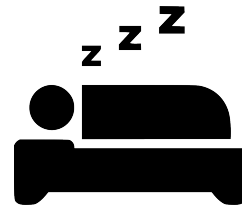
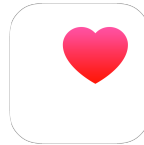
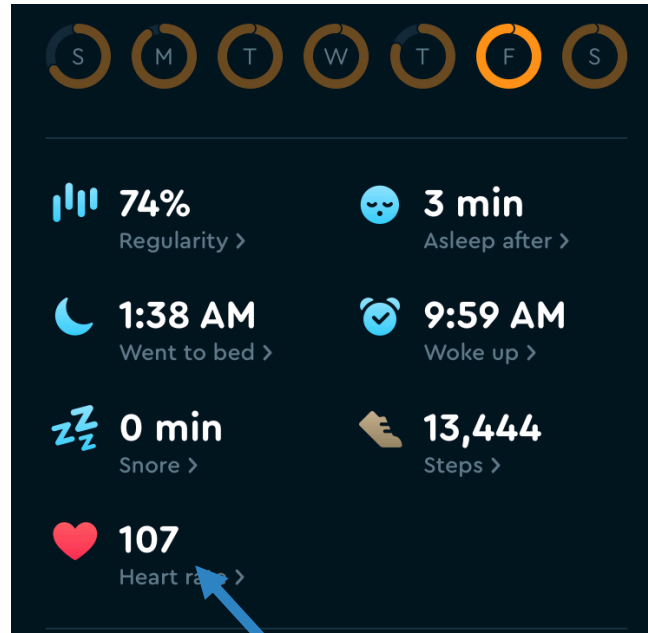




Analyzing Personal Health Data

Robert Swanson





The Question:

What factors influence sleeping heart rate?

Available Data

(Collected from my iPhone and Apple Watch)

- ▶ Sleep Analysis (minutes in bed, asleep)
- ▶ Heart Rate
- ▶ Active Calories
- ▶ Resting Calories
- ▶ Cycling Distance
- ▶ Distance
- ▶ Steps
- ▶ Flights of Stairs Climbed

Step 1: Clean Up and Manipulate

Used readr, dplyr, lubridate, matrixStats libraries

	Start	Finish	Heart Rate (count/min)
1	30-Nov-2018 13:13	30-Nov-2018 13:13	68.00000
2	30-Nov-2018 13:19	30-Nov-2018 13:19	75.00000
3	30-Nov-2018 13:23	30-Nov-2018 13:23	61.62732
4	30-Nov-2018 13:25	30-Nov-2018 13:25	86.00000
5	30-Nov-2018 13:29	30-Nov-2018 13:29	80.00000
6	30-Nov-2018 13:30	30-Nov-2018 13:30	78.00000
7	30-Nov-2018 13:34	30-Nov-2018 13:34	66.00000
8	30-Nov-2018 13:34	30-Nov-2018 13:34	67.00000
9	30-Nov-2018 13:34	30-Nov-2018 13:34	67.00000
10	30-Nov-2018 13:34	30-Nov-2018 13:34	68.00000

Raw Data

Joined to full minute table

	end	bpm
1	2018-11-30 13:13:00	68.00000
2	2018-11-30 13:14:00	68.00000
3	2018-11-30 13:15:00	68.00000
4	2018-11-30 13:16:00	68.00000
5	2018-11-30 13:17:00	68.00000
6	2018-11-30 13:18:00	68.00000
7	2018-11-30 13:19:00	75.00000
8	2018-11-30 13:20:00	75.00000
9	2018-11-30 13:21:00	75.00000
10	2018-11-30 13:22:00	75.00000

*Any Data Points fulfilling over 10 minutes are invalidated (NA)

Filled Down

Averaged

	date	bpm
1	2018-11-30	84.06578
2	2018-12-01	70.59088
3	2018-12-02	73.34748
4	2018-12-03	77.16180
5	2018-12-04	70.84058
6	2018-12-07	76.16701
7	2018-12-08	71.77801
8	2018-12-09	63.97538
9	2018-12-10	65.06643
10	2018-12-11	62.40557

Grouped by date

- Mean
- Median
- Max
- Min

	time	bpm
1	12:00 AM	58.70780
2	12:01 AM	58.78386
3	12:02 AM	58.98739
4	12:03 AM	58.66390
5	12:04 AM	58.44183
6	12:05 AM	58.51704
7	12:06 AM	58.51538
8	12:07 AM	58.64447
9	12:08 AM	58.69603
10	12:09 AM	58.82933

Grouped by time

`pivot_wider(names_from = weekday, values_from = bpm)`

	weekday	bpm
1	Sunday	66.47418
2	Monday	68.84909
3	Tuesday	67.73481
4	Wednesday	68.68466
5	Thursday	70.28824
6	Friday	69.09375
7	Saturday	67.74566

Grouped by weekday

	time	Sunday	Monday	Tuesday	Wednesday	Thursday	Friday	Saturday
1	12:00 AM	55.49152	56.86026	60.47404	61.78257	57.75000	60.65625	57.92175
2	12:01 AM	55.13658	56.76796	60.84848	61.94130	58.26154	60.73438	57.81784
3	12:02 AM	55.60415	57.79873	61.00439	62.00000	58.30769	60.58920	57.55362
4	12:03 AM	55.34341	56.90345	61.42863	61.61290	58.15385	59.71420	57.34265
5	12:04 AM	55.76190	56.41270	61.18620	61.25806	57.46875	59.79688	57.05000

Grouped by weekday and time

	In bed start	In bed Finish	Minutes in bed	Minutes asleep
1011	25-Oct-2019 00:30	25-Oct-2019 06:30	360.5173921	0.00000
1012	28-Oct-2019 00:40	28-Oct-2019 07:25	405.5136193	0.00000
1013	29-Oct-2019 03:15	29-Oct-2019 07:20	245.5044023	0.00000
1014	29-Oct-2019 22:51	30-Oct-2019 07:20	508.5393671	491.58805
1015	30-Oct-2019 23:36	31-Oct-2019 07:42	486.0953667	418.04202
1016	01-Nov-2019 02:36	01-Nov-2019 06:45	248.8458177	163.39927
1017	02-Nov-2019 00:47	02-Nov-2019 09:00	493.0725299	0.00000
1018	03-Nov-2019 01:25	03-Nov-2019 08:25	480.0013566	443.37900
1019	03-Nov-2019 23:01	04-Nov-2019 07:00	478.9337177	0.00000

Raw Data

	time	bed
1	2016-09-13 20:59:00	FALSE
2	2016-09-13 21:21:00	FALSE
3	2016-09-13 21:27:00	FALSE
4	2016-09-13 21:30:00	FALSE
5	2016-09-13 21:33:00	FALSE
6	2016-09-13 21:38:00	FALSE
7	2016-09-13 21:41:00	FALSE
8	2016-09-14 19:04:00	FALSE
9	2016-09-15 07:00:00	FALSE
10	2016-09-16 07:00:00	FALSE

End Times

	time	bed
1	2016-09-13 18:48:00	TRUE
2	2016-09-13 21:04:00	TRUE
3	2016-09-13 21:23:00	TRUE
4	2016-09-13 21:29:00	TRUE
5	2016-09-13 21:32:00	TRUE
6	2016-09-13 21:35:00	TRUE
7	2016-09-13 21:39:00	TRUE
8	2016-09-14 18:15:00	TRUE
9	2016-09-14 23:20:00	TRUE
10	2016-09-15 23:30:00	TRUE

Start Times

	time	bed
129	2016-09-13 20:56:00	TRUE
130	2016-09-13 20:57:00	TRUE
131	2016-09-13 20:58:00	TRUE
132	2016-09-13 20:59:00	FALSE
133	2016-09-13 21:00:00	FALSE
134	2016-09-13 21:01:00	FALSE
135	2016-09-13 21:02:00	FALSE
136	2016-09-13 21:03:00	FALSE
137	2016-09-13 21:04:00	TRUE

Datetime is in bed

Inner Join

	end	bpm
458	2018-12-12 05:55:00	53.00000
459	2018-12-12 05:56:00	52.00000
460	2018-12-12 05:57:00	52.00000
461	2018-12-12 05:58:00	52.00000
462	2018-12-12 05:59:00	52.00000
463	2018-12-12 06:00:00	52.00000
464	2018-12-12 06:01:00	52.00000
465	2018-12-12 06:02:00	52.00000
466	2018-12-14 23:27:00	54.00000
467	2018-12-14 23:28:00	54.00000
468	2018-12-14 23:29:00	57.00000

Bed Heart Rate

Filled Down

	end	bpm
1	2018-11-30 13:13:00	68.00000
2	2018-11-30 13:14:00	68.00000
3	2018-11-30 13:15:00	68.00000
4	2018-11-30 13:16:00	68.00000
5	2018-11-30 13:17:00	68.00000
6	2018-11-30 13:18:00	68.00000
7	2018-11-30 13:19:00	75.00000
8	2018-11-30 13:20:00	75.00000
9	2018-11-30 13:21:00	75.00000
10	2018-11-30 13:22:00	75.00000

Heart Rate Grouped by datetime

Bed Heart Rate Grouped by:

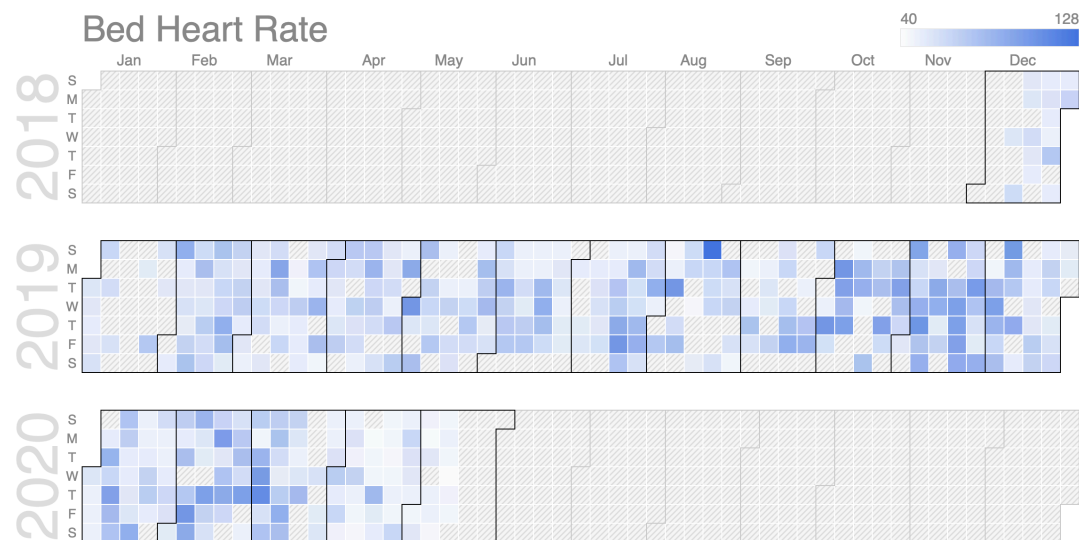
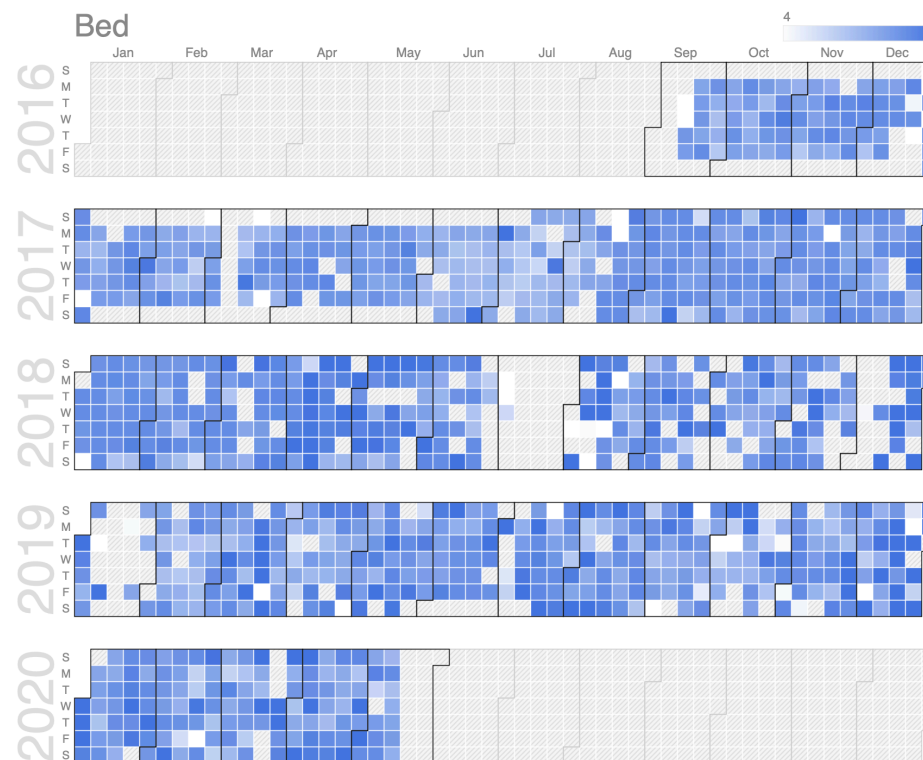
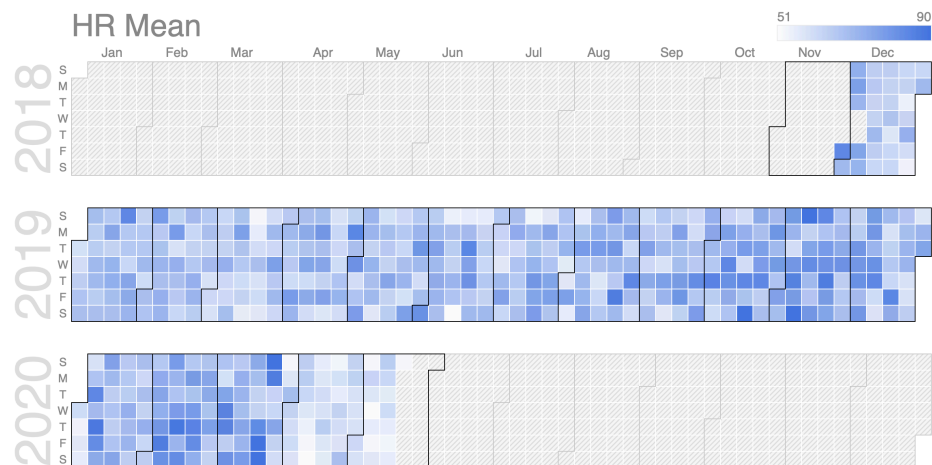
- Datetime
- Date
- Weekday

Step 2: Visualize and Understand

Used ggplot and googleVis libraries

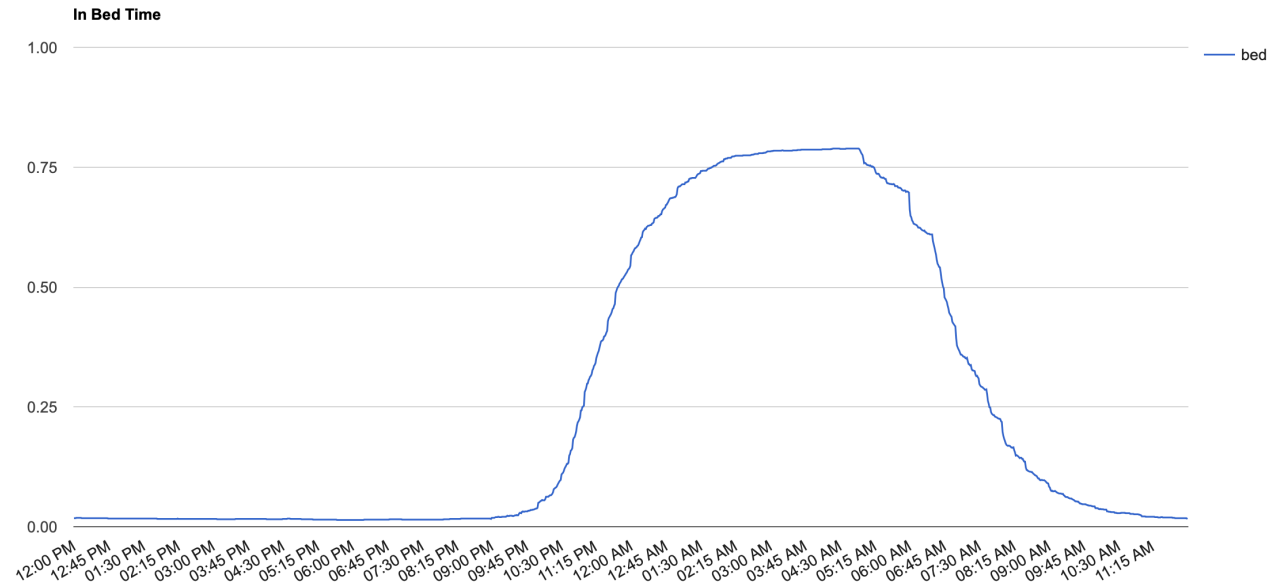
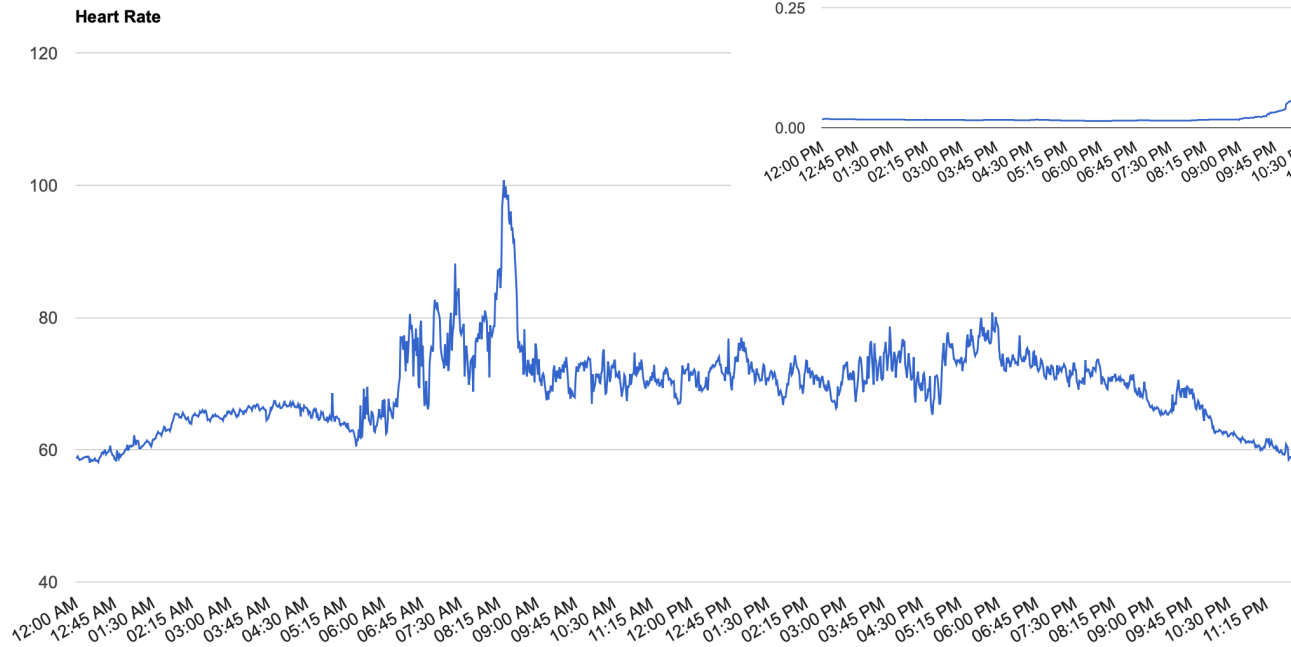
Year Plots

(googleVis)



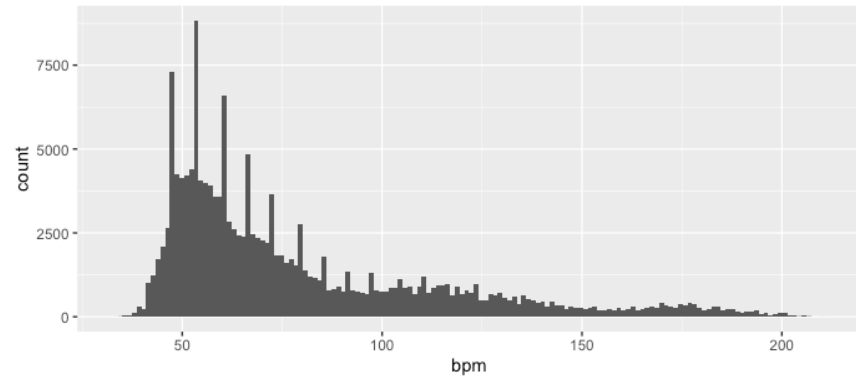
Line Plots

(googleVis)

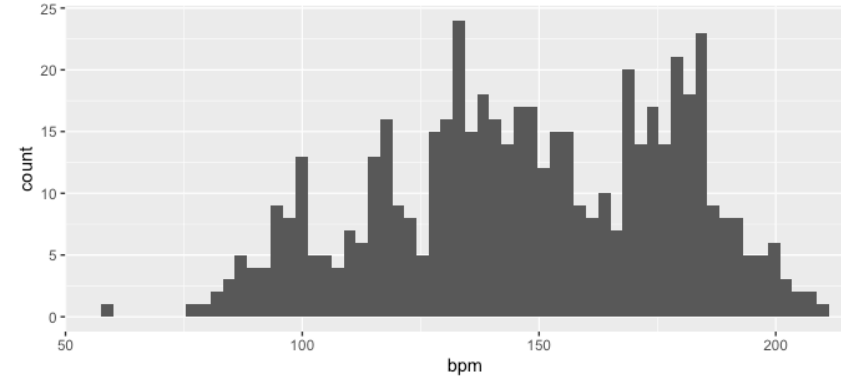


Distribution Histograms

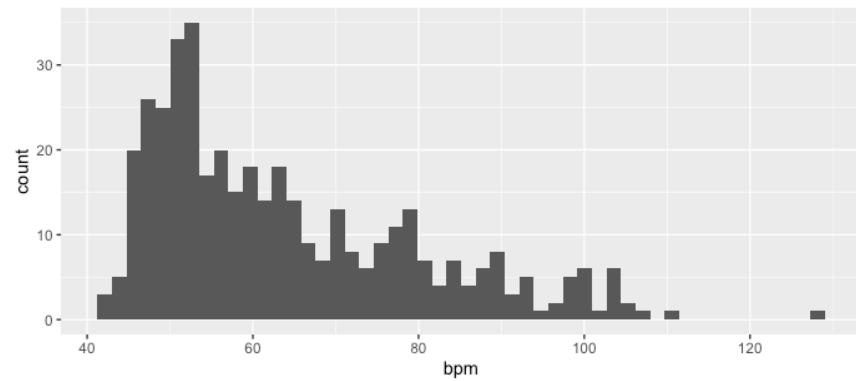
Heart Rate Distribution



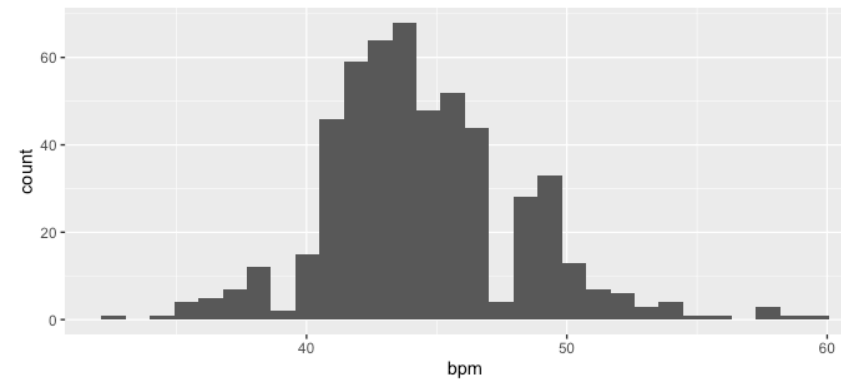
Maximum Heart Rate Distribution



Bed Heart Rate

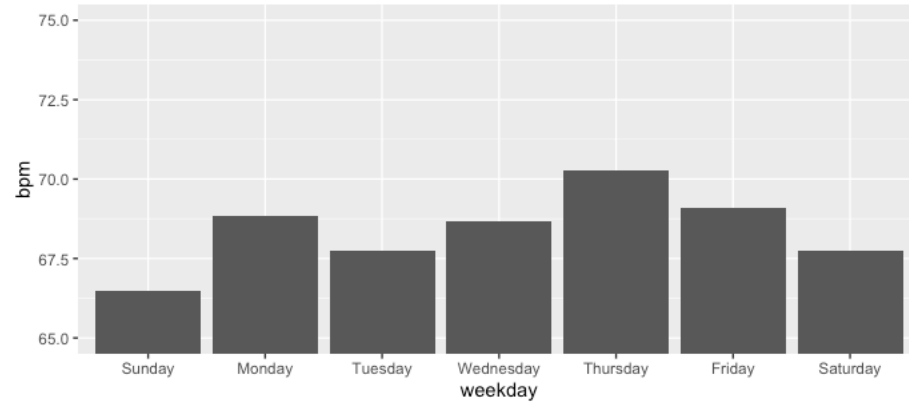


Minimum Heart Rate Distribution

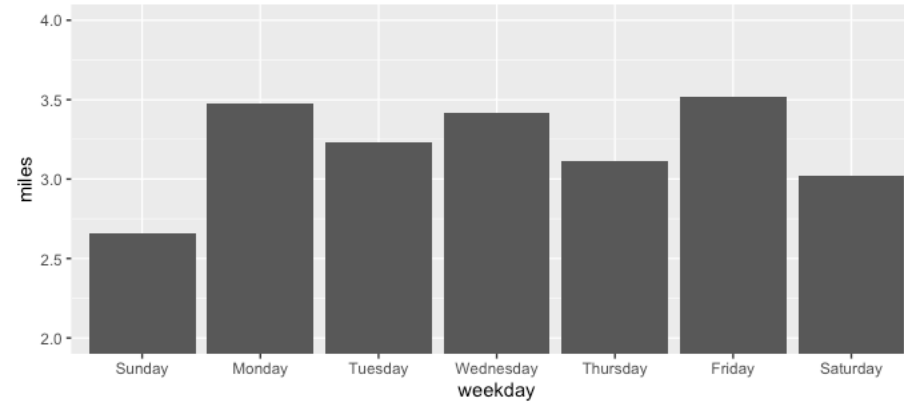


Weekday Charts

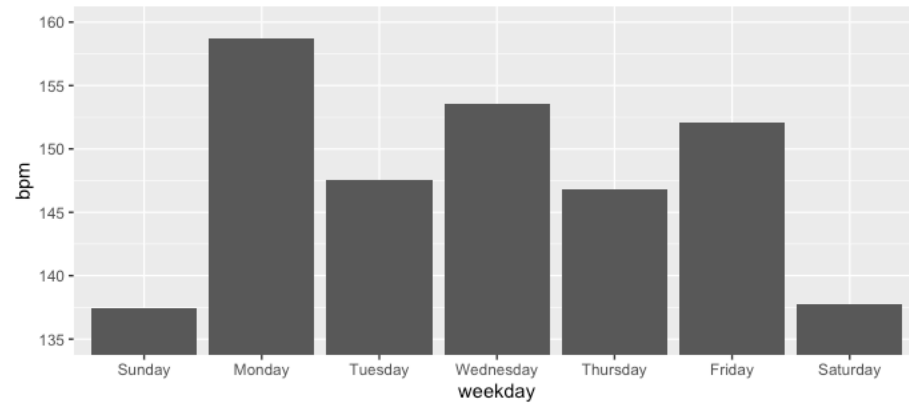
Mean Heart Rate Average Per Weekday



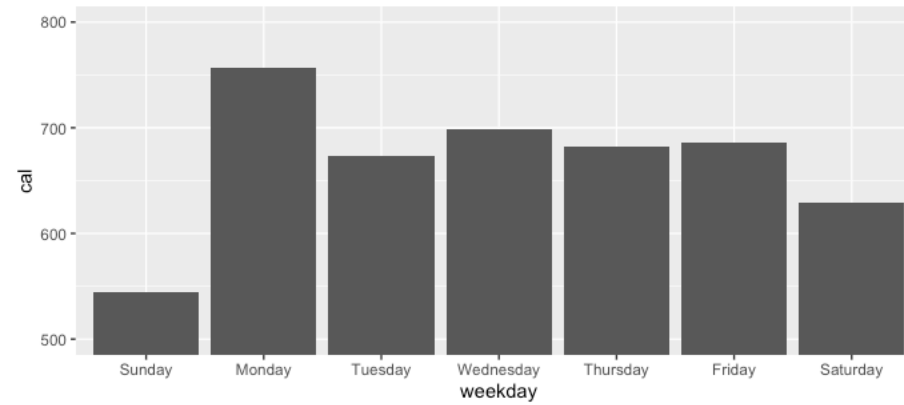
Distance Average Per Weekday



Maximum Heart Rate Average Per Weekday



Active Calories Average Per Weekday



Step 3: Correlate and Analyze

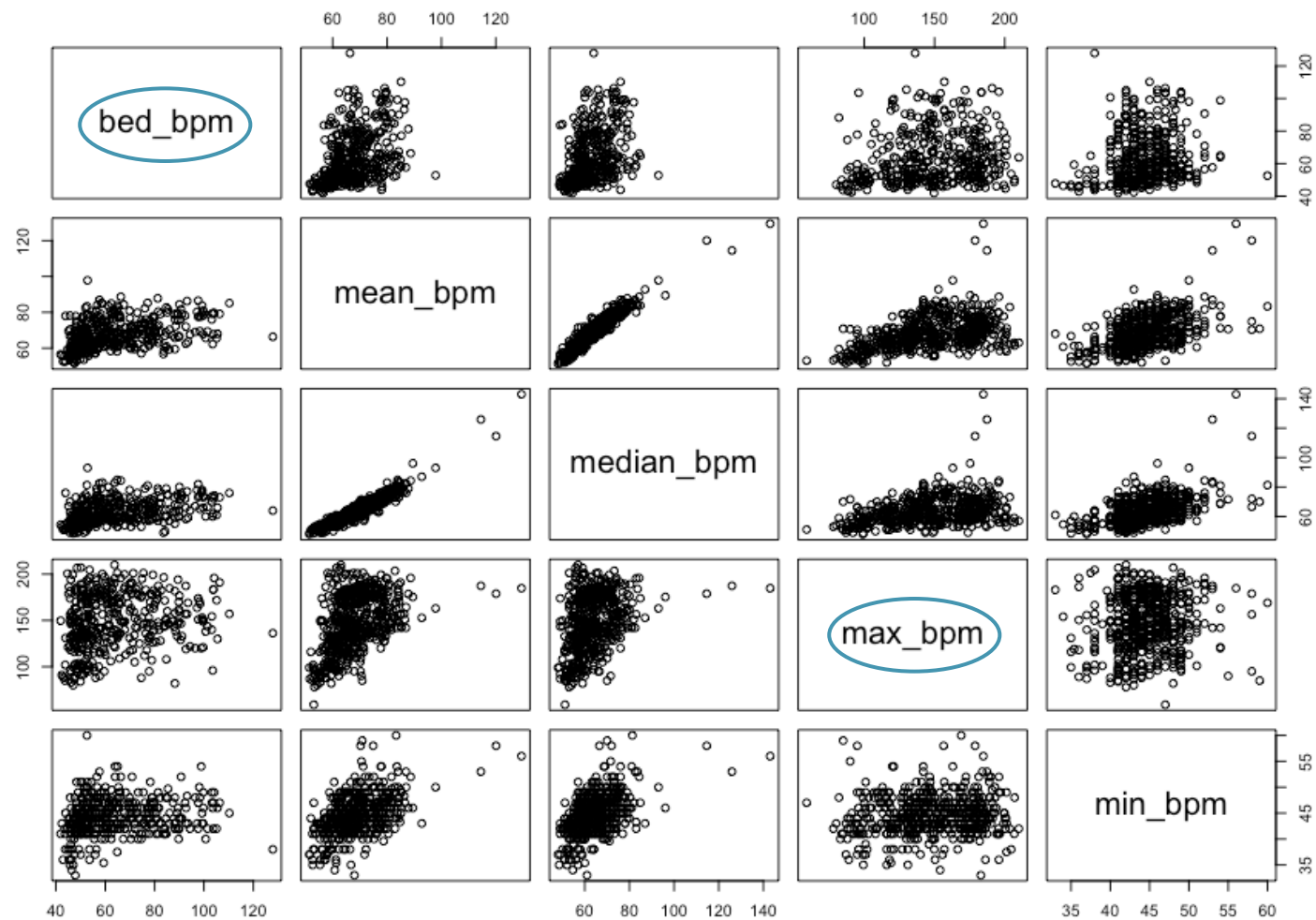
Used zoo and rpart libraries

Unify into single Calendar

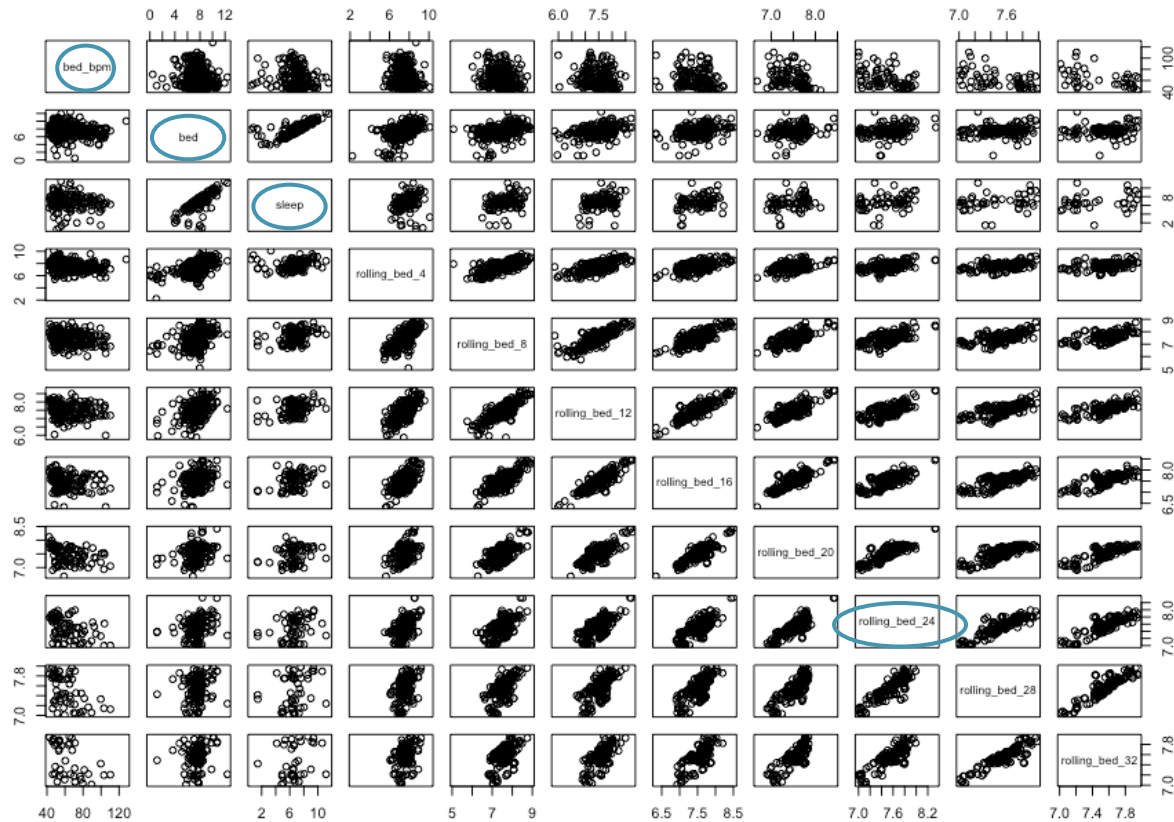
zoo:rollmean()

	date	mean_bpm	median_bpm	max_bpm	min_bpm	bed_bpm	bed	sleep	active_cal	resting_cal	cycling_miles	miles	steps	flights	rolling_bed_4	rolling
1414	2019-11-17	83.50482	74.00000	149.0000	44.00000	NA	6.05	5.52	436.000	1702.855	NA	4.8358216	9912	23	NA	NA
1415	2019-11-18	66.57602	65.00000	147.0000	51.00000	78.25101	7.02	6.74	510.432	1677.948	NA	8.6370404	17841	35	NA	NA
1416	2019-11-19	78.26798	73.00000	150.4513	46.00000	99.76332	8.11	7.74	1008.223	1697.043	7.063051	8.6192282	17377	53	6.9225	NA
1417	2019-11-20	81.00370	74.00000	148.6101	49.00000	99.71138	8.00	7.66	893.806	1695.547	NA	8.2454858	16438	26	7.2950	NA
1418	2019-11-21	86.52842	81.92308	151.0000	47.00000	97.65913	6.19	5.99	899.731	1695.152	NA	11.1345335	22490	46	7.3300	NA
1419	2019-11-22	78.98526	74.00000	133.0000	44.00000	91.09111	8.35	8.07	677.018	1714.166	NA	11.3265267	22810	45	7.6625	NA
1420	2019-11-23	76.16838	66.00000	137.0000	46.00000	62.98199	7.69	7.17	473.207	1688.101	NA	6.2234268	12445	28	7.5575	7.240
1421	2019-11-24	67.34932	59.00000	179.0000	47.00000	62.68566	7.10	6.78	715.480	1726.373	NA	11.7749700	23728	62	7.3325	7.313
1422	2019-11-25	69.36051	66.50000	143.4664	42.00000	101.44025	6.28	6.07	929.731	1700.686	NA	9.9338595	19885	31	7.3550	7.342
1423	2019-11-26	73.96910	65.00000	136.0000	41.00000	76.43431	11.02	9.88	520.200	1717.178	NA	5.6357784	11472	30	8.0225	7.842
1424	2019-11-27	81.05539	79.00000	161.0411	45.00000	56.44060	9.13	8.43	766.837	1698.013	NA	3.7747173	6888	2	8.3825	7.970
1425	2019-11-28	64.00749	57.00000	136.0000	44.01146	52.17128	10.02	9.92	336.928	1673.564	NA	2.1514904	3812	NA	9.1125	8.222
1426	2019-11-29	65.73612	55.08333	158.0000	43.00000	88.23271	8.04	7.45	591.440	1695.791	NA	3.0621577	5629	NA	9.5525	8.453
1427	2019-11-30	68.25555	57.00000	137.0000	43.00000	NA	8.24	6.21	225.530	1632.868	NA	0.6602294	1347	4	8.8575	8.440
1428	2019-12-01	63.19715	60.00000	148.0000	46.00000	46.11420	6.70	6.12	521.754	1767.883	NA	7.0071629	14389	17	8.2500	8.316
1429	2019-12-02	64.69896	60.00000	140.0000	43.00000	81.85713	7.42	6.53	887.696	1700.595	NA	9.0384020	18316	32	7.6000	8.356
1430	2019-12-03	79.61986	78.00000	153.0000	43.00000	99.24706	5.66	5.39	628.124	1698.132	NA	9.2708713	19248	40	7.0050	8.278
1431	2019-12-04	79.93485	75.00000	159.8836	48.00000	85.14748	7.25	6.89	1063.454	1706.949	NA	9.8709562	19389	42	6.7575	7.807
1432	2019-12-05	81.65213	77.00000	140.0000	47.00000	63.43438	1.97	NA	1018.733	1723.212	NA	10.7429430	21313	35	5.5750	6.912
1433	2019-12-06	69.38030	65.72727	145.0000	47.00000	79.42747	9.79	5.41	681.911	1712.359	NA	10.1845135	20504	31	6.1675	6.883
1434	2019-12-07	79.25390	73.50000	154.3058	47.00000	103.19861	6.71	6.36	893.712	1683.194	11.324499	11.0542700	22540	39	6.4300	6.717
1435	2019-12-08	79.04778	73.00000	128.0000	48.00000	82.37521	8.01	7.72	639.757	1713.101	NA	10.5918275	21818	36	6.6200	6.688
1436	2019-12-09	79.81238	76.00000	142.0000	46.00000	72.97912	8.88	8.34	971.866	1702.176	NA	10.4730274	21036	20	8.3475	6.961
1437	2019-12-10	71.45858	66.00000	120.0000	42.00000	NA	6.75	6.48	635.161	1664.303	NA	10.5511241	21912	24	7.5875	6.877
1438	2019-12-11	81.62566	76.00000	146.6046	52.00000	90.72263	9.25	6.98	1192.684	1733.548	NA	11.8085247	23121	33	8.2225	7.326
1439	2019-12-12	84.19135	80.28571	189.0000	48.00000	NA	7.08	6.15	1015.358	1699.277	NA	11.7080434	22725	51	7.9900	7.305
1440	2019-12-13	60.39718	58.00000	103.0000	48.00000	50.68137	6.79	NA	250.954	1690.570	NA	2.3347490	4946	NA	7.4675	7.907

Heart Correlation

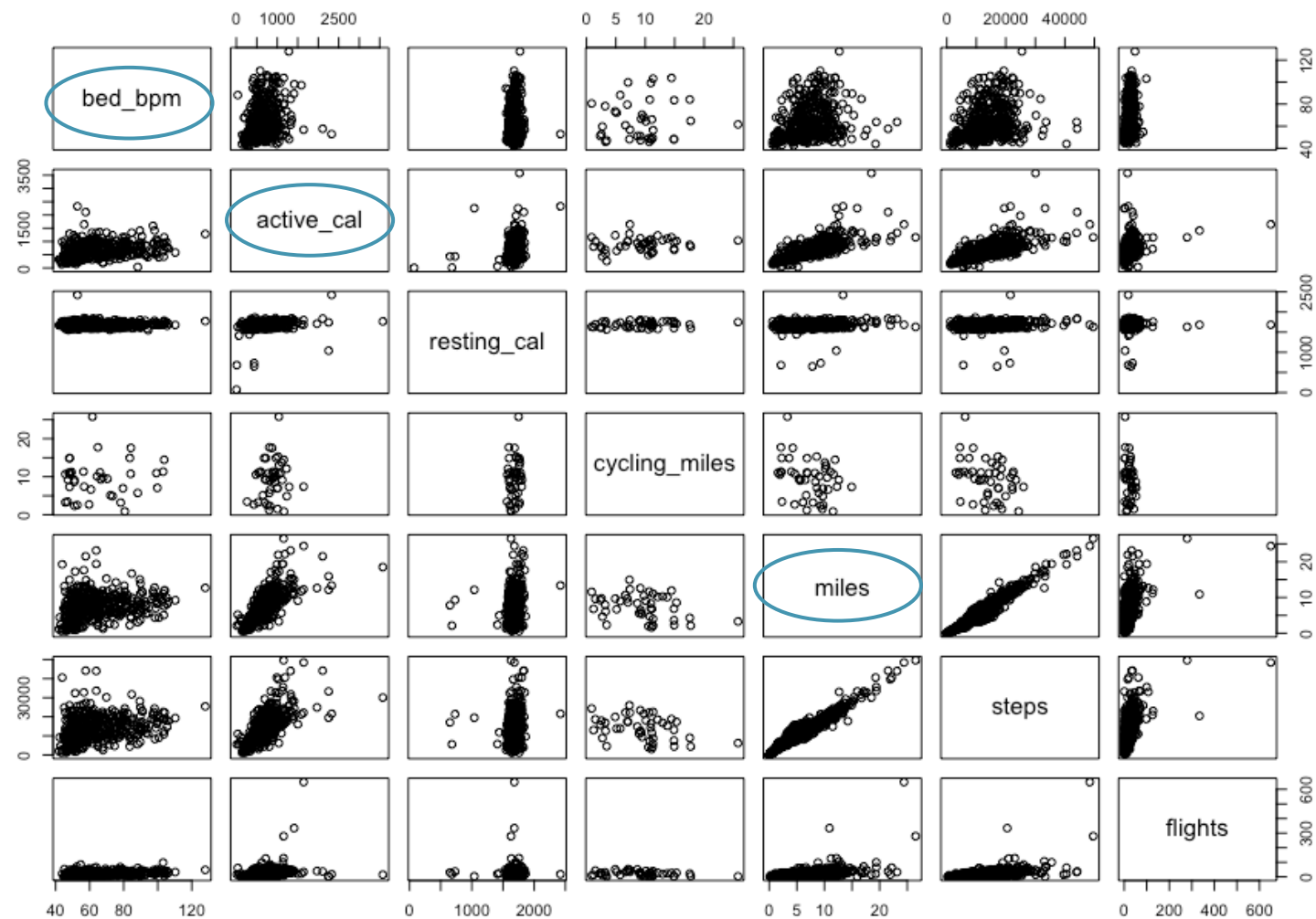


Sleep Correlation



```
> coef(lm(data = calendar, bed_bpm ~ max_bpm + bed + sleep + active_cal + miles + rolling_bed_4))[7]
rolling_bed_4
-5.5886
> coef(lm(data = calendar, bed_bpm ~ max_bpm + bed + sleep + active_cal + miles + rolling_bed_6))[7]
Error in eval(predvars, data, env) : object 'rolling_bed_6' not found
> coef(lm(data = calendar, bed_bpm ~ max_bpm + bed + sleep + active_cal + miles + rolling_bed_8))[7]
rolling_bed_8
-4.697724
> coef(lm(data = calendar, bed_bpm ~ max_bpm + bed + sleep + active_cal + miles + rolling_bed_10))[7]
Error in eval(predvars, data, env) : object 'rolling_bed_10' not found
> coef(lm(data = calendar, bed_bpm ~ max_bpm + bed + sleep + active_cal + miles + rolling_bed_12))[7]
rolling_bed_12
-4.543957
> coef(lm(data = calendar, bed_bpm ~ max_bpm + bed + sleep + active_cal + miles + rolling_bed_16))[7]
rolling_bed_16
-9.688853
> coef(lm(data = calendar, bed_bpm ~ max_bpm + bed + sleep + active_cal + miles + rolling_bed_20))[7]
rolling_bed_20
-20.91376
> coef(lm(data = calendar, bed_bpm ~ max_bpm + bed + sleep + active_cal + miles + rolling_bed_24))[7]
rolling_bed_24
-24.47009
> coef(lm(data = calendar, bed_bpm ~ max_bpm + bed + sleep + active_cal + miles + rolling_bed_28))[7]
rolling_bed_28
-23.65024
> coef(lm(data = calendar, bed_bpm ~ max_bpm + bed + sleep + active_cal + miles + rolling_bed_32))[7]
rolling_bed_32
-18.20744
```

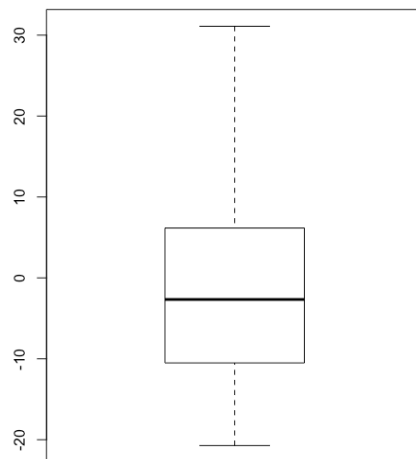
Activity Correlation



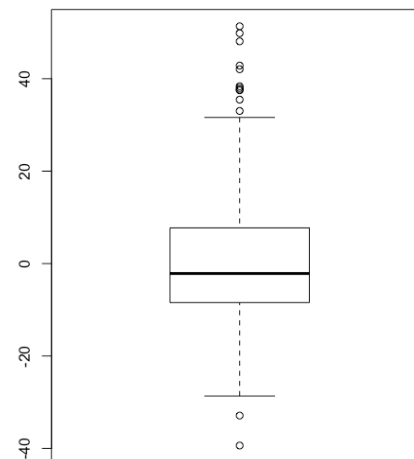
Build the models

```
> formula <- bed_bpm ~ max_bpm + bed + sleep + active_cal + miles + rolling_bed_24  
> lin_reg_model <- lm(data = calendar_train , formula)  
> desc_tree_model <- rpart(formula, calendar_train)  
> coef(lin_reg_model)
```

(Intercept)	max_bpm	bed	sleep	active_cal	miles	rolling_bed_24
188.671445426	0.062080412	9.705837409	-8.544298012	0.009507391	0.119297898	-20.795441859



Distribution of linear
regression model



Distribution of
decision tree model

