

EAS509 Homework 6 (100 points). Key

Submit your answers as a single pdf attach all R code. Failure to do so will result in grade reduction.

Question 1 (100 points)

High-Performance Computing (HPC) resources (a.k.a. supercomputers) are complex systems. Slight changes in hardware or software can drastically affect their performance. For example, a corrupted lookup table in a network switch, an update of a linux kernel, a drop of hardware support in a new software version, and so on.

One way to ensure the top performance of HPC resources is to utilize continuous performance monitoring where the same application is executed with the same input on a regular basis (for example, daily). In a perfect world, the execution time will be exactly the same, but in reality, it varies due to system jitter (this is partially due to system processes taking resources to do their jobs).

So normally, the execution time will be distributed around a certain value. If performance degradation occurs, the execution time will be distributed around different value.

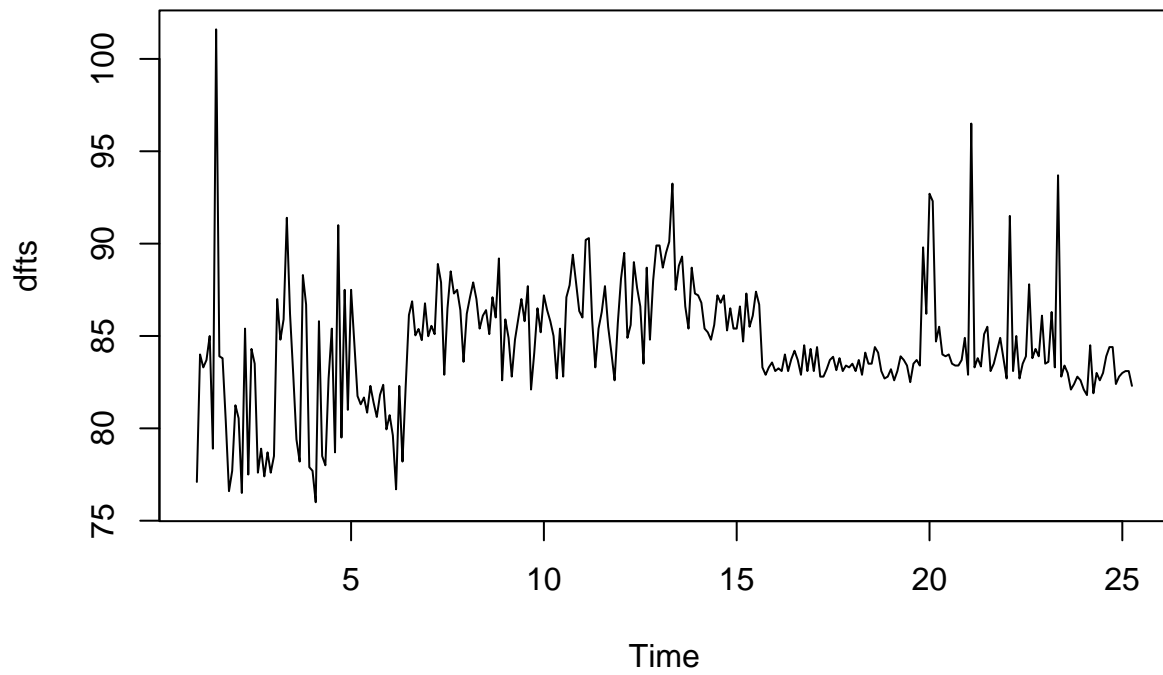
An automated system that inform system administrators on performance change can be a very handy tool.

In this exercise, your task will be to identify the number and location of the change point where performance was changed. NWChem, an Quantum Chemistry application, was used to probe the performance of UB HPC cluster.

1.1 UBHPC_8cores_NWChem_Wall_Clock_Time.csv file contains execution time (same as run time or wall time) of NWChem performing same reference calculation. Read the file and plot it run time on date. (10 points)

```
df <- read.csv('UBHPC_8cores_NWChem_Wall_Clock_Time.csv')
df$date <- as.POSIXct(df$date, format = "%d/%m/%Y %H:%M")
df$date <- as.Date(df$date)

dfts = ts(df$run_time, frequency = 12)
ts.plot(dfts)
```

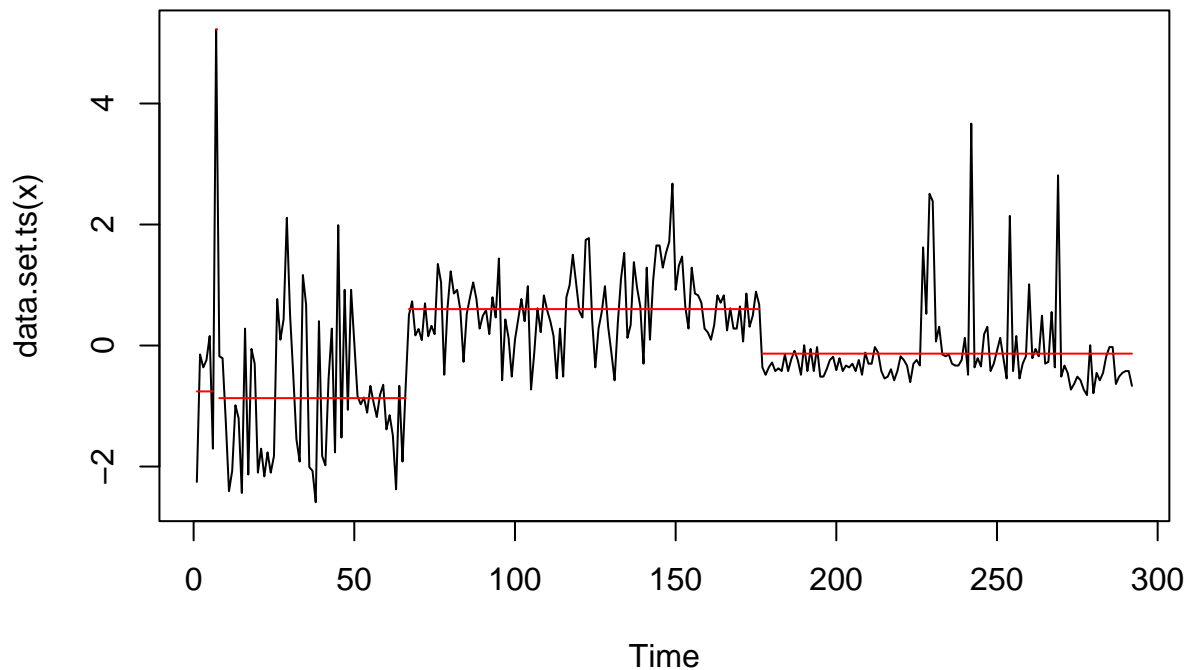


```
#mean changepoint analysis
```

```
mvalue = cpt.mean(as.vector(scale(dfts)), method='PELT')  
cpts(mvalue)
```

```
## [1] 6 7 66 176
```

```
plot(mvalue)
```



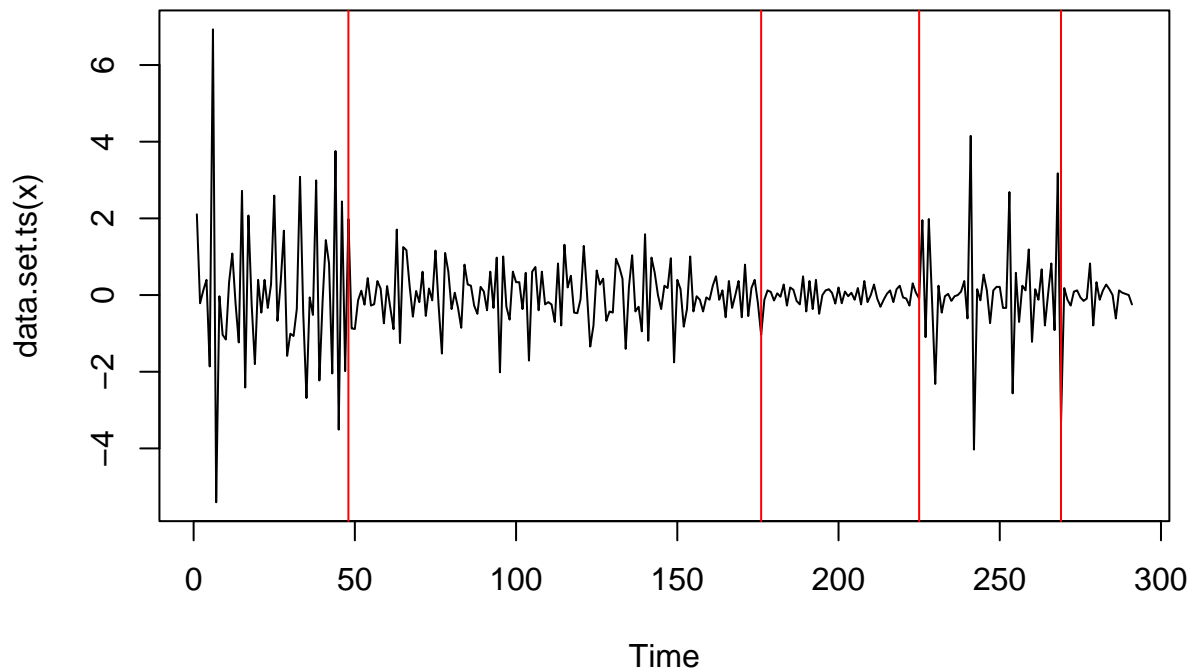
```
summary(mvalue)
```

```
## Created Using changepoint version 2.2.4
## Changepoint type      : Change in mean
## Method of analysis    : PELT
## Test Statistic       : Normal
## Type of penalty       : MBIC with value, 17.03026
## Minimum Segment Length : 1
## Maximum no. of cpts   : Inf
## Changepoint Locations : 6 7 66 176
```

```
# variance changepoint analysis
vnvalue = cpt.var(diff(as.vector(scale(dfts))), method='PELT')
cpts(vnvalue)
```

```
## [1] 48 176 225 269
```

```
plot(vnvalue)
```



```
summary(vnvalue)
```

```
## Created Using changepoint version 2.2.4
## Changepoint type      : Change in variance
## Method of analysis    : PELT
## Test Statistic       : Normal
## Type of penalty       : MBIC with value, 17.01997
## Minimum Segment Length : 2
## Maximum no. of cpts   : Inf
## Changepoint Locations : 48 176 225 269
```

1.2 How many segments/change points can you eyeball? What are they? (10 points)

Usinf cpt mean: 5 changepoints 6, 7, 66, 176 Usinf cpt var: 2 changepoints 48, 176, 225, 269

1.3 Create another column `seg` and assign segment number to it based on previous question. (10 points)

```
seg_data = df %>% mutate(segement = row_number()) %>% mutate(seg = ifelse(row_number() < 6, 1,
                                                                              ifelse(row_number() %in% 6:7, 2,
                                                                              ifelse(row_number() %in% 7:
                                                                              ifelse(row_number() %in%
                                                                              5))))))

seg_data
```

```
##           date run_time segement seg
```

## 1	2017-09-11	77.10000	1	1
## 2	2017-10-11	84.00000	2	1
## 3	2017-11-11	83.30000	3	1
## 4	2017-12-11	83.70000	4	1
## 5	<NA>	85.00000	5	1
## 6	<NA>	78.90000	6	2
## 7	<NA>	101.60000	7	2
## 8	<NA>	83.90000	8	3
## 9	<NA>	83.80000	9	3
## 10	<NA>	80.40000	10	3
## 11	<NA>	76.60000	11	3
## 12	<NA>	77.70000	12	3
## 13	<NA>	81.25000	13	3
## 14	<NA>	80.55000	14	3
## 15	<NA>	76.50000	15	3
## 16	<NA>	85.40000	16	3
## 17	<NA>	77.50000	17	3
## 18	<NA>	84.30000	18	3
## 19	2017-02-12	83.50000	19	3
## 20	2017-03-12	77.60000	20	3
## 21	2017-04-12	78.90000	21	3
## 22	2017-05-12	77.40000	22	3
## 23	2017-06-12	78.70000	23	3
## 24	2017-07-12	77.60000	24	3
## 25	2017-08-12	78.50000	25	3
## 26	2017-09-12	87.00000	26	3
## 27	2017-11-12	84.80000	27	3
## 28	2017-12-12	85.90000	28	3
## 29	<NA>	91.40000	29	3
## 30	<NA>	86.20000	30	3
## 31	<NA>	82.90000	31	3
## 32	<NA>	79.40000	32	3
## 33	<NA>	78.20000	33	3
## 34	<NA>	88.30000	34	3
## 35	<NA>	86.70000	35	3
## 36	<NA>	77.90000	36	3
## 37	<NA>	77.70000	37	3
## 38	<NA>	76.00000	38	3
## 39	<NA>	85.80000	39	3
## 40	<NA>	78.50000	40	3
## 41	<NA>	78.00000	41	3
## 42	<NA>	82.70000	42	3
## 43	<NA>	85.40000	43	3
## 44	<NA>	78.70000	44	3
## 45	<NA>	91.00000	45	3
## 46	2018-01-01	79.50000	46	3
## 47	2018-03-01	87.50000	47	3
## 48	2018-04-01	81.00000	48	3
## 49	2018-08-01	87.50000	49	3
## 50	<NA>	84.65000	50	3
## 51	<NA>	81.75000	51	3
## 52	<NA>	81.30000	52	3
## 53	<NA>	81.67000	53	3
## 54	<NA>	80.85000	54	3

## 55	2018-02-02	82.30000	55	3
## 56	2018-03-02	81.40000	56	3
## 57	2018-04-02	80.61667	57	3
## 58	2018-05-02	81.81111	58	3
## 59	2018-06-02	82.35714	59	3
## 60	2018-07-02	79.94286	60	3
## 61	2018-08-02	80.71429	61	3
## 62	2018-09-02	79.61667	62	3
## 63	2018-10-02	76.70000	63	3
## 64	2018-11-02	82.30000	64	3
## 65	<NA>	78.20000	65	3
## 66	<NA>	82.30000	66	3
## 67	<NA>	86.13333	67	4
## 68	<NA>	86.87778	68	4
## 69	<NA>	85.03333	69	4
## 70	<NA>	85.38000	70	4
## 71	<NA>	84.77500	71	4
## 72	<NA>	86.77000	72	4
## 73	<NA>	84.98750	73	4
## 74	<NA>	85.55000	74	4
## 75	<NA>	85.10000	75	4
## 76	<NA>	88.90000	76	4
## 77	<NA>	87.90000	77	4
## 78	<NA>	82.90000	78	4
## 79	2018-02-03	86.50000	79	4
## 80	2018-03-03	88.50000	80	4
## 81	2018-04-03	87.30000	81	4
## 82	2018-05-03	87.50000	82	4
## 83	2018-06-03	86.40000	83	4
## 84	2018-07-03	83.60000	84	4
## 85	2018-08-03	86.20000	85	4
## 86	2018-09-03	87.10000	86	4
## 87	2018-10-03	87.90000	87	4
## 88	2018-11-03	87.00000	88	4
## 89	2018-12-03	85.40000	89	4
## 90	<NA>	86.10000	90	4
## 91	<NA>	86.40000	91	4
## 92	<NA>	85.10000	92	4
## 93	<NA>	87.10000	93	4
## 94	<NA>	86.00000	94	4
## 95	<NA>	89.20000	95	4
## 96	<NA>	82.60000	96	4
## 97	<NA>	85.90000	97	4
## 98	<NA>	84.90000	98	4
## 99	<NA>	82.80000	99	4
## 100	<NA>	84.80000	100	4
## 101	<NA>	85.90000	101	4
## 102	<NA>	87.00000	102	4
## 103	<NA>	85.80000	103	4
## 104	<NA>	87.70000	104	4
## 105	<NA>	82.10000	105	4
## 106	<NA>	84.10000	106	4
## 107	<NA>	86.50000	107	4
## 108	2018-02-04	85.20000	108	4

## 109	2018-03-04	87.20000	109	4
## 110	2018-04-04	86.40000	110	4
## 111	2018-05-04	85.80000	111	4
## 112	2018-06-04	85.00000	112	4
## 113	2018-07-04	82.70000	113	4
## 114	2018-09-04	85.40000	114	4
## 115	2018-10-04	82.80000	115	4
## 116	2018-12-04	87.10000	116	4
## 117	<NA>	87.75000	117	4
## 118	<NA>	89.40000	118	4
## 119	<NA>	87.90000	119	4
## 120	<NA>	86.35000	120	4
## 121	<NA>	86.00000	121	4
## 122	<NA>	90.20000	122	4
## 123	<NA>	90.30000	123	4
## 124	<NA>	85.90000	124	4
## 125	<NA>	83.30000	125	4
## 126	<NA>	85.40000	126	4
## 127	<NA>	86.30000	127	4
## 128	<NA>	87.70000	128	4
## 129	<NA>	85.50000	129	4
## 130	2018-02-05	84.10000	130	4
## 131	2018-03-05	82.60000	131	4
## 132	2018-05-05	85.70000	132	4
## 133	2018-06-05	88.10000	133	4
## 134	2018-07-05	89.50000	134	4
## 135	2018-08-05	84.90000	135	4
## 136	2018-09-05	85.60000	136	4
## 137	2018-10-05	89.00000	137	4
## 138	2018-11-05	87.60000	138	4
## 139	<NA>	86.60000	139	4
## 140	<NA>	83.50000	140	4
## 141	<NA>	88.70000	141	4
## 142	<NA>	84.80000	142	4
## 143	<NA>	88.00000	143	4
## 144	<NA>	89.90000	144	4
## 145	<NA>	89.90000	145	4
## 146	<NA>	88.70000	146	4
## 147	<NA>	89.50000	147	4
## 148	<NA>	90.10000	148	4
## 149	<NA>	93.25000	149	4
## 150	<NA>	87.50000	150	4
## 151	<NA>	88.80000	151	4
## 152	<NA>	89.30000	152	4
## 153	<NA>	86.60000	153	4
## 154	<NA>	85.40000	154	4
## 155	2018-01-06	88.70000	155	4
## 156	2018-02-06	87.30000	156	4
## 157	2018-04-06	87.20000	157	4
## 158	2018-05-06	86.80000	158	4
## 159	2018-06-06	85.40000	159	4
## 160	2018-07-06	85.20000	160	4
## 161	2018-08-06	84.80000	161	4
## 162	2018-09-06	85.60000	162	4

## 163	2018-10-06	87.20000	163	4
## 164	2018-11-06	86.80000	164	4
## 165	<NA>	87.20000	165	4
## 166	<NA>	85.30000	166	4
## 167	<NA>	86.50000	167	4
## 168	<NA>	85.40000	168	4
## 169	<NA>	85.40000	169	4
## 170	<NA>	86.60000	170	4
## 171	<NA>	84.70000	171	4
## 172	<NA>	87.30000	172	4
## 173	<NA>	85.50000	173	4
## 174	<NA>	86.10000	174	4
## 175	<NA>	87.40000	175	4
## 176	<NA>	86.70000	176	4
## 177	<NA>	83.30000	177	5
## 178	<NA>	82.90000	178	5
## 179	2018-02-07	83.30000	179	5
## 180	2018-04-07	83.56667	180	5
## 181	2018-05-07	83.10000	181	5
## 182	2018-06-07	83.25000	182	5
## 183	2018-07-07	83.10000	183	5
## 184	2018-09-07	84.00000	184	5
## 185	2018-11-07	83.10000	185	5
## 186	2018-12-07	83.75000	186	5
## 187	<NA>	84.20000	187	5
## 188	<NA>	83.70000	188	5
## 189	<NA>	82.90000	189	5
## 190	<NA>	84.50000	190	5
## 191	<NA>	83.10000	191	5
## 192	<NA>	84.30000	192	5
## 193	<NA>	83.10000	193	5
## 194	<NA>	84.40000	194	5
## 195	<NA>	82.80000	195	5
## 196	<NA>	82.80000	196	5
## 197	<NA>	83.20000	197	5
## 198	<NA>	83.70000	198	5
## 199	<NA>	83.87778	199	5
## 200	<NA>	83.15000	200	5
## 201	<NA>	83.80000	201	5
## 202	<NA>	83.10000	202	5
## 203	2018-01-08	83.40000	203	5
## 204	2018-02-08	83.30000	204	5
## 205	2018-03-08	83.50000	205	5
## 206	2018-04-08	83.10000	206	5
## 207	2018-06-08	83.70000	207	5
## 208	2018-07-08	82.90000	208	5
## 209	2018-08-08	84.10000	209	5
## 210	2018-09-08	83.50000	210	5
## 211	2018-10-08	83.50000	211	5
## 212	2018-11-08	84.40000	212	5
## 213	2018-12-08	84.10000	213	5
## 214	<NA>	83.10000	214	5
## 215	<NA>	82.70000	215	5
## 216	<NA>	82.80000	216	5

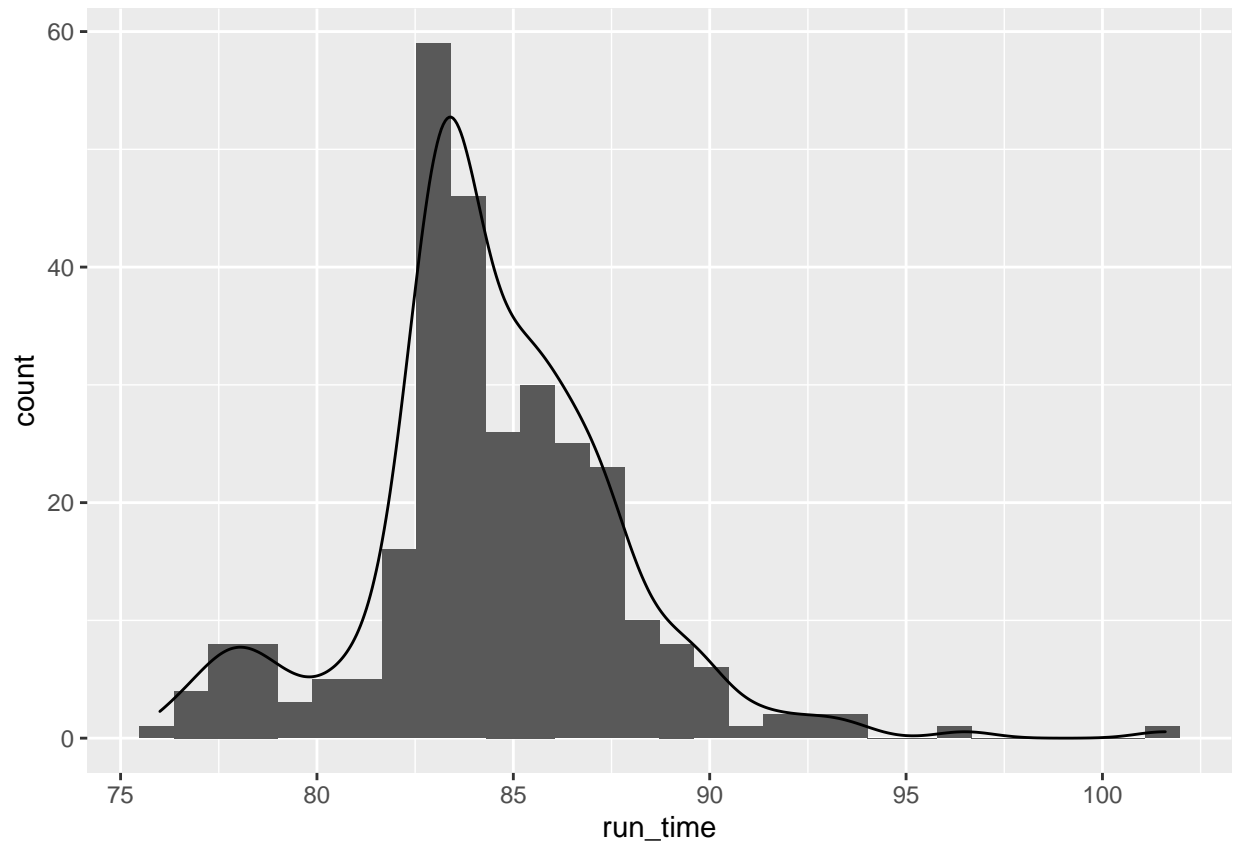
## 217	<NA>	83.20000	217	5
## 218	<NA>	82.60000	218	5
## 219	<NA>	83.10000	219	5
## 220	<NA>	83.90000	220	5
## 221	<NA>	83.70000	221	5
## 222	<NA>	83.40000	222	5
## 223	<NA>	82.50000	223	5
## 224	<NA>	83.50000	224	5
## 225	<NA>	83.70000	225	5
## 226	<NA>	83.40000	226	5
## 227	<NA>	89.80000	227	5
## 228	<NA>	86.20000	228	5
## 229	<NA>	92.70000	229	5
## 230	<NA>	92.30000	230	5
## 231	<NA>	84.70000	231	5
## 232	<NA>	85.50000	232	5
## 233	2018-01-09	84.00000	233	5
## 234	2018-02-09	83.90000	234	5
## 235	2018-03-09	84.00000	235	5
## 236	2018-04-09	83.50000	236	5
## 237	2018-05-09	83.40000	237	5
## 238	2018-06-09	83.40000	238	5
## 239	2018-07-09	83.70000	239	5
## 240	2018-09-09	84.90000	240	5
## 241	2018-11-09	82.90000	241	5
## 242	<NA>	96.50000	242	5
## 243	<NA>	83.30000	243	5
## 244	<NA>	83.80000	244	5
## 245	<NA>	83.35000	245	5
## 246	<NA>	85.10000	246	5
## 247	<NA>	85.50000	247	5
## 248	<NA>	83.10000	248	5
## 249	<NA>	83.50000	249	5
## 250	<NA>	84.20000	250	5
## 251	<NA>	84.90000	251	5
## 252	<NA>	83.80000	252	5
## 253	<NA>	82.70000	253	5
## 254	<NA>	91.50000	254	5
## 255	<NA>	83.10000	255	5
## 256	2018-01-10	85.00000	256	5
## 257	2018-02-10	82.70000	257	5
## 258	2018-03-10	83.50000	258	5
## 259	2018-04-10	83.90000	259	5
## 260	2018-05-10	87.80000	260	5
## 261	2018-06-10	83.80000	261	5
## 262	2018-07-10	84.30000	262	5
## 263	2018-09-10	83.90000	263	5
## 264	2018-10-10	86.10000	264	5
## 265	2018-11-10	83.50000	265	5
## 266	2018-12-10	83.60000	266	5
## 267	<NA>	86.30000	267	5
## 268	<NA>	83.30000	268	5
## 269	<NA>	93.70000	269	5
## 270	<NA>	82.80000	270	5

```
## 271      <NA> 83.40000      271  5
## 272      <NA> 83.00000      272  5
## 273      <NA> 82.10000      273  5
## 274      <NA> 82.40000      274  5
## 275      <NA> 82.80000      275  5
## 276      <NA> 82.60000      276  5
## 277      <NA> 82.10000      277  5
## 278      <NA> 81.80000      278  5
## 279      <NA> 84.50000      279  5
## 280      <NA> 81.90000      280  5
## 281      <NA> 83.00000      281  5
## 282      <NA> 82.60000      282  5
## 283      <NA> 83.00000      283  5
## 284      <NA> 83.90000      284  5
## 285      <NA> 84.40000      285  5
## 286 2018-01-11 84.40000      286  5
## 287 2018-02-11 82.40000      287  5
## 288 2018-03-11 82.80000      288  5
## 289 2018-04-11 83.00000      289  5
## 290 2018-05-11 83.10000      290  5
## 291 2018-06-11 83.10000      291  5
## 292 2018-07-11 82.30000      292  5
```

1.4 Make a histogram plot of all run times. (10 points)

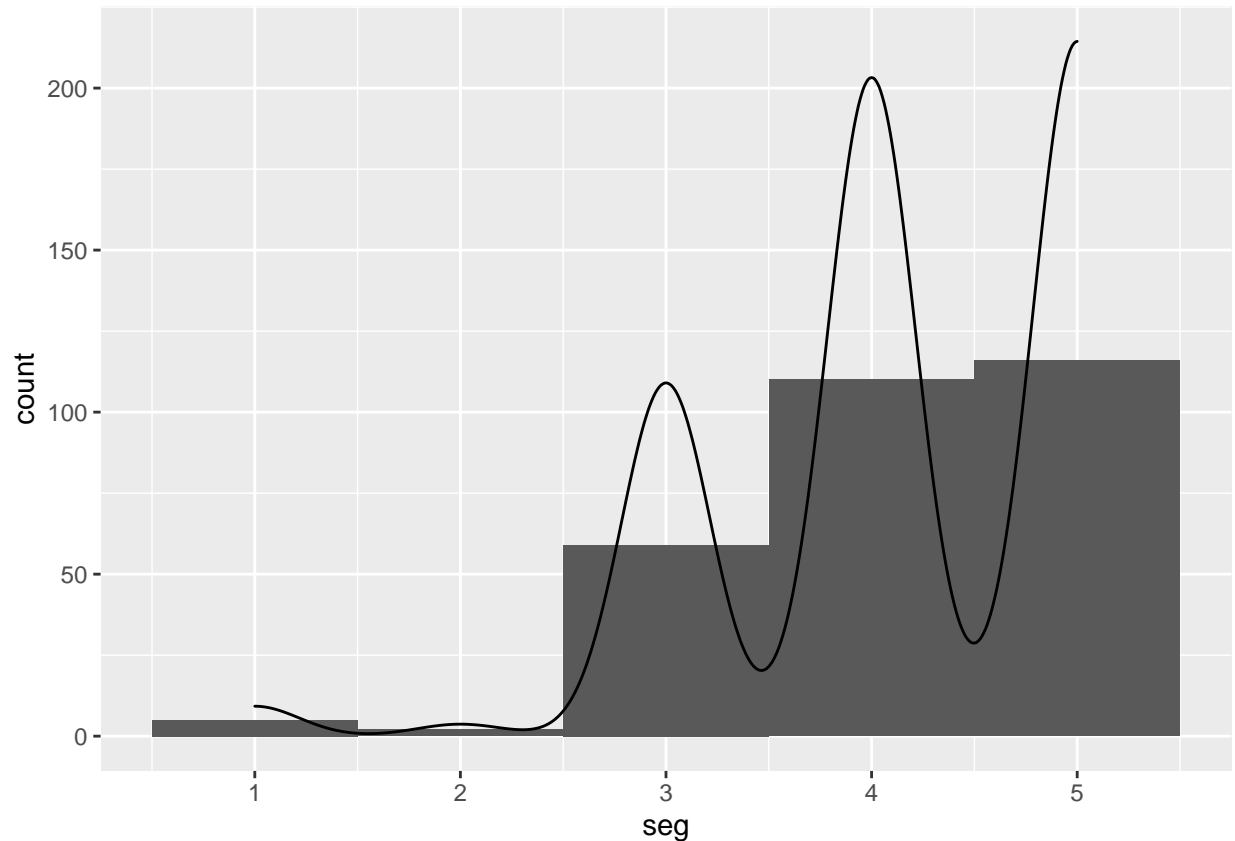
```
ggplot(seg_data, aes(x =run_time)) + geom_histogram()+ geom_density(aes(y=..count..))
```

```
## 'stat_bin()' using 'bins = 30'. Pick better value with 'binwidth'.
```



1.5 Make a histogram plot of for each segments. (10 points)

```
ggplot(seg_data, aes(x=seg))+geom_histogram(binwidth=1)+geom_density(aes(y=..count..))
```



1.6 Does it look reasonably normal? (10 points)

From figure 1.4, we can tell that it is a right skewed but a reasonably normal curve.

1.7 Identify change points with `cpt.meanvar` function. Use PELT method and Normal for `test.stat`. Plot your data with identified segments mean. (10 points)

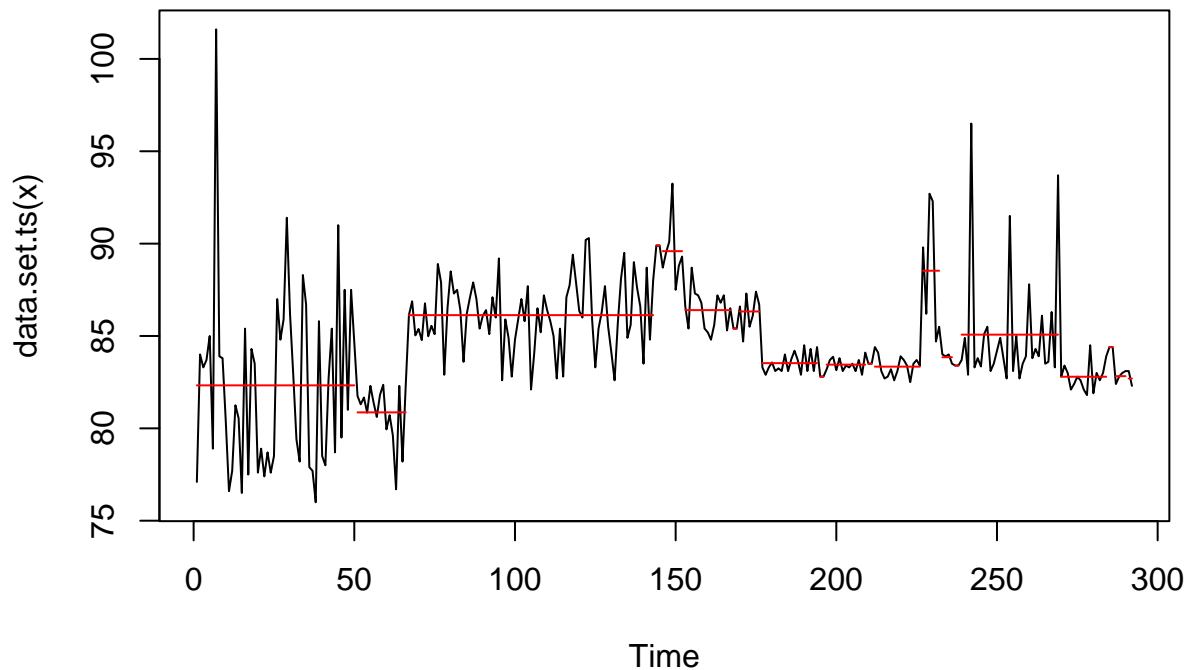
```
set.seed(49)
df_meanvar = cpt.meanvar(df$run_time, test.stat='Normal', method='PELT',penalty = 'SIC')
cpts(df_meanvar)
```

```
## [1] 50 66 143 145 152 167 169 176 194 196 209 211 226 232 236 238 269 284 286
## [20] 290
```

```
pen.value(df_meanvar)
```

```
## [1] 17.03026
```

```
plot(df_meanvar)
```



hints: run `cpt.meanvar` on the `run_time` column (i.e. `df$run_time`)

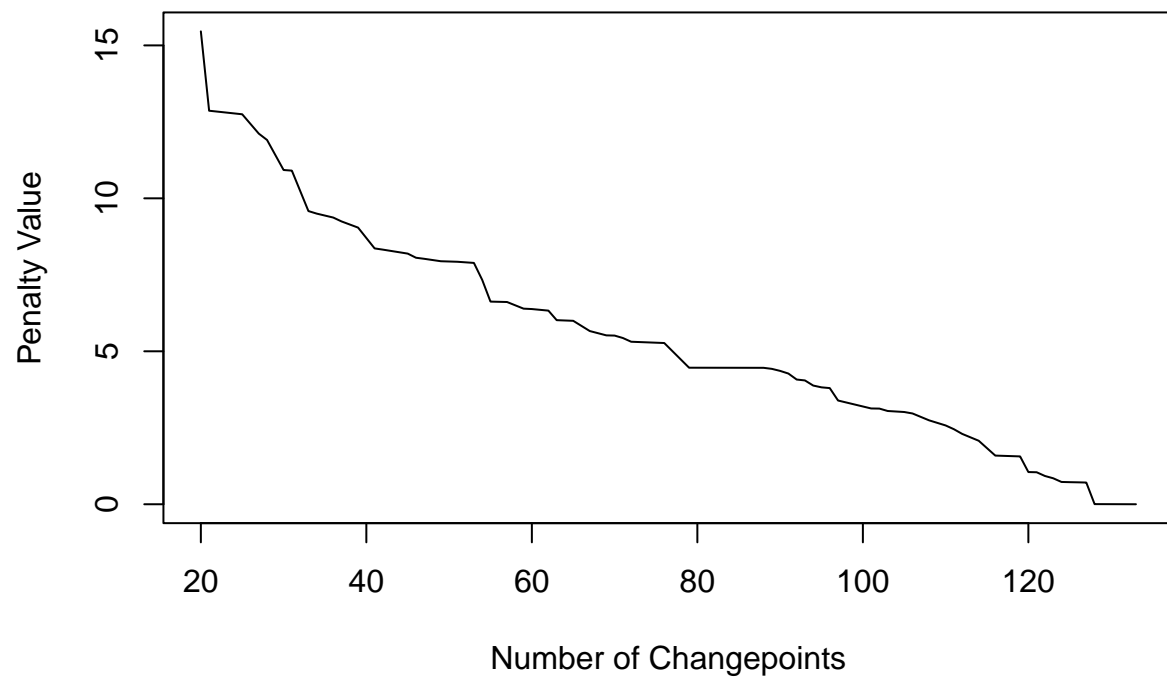
use `pen.value` function to see current value of penalty (MBIC value), use that value as guide for your penalty range in next question.

1.8 Using CROPS procedure find optimal number of points. Plot data with optimal number of segments. (10 points)

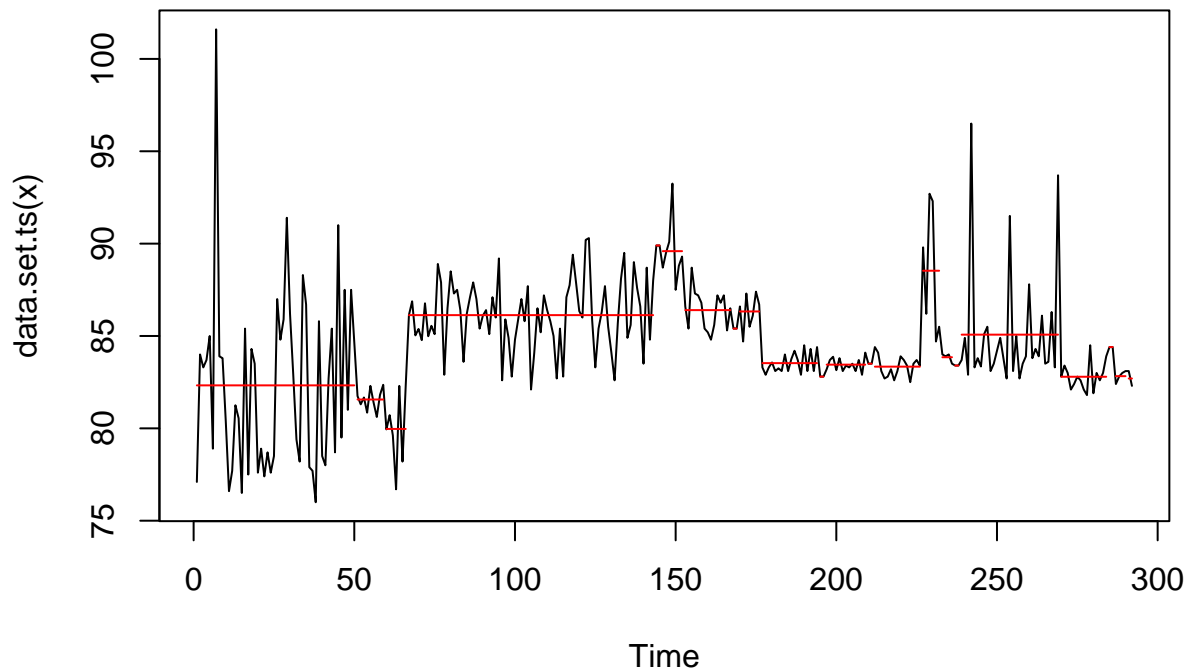
```
df_meanvar_crops = cpt.meanvar(df$run_time, method="PELT", penalty="CROPS",
                               pen.value=c(0, pen.value(df_meanvar)))
```

```
## [1] "Maximum number of runs of algorithm = 115"
## [1] "Completed runs = 2"
## [1] "Completed runs = 3"
## [1] "Completed runs = 5"
## [1] "Completed runs = 9"
## [1] "Completed runs = 17"
## [1] "Completed runs = 33"
## [1] "Completed runs = 59"
## [1] "Completed runs = 85"
## [1] "Completed runs = 93"
## [1] "Completed runs = 95"
```

```
plot(df_meanvar_crops, diagnostic=TRUE)
```



```
plot(df_meanvar_crops, ncpts=21)
```



1.9 Does your initial segment guess matches with optimized by CROPS? (10 points)

our guess was 5 with mean and 20 using mean var. Our guess matches with the one optimized with CROPS

1.10 The run-time in this example does not really follow normal distribution. What to do you think can we still use this method to identify changepoints? (10 points)

mean, var and meanvar assumes that the data is normalized. We need to normalize the data by differencing it such that the mean fits over x axis(mean =0) and the variance is 1.

PS. Just in case if you wonder. On 2018-02-21 system got a critical linux kernel update to alleviate Meltdown-Spectre vulnerabilities. On 2018-06-28 system got another kernel update which is more robust and hit the performance less