		

## 12	45	0	420	55	215	29.15606
## 13	0	40	105	140	258	34.98727
## 14	0	0	10	106	165	24.36358
## 15	90	60	60	12	189	30.50207
## 16	225	360	70	111	194	27.83306
## 17	90	60	420	78	124	21.96321
## 18	0	30	100	133	103	22.28655
## 19	0	0	30	183	NA	NA
## 20	180	0	0	35	125	21.45386
## 21	0	30	50	95	NA	NA
## 22	210	210	105	97	185	23.12089
## 23	0	120	120	176	NA	NA
## 24	0	0	60	64	120	22.67132
## 25	0	0	49	45	100	18.28824
## 26	20	0	20	76	NA	NA
## 27	300	80	189	108	139	25.42066
## 28	100	60	90	26	165	24.36358
## 29	75	0	20	156	103	17.13822
## 30	300	300	60	122	115	19.73755
## 31	60	0	90	163	NA	NA
## 32	160	60	80	56	125	21.45386
## 33	0	0	68	75	175	26.60575
## 34	180	30	100	61	125	21.45386
## 35	120	30	600	101	168	26.30965
## 36	0	0	35	11	172	32.49557
## 37	30	60	100	166	130	21.63077
## 38	0	90	150	173	NA	NA
## 39	0	0	140	134	259	44.45239
## 40	180	91	70	67	124	21.96321
## 41	60	180	60	33	142	20.96744
## 42	0	1680	0	150	103	18.24364
## 43	150	150	90	69	150	23.49076
## 44	60	50	150	13	NA	NA
## 45	0	120	0	136	130	20.98026
## 46	0	0	20	155	NA	NA
## 47	180	75	35	60	NA	NA
## 48	0	0	420	32	163	NA
## 49	160	0	420	91	155	NA
## 50	80	0	60	171	165	24.36358
## 51	160	0	0	21	220	29.83410
## 52	0	20	80	109	149	27.24948
## 53	0	60	60	130	135	21.14168
## 54	0	9	0	113	207	41.80437
## 55	60	210	60	38	140	27.33889
## 56	80	0	0	153	NA	NA
## 57	0	135	0	144	NA	NA
## 58	0	0	120	125	125	19.57563
## 59	360	0	0	98	NA	NA
## 60	15	10	0	42	NA	NA
## 61	16	16	90	135	NA	NA
## 62	0	0	135	131	NA	NA
## 63	90	40	30	182	134	25.31631
## 64	315	60	16	94	205	36.31015
## 65	0	0	210	158	118	19.04362

## 66	180	180	0	71	200	29.53161
## 67	200	240	140	100	210	28.47801
## 68	60	60	60	178	195	25.72434
## 69	0	0	210	158	186	NA
## 70	0	100	60	149	180	27.36592
## 71	40	30	280	29	115	21.72669
## 72	90	9	9	164	117	17.27599
## 73	200	60	0	58	260	37.30204
## 74	0	0	225	162	197	29.95048
## 75	0	30	30	72	145	21.41042
## 76	180	150	50	74	175	23.08594
## 77	0	0	210	158	155	25.01492
## 78	0	20	60	93	NA	NA
## 79	0	240	60	12	160	25.82185
## 80	90	60	20	6	NA	NA
## 81	120	120	60	46	220	31.56327
## 82	0	140	75	165	180	18.81916
## 83	0	60	0	80	178	27.87570
## 84	225	0	140	77	173	31.63866
## 85	60	30	210	23	155	23.56510
## 86	0	90	140	169	146	25.05811
## 87	120	120	80	51	160	27.46094
## 88	0	45	90	135	168	26.30965
## 89	0	0	30	183	130	NA
## 90	135	30	24	20	172	23.98651
## 91	180	180	0	71	160	24.32526
## 92	165	90	225	82	228	33.66604
## 93	0	0	120	125	NA	NA
## 94	0	0	60	64	NA	NA
## 95	0	60	90	143	168	24.10286
## 96	0	15	0	150	110	18.30296
## 97	0	225	300	61	NA	NA
## 98	0	0	60	64	130	19.76427
## 99	120	105	140	59	NA	NA
## 100	225	210	420	127	147	25.22974
## 101	60	0	60	157	123	22.49454
## 102	180	180	0	71	113	17.69637
## 103	0	90	60	147	NA	NA
## 104	0	60	300	17	196	33.63965
## 105	0	360	120	57	130	23.77471
## 106	0	0	80	89	150	25.74463
## 107	0	0	480	41	134	21.62580
## 108	12	16	210	172	169	21.12124
## 109	0	135	20	152	155	22.23776
## 110	0	140	140	5	169	29.93374
## 111	60	120	0	180	200	27.12191
## 112	240	40	120	83	NA	NA
## 113	120	100	100	50	138	23.68506
## 114	240	120	40	86	220	31.56327
## 115	0	30	40	85	210	30.12857
## 116	60	0	0	136	147	24.45941
## 117	0	0	100	107	186	29.12854
## 118	8	56	20	93	135	23.17017
## 119	200	40	75	65	133	21.46442

## 120	0	0	120	125	130 19.76427
## 121	360	120	120	121	123 19.85055
## 122	45	240	210	66	172 31.45578
## 123	0	40	150	154	95 16.30493
## 124	180	210	300	104	180 26.57845
## 125	30	45	360	44	NA NA
## 126	0	0	210	158	263 37.73245
## 127	50	75	135	10	140 21.28460
## 128	30	60	20	145	200 29.53161
## 129	135	90	160	63	113 19.39429
## 130	30	45	20	137	160 19.99644
## 131	0	0	80	89	175 25.10714
## 132	240	120	210	102	130 19.19555
## 133	360	0	240	116	180 NA
## 134	0	0	160	142	155 25.79053
## 135	0	30	15	49	153 27.09977
## 136	180	20	0	39	130 20.98026
## 137	0	240	300	62	118 20.25244
## 138	135	105	210	73	205 26.31757
## 139	180	20	60	53	138 19.79878
## 140	120	40	40	18	118 22.29347
## 141	75	0	0	151	285 39.74509
## 142	240	30	0	68	240 32.54630
## 143	60	90	40	181	190 26.49673
## 144	480	240	60	139	158 22.03412
## 145	90	90	105	34	NA NA
## 146	0	0	315	7	150 23.49076
## 147	0	0	140	134	NA NA
## 148	480	0	0	124	150 24.95858
## 149	0	0	180	148	210 24.89965
##	BMChange weightgain.b				
## 1	-16.26405	1			
## 2	-20.90296	1			
## 3	-21.82171	1			
## 4	-19.80442	1			
## 5	NA	1			
## 6	NA	1			
## 7	-22.25891	1			
## 8	-16.00089	1			
## 9	-21.51740	1			
## 10	-26.20695	1			
## 11	NA	1			
## 12	-25.69439	1			
## 13	-31.28819	1			
## 14	-21.05596	1			
## 15	-26.98368	1			
## 16	-24.45650	1			
## 17	-18.83426	1			
## 18	-19.08992	1			
## 19	NA	1			
## 20	-18.27455	1			
## 21	NA	0			
## 22	-19.78441	1			
## 23	NA	1			

## 24	-19.43225	1
## 25	-15.19947	1
## 26	NA	1
## 27	-21.98377	1
## 28	-21.14071	1
## 29	-14.24952	1
## 30	-16.71226	1
## 31	NA	1
## 32	-18.33326	1
## 33	-23.19098	1
## 34	-18.34882	1
## 35	-23.01048	1
## 36	-28.85378	1
## 37	-18.51903	1
## 38	NA	1
## 39	-40.62006	1
## 40	-18.81133	1
## 41	-17.86961	1
## 42	-15.20389	1
## 43	-20.30151	1
## 44	NA	1
## 45	-17.86275	1
## 46	NA	1
## 47	NA	1
## 48	NA	1
## 49	NA	1
## 50	-21.08342	1
## 51	-26.31069	1
## 52	-23.83044	1
## 53	-17.95244	1
## 54	-38.02428	1
## 55	-23.83634	1
## 56	NA	1
## 57	NA	1
## 58	-16.45282	1
## 59	NA	1
## 60	NA	1
## 61	NA	1
## 62	NA	1
## 63	-22.04827	1
## 64	-32.58157	1
## 65	-16.01542	1
## 66	-26.05099	1
## 67	-24.95459	1
## 68	-22.45173	1
## 69	NA	1
## 70	-23.95115	1
## 71	-18.48762	1
## 72	-14.36039	1
## 73	-33.62701	1
## 74	-26.50144	1
## 75	-18.31258	1
## 76	-19.89095	1
## 77	-21.73288	1

## 78	NA	1
## 79	-22.50985	1
## 80	NA	1
## 81	-28.04517	1
## 82	-15.80411	1
## 83	-24.49334	1
## 84	-28.11731	1
## 85	-20.34263	1
## 86	-21.81006	1
## 87	-24.03050	1
## 88	-22.98174	1
## 89	NA	1
## 90	-20.72534	1
## 91	-21.01581	1
## 92	-30.10670	1
## 93	NA	1
## 94	NA	1
## 95	-20.80813	1
## 96	-15.35170	1
## 97	NA	1
## 98	-16.70651	1
## 99	NA	1
## 100	-21.90470	1
## 101	-19.26611	1
## 102	-14.76304	1
## 103	NA	1
## 104	-30.06450	1
## 105	-20.46308	1
## 106	-22.37115	1
## 107	-18.50830	1
## 108	-17.98055	1
## 109	-19.04358	1
## 110	-26.44422	1
## 111	-23.77271	1
## 112	NA	1
## 113	-20.47059	1
## 114	-28.02434	1
## 115	-26.69943	1
## 116	-21.19671	1
## 117	-25.67918	1
## 118	-19.95570	1
## 119	-18.33241	1
## 120	-16.51196	1
## 121	-16.74731	1
## 122	-27.92355	1
## 123	-13.19990	1
## 124	-23.21842	1
## 125	NA	1
## 126	-34.06454	1
## 127	-18.19174	1
## 128	-26.09730	1
## 129	-16.26621	1
## 130	-16.91133	1
## 131	-21.67801	1

```
## 132 -16.13169      1
## 133      NA      1
## 134 -22.47817      1
## 135 -23.62131      1
## 136 -17.86275      1
## 137 -17.18672      1
## 138 -22.93216      1
## 139 -16.67773      1
## 140 -19.14759      1
## 141 -35.99488      1
## 142 -28.90950      1
## 143 -23.09620      1
## 144 -18.88010      1
## 145      NA      1
## 146 -20.23883      1
## 147      NA      1
## 148 -21.70859      1
## 149 -21.63809      1
```

```
head(dat)
```

```
##      Snumber Date_Started      dept      Job shift gender Age height head
## 1         1      4/6/2005 Training      Other   8am Female  30     66    3
## 2         2      4/10/2005 Training      Other   7am Female  31     65    0
## 3         3      4/11/2005 Training      Other   8am Female  26     62    0
## 4         4      4/11/2005 Training      Other   7am Female  34     62    5
## 5         5      4/11/2005      CS      Other   9am Female  27     65    0
## 6         6      4/14/2005      CFS Collections 11am   Male  21     74    0
##      neck rshoul relbow rwrist lback rleg rknee rfoot eyes uback lshould
## 1         5         5         0         0         5         0         0         0         3         5         5
## 2         5         0         0         0         4         0         4         0         0         0         0
## 3         3         3         0         3         1         0         0         0         4         3         3
## 4         5         4         2         3         0         0         2         1         4         0         0
## 5         0         0         0         0         1         0         0         0         2         0         0
## 6         0         0         0         1         1         0         0         0         1         0         0
##      lelbow lwrist butt lleg lknee lfoot ddis ddiscomfort      oftendis
## 1         0         0         0         0         0         0      NA      #VALUE! Occasionally
## 2         0         0         0         0         0         0      NA      #VALUE! When Active
## 3         0         3         0         0         0         0    730      730 When Active
## 4         0         1         0         0         2         1   365      365 Occasionally
## 5         0         0         0         0         0         0      NA      #VALUE! Occasionally
## 6         1         0         0         0         0         0   182      182 Occasionally
##      discomorig      workplace1      Job1      tool
## 1 Gradually over time      Seating No Concern No Concern
## 2 Gradually over time      Reaches No Concern No Concern
## 3 Gradually over time Adjustability; Seating      Breaks
## 4 Gradually over time      No Concern No Concern No Concern
## 5 Gradually over time      No Concern      Methods No Concern
## 6 Gradually over time      No Concern      Breaks No Concern
##      handling      environment
## 1      Lifting      Temperature
## 2 Carrying; Lifting; Push/Pull      Temperature; Lighting; Ventilation
## 3      Lifting; Lift Assists      Ventilation; Lighting; Temperature
## 4      Carrying; Lifting      Temperature; Noise
```

## 5	No Concern						Lighting
## 6	No Concern Noise; Temperature; Lighting; Ventilation						
##	fatigue5min	fatigue2	fatigue4	fatigue6	fatigue8	fatigue10	fatigue12
## 1	5	3	3	3	3	3	4
## 2	0	1	2	5	5	NA	NA
## 3	1	2	3	3	4	0	0
## 4	4	2	0	2	2	NA	NA
## 5	0	0	0	3	1	2	4
## 6	0	0	0	0	0	0	0
##	Days_to_less_fatigue		injury		injurypart		
## 1	14.0		No				
## 2	30.0		Yes				
## 3	60.0		Yes		Neck; Hand/wrist; Lower Back; Foot		
## 4	1.5		Yes		Neck		
## 5	NA		No				
## 6	NA		No				
##	plabor		prevposition		weightgain		lbs_gained
## 1	Very little		No		No		NA
## 2	Moderate		No		No		NA
## 3	Moderate		No		Yes		18
## 4	Moderate		No		Yes		NA
## 5	Very little		No		Yes		6
## 6	Very little		No		No		NA
##	weight_situation		bweight		exbreak		expartic
## 1	I am AT my ideal weight		135		Yes		No
## 2	I am ABOVE my ideal weight		135		No		No
## 3	I am ABOVE my ideal weight		124		Yes		Yes
## 4	I am ABOVE my ideal weight		NA		Yes		Yes
## 5	I am ABOVE my ideal weight		151		Yes		Yes
## 6	I am ABOVE my ideal weight		280		Yes		Yes
##	daysVHW		Veg_Serv		fruit_servings		fried_servings
## 1	3.5		NA		NA		NA
## 2	0.0		2		0		0
## 3	0.0		3		5		2
## 4	NA		2		2		0
## 5	0.0		2		3		2
## 6	0.0		0		0		0
##	vig7		modur		moderatet		modtur
## 1	NA		0		NA		0
## 2	NA		0		NA		3
## 3	NA		4		NA		5
## 4	8		7		NA		8
## 5	NA		0		NA		0
## 6	NA		3		NA		60
##	sitt7t		situr		Field66		BMI
## 1	NA		8		1		21.79
## 2	300		NA		2		22.46
## 3	600		NA		3		22.68
## 4	240		NA		4		NA
## 5	540		NA		5		25.12
## 6	600		NA		6		35.95
##	Total_ex_time		Total_Ex_Time_Exclude				
## 1	330		0		0		
## 2	90		0		180		
##	Vig.ex.Time		Mod.ex.time		Walk.ex.Time		Total_Met_Min
## 1	330		0		0		2640
## 2	90		0		180		1314


```
## 3      180      160      100      2410      NA
## 4       24       56       24       320      NA
## 5       40        0        0       320      NA
## 6      360      180      630      5679
```

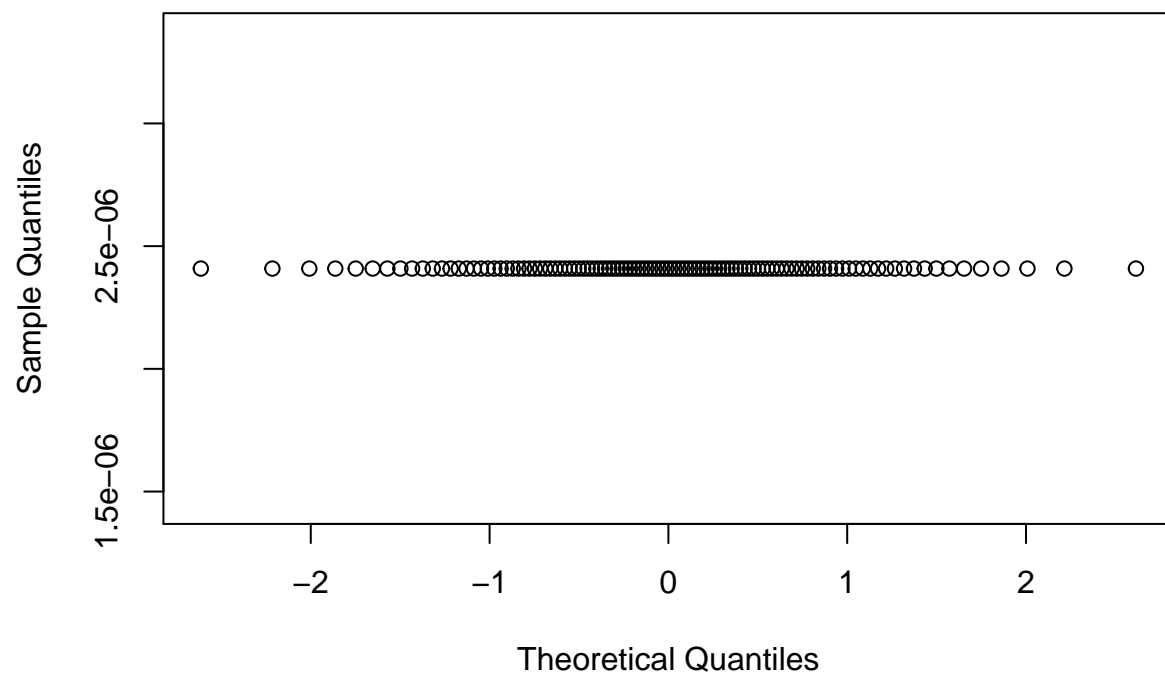
```
dat1$gender <- factor(dat1$gender)
dat1$shift <- factor(dat1$shift)
```

```
mod1 <- glm(weightgain.b ~ shift + gender + Age + height + Beg_BMI + Total_Met_Min, family = binomial, na.rm = TRUE)
```

```
## Warning: glm.fit: algorithm did not converge
```

```
##qq plot
qqnorm(residuals(mod1))
```

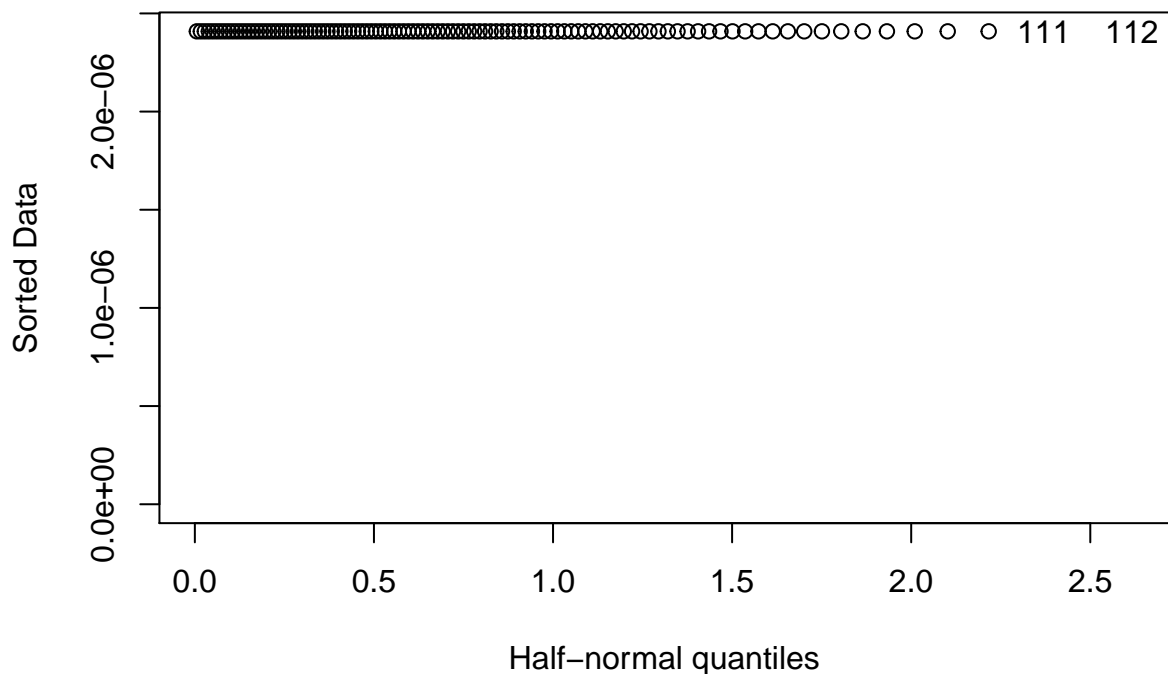
Normal Q-Q Plot



```
##Outliers: no obvious outliers
library(faraway)
```

```
## Warning: package 'faraway' was built under R version 3.6.3
```

```
halfnorm(residuals(mod1))
```



```
## test if shift should be removed from the model
```

```
mod0 <- glm(weightgain.b ~ gender + Age + height + Beg_BMI + Total_Met_Min, family = binomial, na.omit(d
```

```
## Warning: glm.fit: algorithm did not converge
```

```
summary(mod1)
```

```
##
```

```
## Call:
```

```
## glm(formula = weightgain.b ~ shift + gender + Age + height +
```

```
##     Beg_BMI + Total_Met_Min, family = binomial, data = na.omit(dat1))
```

```
##
```

```
## Deviance Residuals:
```

```
##      Min       1Q   Median       3Q      Max
```

```
## 2.409e-06 2.409e-06 2.409e-06 2.409e-06 2.409e-06
```

```
##
```

```
## Coefficients:
```

```
##           Estimate Std. Error z value Pr(>|z|)
```

```
## (Intercept)  2.657e+01 9.919e+05      0      1
```

```
## shift10am    -4.695e-06 2.728e+05      0      1
```

```
## shift11am    -2.580e-06 2.789e+05      0      1
```

```
## shift12pm     1.584e-07 2.974e+05      0      1
```

```
## shift1pm     -8.091e-07 3.662e+05      0      1
```

```
## shift2pm     -2.699e-06 3.245e+05      0      1
```

```
## shift7am     -5.463e-06 2.889e+05      0      1
```

```
## shift8am      -1.577e-06  2.691e+05      0      1
## shift9am      -1.728e-06  2.820e+05      0      1
## shifttother   -5.322e-07  3.117e+05      0      1
## genderFemale  -9.812e-08  3.916e+05      0      1
## genderMale     9.053e-07  4.036e+05      0      1
## Age           -1.096e-06  1.174e+05      0      1
## height        -1.187e-07  1.186e+04      0      1
## Beg_BMI       -1.285e-07  6.968e+03      0      1
## Total_Met_Min -1.341e-10  7.348e+02      0      1
##
## (Dispersion parameter for binomial family taken to be 1)
##
## Null deviance: 0.0000e+00 on 111 degrees of freedom
## Residual deviance: 6.4978e-10 on 96 degrees of freedom
## AIC: 32
##
## Number of Fisher Scoring iterations: 25
```

```
summary(mod0)
```

```
##
## Call:
## glm(formula = weightgain.b ~ gender + Age + height + Beg_BMI +
## Total_Met_Min, family = binomial, data = na.omit(dat1))
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## 2.409e-06  2.409e-06  2.409e-06  2.409e-06  2.409e-06
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)   2.657e+01  9.572e+05      0      1
## genderFemale  -1.074e-08  3.593e+05      0      1
## genderMale    -8.714e-09  3.693e+05      0      1
## Age           3.807e-09  1.138e+05      0      1
## height       -3.932e-11  1.167e+04      0      1
## Beg_BMI       1.322e-10  6.518e+03      0      1
## Total_Met_Min -2.871e-11  6.991e+02      0      1
##
## (Dispersion parameter for binomial family taken to be 1)
##
## Null deviance: 0.0000e+00 on 111 degrees of freedom
## Residual deviance: 6.4978e-10 on 105 degrees of freedom
## AIC: 14
##
## Number of Fisher Scoring iterations: 25
```

```
anova(mod0, mod1, test = "Chi")
```

```
## Analysis of Deviance Table
##
## Model 1: weightgain.b ~ gender + Age + height + Beg_BMI + Total_Met_Min
## Model 2: weightgain.b ~ shift + gender + Age + height + Beg_BMI + Total_Met_Min
```

```
##   Resid. Df Resid. Dev Df Deviance Pr(>Chi)
## 1      105 6.4978e-10
## 2       96 6.4978e-10  9         0      1
```

##do not reject the null hypothesis that the reduced model is better(the model includes shift is less d

Solely from the model, we can see that only three variables have significant impact on the predictor: gender, Beg_BMI and Total_Met_Min. We perform a Chi-square test to test if the model contains shift is better than the model without shifts. The null hypothesis is that the model without shift is a better model, and the alternative hypothesis is that the model without shift is not a better model. In order to compare two models, we omit the NA values in the dataset to make sure the number of cases used in each model is the same. The test statistic is very large given a 95% confidence interval, therefore we fail to reject the null hypothesis that the reduced model is better, which means that the model contains shift is less favorable than the reduced model. The test result is shown below:

Analysis of Deviance Table

```
Model 1: weightgain.b ~ gender + Age + height + Beg_BMI + Total_Met_Min
Model 2: weightgain.b ~ shift + gender + Age + height + Beg_BMI + Total_Met_Min
Resid. Df Resid. Dev Df Deviance Pr(>Chi)
1 202 236.06
2 194 228.44 8 7.6207 0.4714
```

Since the initial model shows that Total_Met_Min is a significant variable, we performed another analysis, using a forward stepwise selection to select the model with the most appropriate variables that produces the lowest AIC. The selected variables are exactly the same as our previous analysis, which are gender, Beg_BMI and Total_Met_Min, which reassures that total metabolic minutes do have an effect on weight gain and shift does not have an effect on weightgain. The results are shown below:

```
Call: glm(formula = weightgain.b ~ gender + Beg_BMI + Total_Met_Min, family = binomial, data =
na.omit(dat1))
```

```
Coefficients: (Intercept) genderMale Beg_BMI Total_Met_Min
3.0281421 -0.7079166 -0.0615582 -0.0001601
```

```
Degrees of Freedom: 207 Total (i.e. Null); 204 Residual Null Deviance: 251.7 Residual Deviance: 237.4 AIC:
245.4
```

Using stepwise to select the best model

```
mod00 <- glm(weightgain.b ~ 1, family = binomial, na.omit(dat1))
```

```
## Warning: glm.fit: algorithm did not converge
```

```
m.sw1 = step(mod00,
             scope=list(lower=mod00, upper=mod1),
             direction = "both", trace = F)
```

```
## Warning: glm.fit: algorithm did not converge
```

```
## Warning: glm.fit: algorithm did not converge
```

```
## Warning: glm.fit: algorithm did not converge
```

```
## Warning: glm.fit: algorithm did not converge
```

```
## Warning: glm.fit: algorithm did not converge
```

```
## Warning: glm.fit: algorithm did not converge
```

```
## Warning: glm.fit: algorithm did not converge
```

```
m.sw1$anova
```

```
##   Step Df Deviance Resid. Df   Resid. Dev AIC
## 1      NA      NA      111 6.497771e-10   2
```

```
m.sw1
```

```
##
## Call:  glm(formula = weightgain.b ~ 1, family = binomial, data = na.omit(dat1))
##
## Coefficients:
## (Intercept)
##      26.57
##
## Degrees of Freedom: 111 Total (i.e. Null);  111 Residual
## Null Deviance:      0
## Residual Deviance: 6.498e-10      AIC: 2
```

```
## The selected model with the lowest AIC contains only three variables: gender, BMI and Total_Met_Min,
```

The Relationship Between Total Metabolic Minutes and Weight Gain

Using various methods of analysis, we found that total metabolic minutes is a significant variable. The initial model demonstrated that there is a negative relationship between total metabolic minutes and weight gain, and this was corroborated by the subsequent forward stepwise model.

The Relationship Between Shift Time and Weight Gain

It does not appear that shift has significant impact on weight gain. From the Chi-square test, we found that the model without shift performs better than the model with shift. As mentioned, we failed to reject the null hypothesis that the reduced model is better, so we do not find it beneficial to include shift in our final model to predict weight gain.

Conclusion

Our initial goal was to see whether shift and total metabolic minutes have an impact on weight gain. Multiple forms of analysis indicate that shift is not a relevant factor in determining weight gain, while total metabolic minutes is. Total metabolic minutes has a slightly adverse relationship to weight gain. We conducted further analysis to go beyond the requested covariate relationships. For a final model selection, gender, Beg_BMI, and Total_Met_Min appear to be the most significant and useful predictors of weight gain.