

Final Exam

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PRE-PROCESSING

We start by setting the random seed and loading useful libraries.

```
set.seed(02022018)
require(RCurl)
require(gdata)
library(dplyr)
library(magrittr)
library(ggplot2)
library(MASS)
#library(MVN) #multivariate normality test
#library(MASS)
#library(caret)
```

Now we load our data frames for the exercises.

```
urls <- c("https://www.mimuw.edu.pl/~noble/courses/SDA/data/ushighways.txt",
          "https://www.mimuw.edu.pl/~noble/courses/SDA/data/PET.txt",
          "https://www.mimuw.edu.pl/~noble/courses/SDA/data/ozone.csv",
          "https://www.mimuw.edu.pl/~noble/courses/SDA/data/pendigits.txt",
          "https://www.mimuw.edu.pl/~noble/courses/SDA/data/carmarks.txt",
          "https://www.mimuw.edu.pl/~noble/courses/SDA/data/primate.scapulae.txt")
ushighways <- read.table(urls[1], header=T)
#download.file("https://www.mimuw.edu.pl/~noble/courses/SDA/data/bodyfat2.XLS", destfile="bodyfat2.xls")
bodyfat <- read.xls("bodyfat2.xlsx")
#data(yarn, package="pls")
yarn <- read.table(urls[2], header=T)
ozone <- read.csv(urls[3])
pendigits <- read.table(urls[4])
carmarks <- read.table(urls[5], sep=";", header=T)
scapular <- read.table(urls[6], sep=" ", header=T)
```

EXERCISE 1

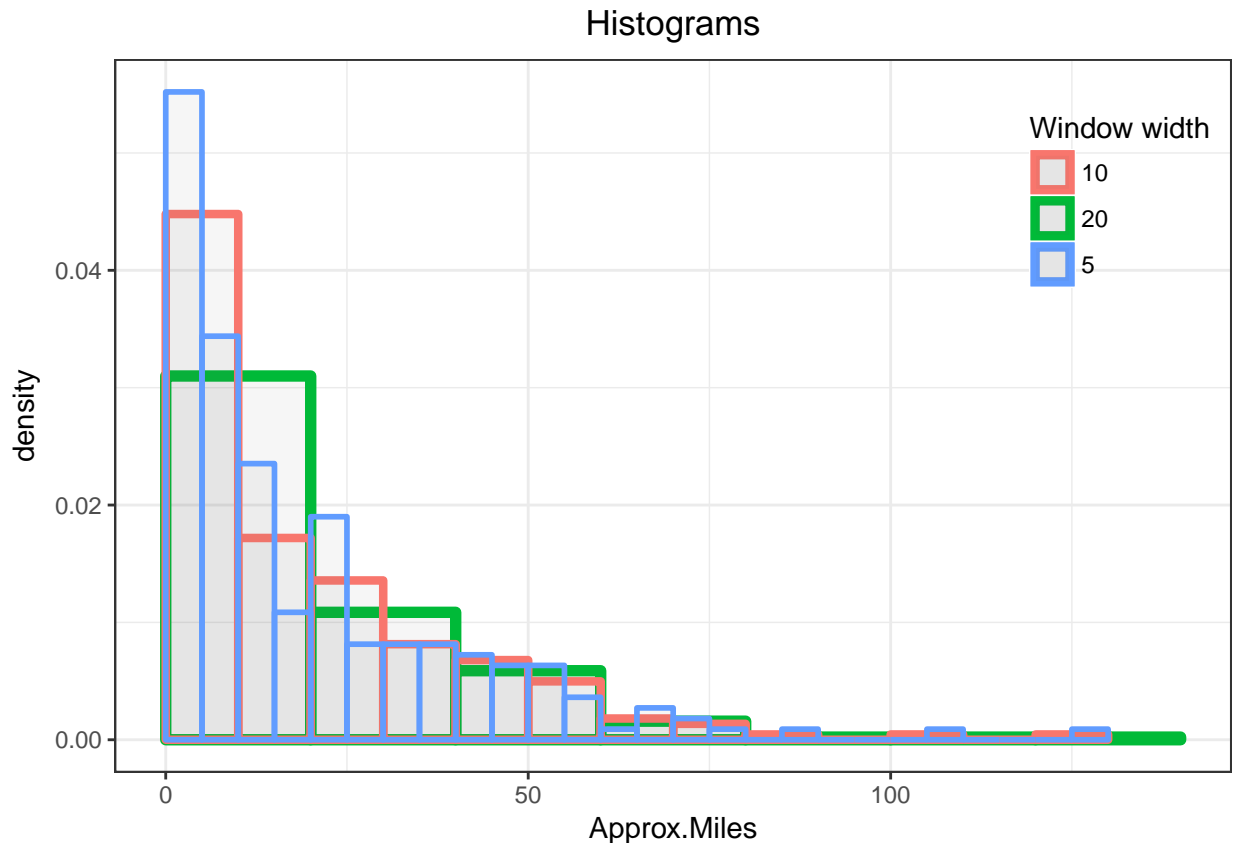
Let us take a look at data.

```
head(ushighways)
```

##	Interstate..	State	Approx.Miles	Location
## 1	165	AL	4	MOBILE
## 2	359	AL	4	TUSCALOOSA
## 3	459	AL	32	BIRMINGHAM
## 4	759	AL	4	GADSDEN
## 5	565	AL	21	HUNTSVILLE
## 6	430	AR	13	LITTLE ROCK

We first draw some histograms with different window widths.

```
ggplot(ushighways, aes(x=Approx.Miles)) +
  geom_histogram(size=2, alpha=0.05, binwidth=20, boundary=0,
    aes(color="20", y=..density..)) +
  geom_histogram(size=1.5, alpha=0.05, binwidth=10, boundary=0,
    aes(color="10", y=..density..)) +
  geom_histogram(size=1, alpha=0.05, binwidth=5, boundary=0,
    aes(color="5", y=..density..)) +
  theme_bw() + theme(plot.title = element_text(hjust = 0.5)) + ggtitle("Histograms") +
  guides(color=guide_legend(title="Window width")) +
  theme(legend.position=c(0.9,0.8), legend.background=element_rect(fill=alpha('blue',0)))
```

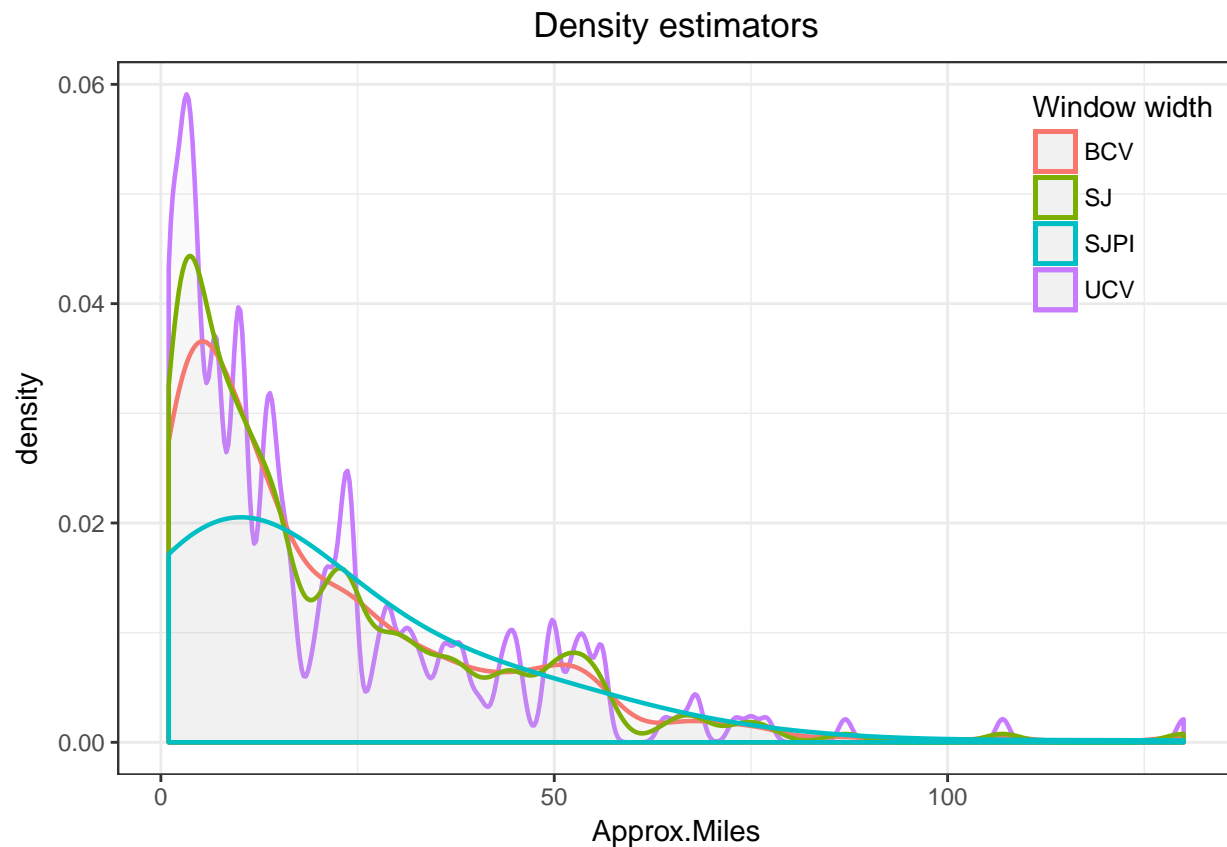


The second picture with window width of 10 miles looks quite good. It is well-splitted and shows the monotonicity of the distribution.

Now we will use UCV, BCV and SJPI estimators for window width to plot densities with gaussian kernel.

```
sjpi <- width.SJ(ushighways$Approx.Miles, method="dpi")
ggplot(ushighways, aes(x=Approx.Miles)) +
  geom_density(fill="grey", size=0.8, alpha=0.05, bw="ucv", aes(color="UCV")) +
  geom_density(fill="grey", size=0.8, alpha=0.05, bw="bcv", aes(color="BCV")) +
  geom_density(fill="grey", size=0.8, alpha=0.05, bw="sj", aes(color="SJ")) +
  geom_density(fill="grey", size=0.8, alpha=0.05, bw=sjpi, aes(color="SJPI")) +
  theme_bw() + theme(plot.title = element_text(hjust = 0.5)) +
  ggtitle("Density estimators") + guides(color=guide_legend(title="Window width")) +
  theme(legend.position=c(0.9,0.8), legend.background=element_rect(fill=alpha('blue',0)))
```

```
## Warning in bw.ucv(x): minimum occurred at one end of the range
```



The BCV estimator (red one) looks quite well-fitted and monotonic. The other kernel choices gave very similar results.

EXERCISE 2

Let us take a look at data.

```
head(bodyfat)
```

```
##   density bodyfat age weight height neck chest abdomen  hip thigh knee
## 1  1.0708   12.3  23 154.25  67.75 36.2  93.1    85.2  94.5  59.0 37.3
## 2  1.0853    6.1  22 173.25  72.25 38.5  93.6    83.0  98.7  58.7 37.3
## 3  1.0414   25.3  22 154.00  66.25 34.0  95.8    87.9  99.2  59.6 38.9
## 4  1.0751   10.4  26 184.75  72.25 37.4 101.8    86.4 101.2  60.1 37.3
## 5  1.0340   28.7  24 184.25  71.25 34.4  97.3   100.0 101.9  63.2 42.2
## 6  1.0502   20.9  24 210.25  74.75 39.0 104.5    94.4 107.8  66.0 42.0
##   ankle biceps forearm wrist
## 1  21.9   32.0   27.4  17.1
## 2  23.4   30.5   28.9  18.2
## 3  24.0   28.8   25.2  16.6
## 4  22.8   32.4   29.4  18.2
## 5  24.0   32.2   27.7  17.7
## 6  25.6   35.7   30.6  18.8
```

EXERCISE 3

Let us take a look at data.

```
head(yarn)
```

##	X.1	X.2	X.3	X.4	X.5	X.6	X.7	X.8	X.9	X.10
## 1	3.0663	3.0861	3.1079	3.0972	2.9979	2.8273	2.6233	2.4039	2.1931	2.0058
## 2	3.0675	3.0857	3.0958	3.0692	2.9818	2.8408	2.6760	2.5059	2.3506	2.2230
## 3	3.0750	3.0966	3.0916	3.0288	2.8849	2.6885	2.4764	2.2694	2.0824	1.9195
## 4	3.0828	3.0973	3.1010	3.0735	2.9913	2.8709	2.7392	2.6102	2.5000	2.4237
## 5	3.1029	3.1034	3.0848	3.0228	2.8927	2.7159	2.5384	2.3764	2.2397	2.1346
## 6	3.0815	3.0849	3.0487	2.9305	2.7323	2.5089	2.2944	2.0995	1.9328	1.7925
##	X.11	X.12	X.13	X.14	X.15	X.16	X.17	X.18	X.19	X.20
## 1	1.8379	1.6907	1.5777	1.5033	1.4381	1.3373	1.2206	1.1410	1.1161	1.1329
## 2	2.1192	2.0352	1.9795	1.9554	1.9396	1.8931	1.8198	1.7550	1.7122	1.6832
## 3	1.7747	1.6490	1.5533	1.4907	1.4363	1.3540	1.2603	1.1979	1.1788	1.1900
## 4	2.3774	2.3526	2.3506	2.3721	2.4009	2.4042	2.3756	2.3321	2.2768	2.2053
## 5	2.0534	1.9909	1.9522	1.9393	1.9337	1.9047	1.8535	1.8047	1.7650	1.7266
## 6	1.6693	1.5633	1.4836	1.4310	1.3851	1.3193	1.2468	1.1999	1.1851	1.1896
##	X.21	X.22	X.23	X.24	X.25	X.26	X.27	X.28	X.29	X.30
## 1	1.1604	1.1638	1.1527	1.1515	1.1277	1.0681	1.0197	1.0084	1.0321	1.0828
## 2	1.6461	1.5810	1.5091	1.4521	1.3802	1.2865	1.2045	1.1508	1.1308	1.1438
## 3	1.2053	1.1985	1.1804	1.1727	1.1513	1.1042	1.0665	1.0593	1.0807	1.1245
## 4	2.1085	1.9815	1.8544	1.7459	1.6310	1.5083	1.3954	1.2976	1.2297	1.2022
## 5	1.6753	1.5999	1.5207	1.4575	1.3898	1.3114	1.2421	1.1904	1.1630	1.1655
## 6	1.1931	1.1778	1.1538	1.1405	1.1220	1.0875	1.0604	1.0562	1.0742	1.1106
##	X.31	X.32	X.33	X.34	X.35	X.36	X.37	X.38	X.39	X.40
## 1	1.1408	1.1788	1.2018	1.2349	1.2776	1.3125	1.3431	1.3766	1.4145	1.4571
## 2	1.1761	1.2041	1.2292	1.2690	1.3180	1.3574	1.3924	1.4335	1.4792	1.5239
## 3	1.1751	1.2105	1.2357	1.2724	1.3204	1.3659	1.4096	1.4529	1.4976	1.5460
## 4	1.2088	1.2278	1.2553	1.3016	1.3567	1.4006	1.4399	1.4881	1.5412	1.5877
## 5	1.1901	1.2163	1.2443	1.2880	1.3421	1.3918	1.4400	1.4908	1.5430	1.5931
## 6	1.1542	1.1873	1.2147	1.2547	1.3070	1.3618	1.4175	1.4692	1.5190	1.5724
##	X.41	X.42	X.43	X.44	X.45	X.46	X.47	X.48	X.49	X.50
## 1	1.5031	1.5457	1.5762	1.6084	1.6605	1.7264	1.7977	1.8791	1.9835	2.1216
## 2	1.5657	1.6045	1.6366	1.6725	1.7260	1.7932	1.8686	1.9561	2.0683	2.2159
## 3	1.5938	1.6320	1.6575	1.6918	1.7518	1.8215	1.8862	1.9618	2.0727	2.2230
## 4	1.6249	1.6592	1.6928	1.7325	1.7870	1.8548	1.9343	2.0282	2.1481	2.3036
## 5	1.6366	1.6709	1.6982	1.7365	1.7976	1.8684	1.9377	2.0199	2.1391	2.2995
## 6	1.6216	1.6547	1.6754	1.7124	1.7793	1.8510	1.9086	1.9792	2.0954	2.2553
##	X.51	X.52	X.53	X.54	X.55	X.56	X.57	X.58	X.59	X.60
## 1	2.2734	2.4047	2.5122	2.6176	2.7415	2.8889	3.0358	3.1401	3.1976	3.2279
## 2	2.3712	2.4851	2.5590	2.6399	2.7597	2.9196	3.0851	3.1855	3.2077	3.2086
## 3	2.3876	2.5381	2.6614	2.7637	2.8644	2.9774	3.0972	3.1887	3.2271	3.2448
## 4	2.4589	2.5513	2.5923	2.6528	2.7686	2.9346	3.1110	3.2062	3.2019	3.1915
## 5	2.4665	2.5950	2.6826	2.7628	2.8580	2.9830	3.1250	3.2115	3.2239	3.2267
## 6	2.4310	2.5975	2.7325	2.8296	2.9058	2.9825	3.0743	3.1621	3.2042	3.2293
##	X.61	X.62	X.63	X.64	X.65	X.66	X.67	X.68	X.69	X.70
## 1	3.2512	3.2835	3.3212	3.3514	3.3670	3.3590	3.3159	3.2437	3.1262	2.9523
## 2	3.2257	3.2590	3.2948	3.3264	3.3504	3.3609	3.3297	3.2577	3.1240	2.8986
## 3	3.2838	3.3285	3.3538	3.3764	3.3896	3.3667	3.3145	3.2266	3.0409	2.8017
## 4	3.2100	3.2352	3.2603	3.2975	3.3337	3.3377	3.2949	3.2351	3.0970	2.8169
## 5	3.2484	3.2761	3.2982	3.3229	3.3476	3.3372	3.2867	3.2119	3.0177	2.7192
## 6	3.2794	3.3154	3.3222	3.3380	3.3533	3.3176	3.2418	3.1224	2.8633	2.5717

##	X.71	X.72	X.73	X.74	X.75	X.76	X.77	X.78	X.79	X.80
## 1	2.7794	2.6562	2.5837	2.5444	2.5277	2.5333	2.5613	2.6092	2.6722	2.7478
## 2	2.6815	2.5399	2.4638	2.4274	2.4160	2.4249	2.4540	2.5024	2.5677	2.6458
## 3	2.6094	2.4879	2.4199	2.3875	2.3800	2.3926	2.4244	2.4747	2.5401	2.6181
## 4	2.5605	2.4060	2.3280	2.2933	2.2850	2.2969	2.3275	2.3747	2.4371	2.5129
## 5	2.4807	2.3432	2.2758	2.2470	2.2416	2.2552	2.2888	2.3401	2.4053	2.4832
## 6	2.3675	2.2514	2.1925	2.1672	2.1641	2.1799	2.2140	2.2654	2.3321	2.4104
##	X.81	X.82	X.83	X.84	X.85	X.86	X.87	X.88	X.89	X.90
## 1	2.8329	2.9196	3.0015	3.0733	3.1329	3.1796	3.2105	3.2331	3.2502	3.2586
## 2	2.7312	2.8192	2.9053	2.9842	3.0542	3.1125	3.1623	3.2021	3.2261	3.2396
## 3	2.7061	2.7947	2.8782	2.9610	3.0397	3.1053	3.1609	3.2047	3.2294	3.2436
## 4	2.5980	2.6879	2.7764	2.8610	2.9408	3.0118	3.0706	3.1137	3.1476	3.1750
## 5	2.5701	2.6603	2.7483	2.8323	2.9123	2.9827	3.0424	3.0970	3.1408	3.1664
## 6	2.4958	2.5848	2.6741	2.7634	2.8494	2.9249	2.9928	3.0519	3.0922	3.1194
##	X.91	X.92	X.93	X.94	X.95	X.96	X.97	X.98	X.99	X.100
## 1	3.2692	3.2772	3.2773	3.2683	3.2452	3.2278	3.2205	3.2091	3.1986	3.1937
## 2	3.2471	3.2524	3.2528	3.2509	3.2488	3.2413	3.2300	3.2125	3.1908	3.1746
## 3	3.2559	3.2638	3.2606	3.2489	3.2372	3.2255	3.2127	3.1988	3.1850	3.1696
## 4	3.1908	3.1965	3.1910	3.1808	3.1767	3.1746	3.1650	3.1493	3.1331	3.1134
## 5	3.1820	3.1919	3.1930	3.1874	3.1768	3.1660	3.1556	3.1399	3.1204	3.0996
## 6	3.1398	3.1507	3.1530	3.1497	3.1416	3.1266	3.1072	3.0852	3.0615	3.0416
##	X.101	X.102	X.103	X.104	X.105	X.106	X.107	X.108	X.109	X.110
## 1	3.1872	3.1732	3.1540	3.1392	3.1267	3.1084	3.0922	3.0802	3.0668	3.0553
## 2	3.1665	3.1568	3.1366	3.1168	3.1036	3.0870	3.0661	3.0470	3.0316	3.0190
## 3	3.1501	3.1299	3.1112	3.0920	3.0760	3.0619	3.0432	3.0224	3.0055	2.9947
## 4	3.0893	3.0717	3.0611	3.0451	3.0224	3.0018	2.9854	2.9685	2.9541	2.9456
## 5	3.0812	3.0595	3.0324	3.0106	2.9907	2.9682	2.9469	2.9272	2.9113	2.8994
## 6	3.0204	2.9934	2.9691	2.9451	2.9194	2.8962	2.8758	2.8590	2.8415	2.8221
##	X.111	X.112	X.113	X.114	X.115	X.116	X.117	X.118	X.119	X.120
## 1	3.0462	3.0344	3.0206	3.0060	2.9881	2.9677	2.9488	2.9304	2.9088	2.8820
## 2	3.0068	2.9941	2.9808	2.9678	2.9513	2.9296	2.9083	2.8884	2.8676	2.8408
## 3	2.9810	2.9622	2.9463	2.9300	2.9091	2.8866	2.8636	2.8374	2.8058	2.7747
## 4	2.9372	2.9255	2.9097	2.8933	2.8791	2.8601	2.8397	2.8201	2.7973	2.7751
## 5	2.8890	2.8770	2.8599	2.8403	2.8206	2.7990	2.7748	2.7497	2.7230	2.6925
## 6	2.8058	2.7915	2.7768	2.7584	2.7324	2.7034	2.6764	2.6493	2.6189	2.5847
##	X.121	X.122	X.123	X.124	X.125	X.126	X.127	X.128	X.129	X.130
## 1	2.8537	2.8241	2.7893	2.7557	2.7260	2.6924	2.6536	2.6072	2.5405	2.4500
## 2	2.8082	2.7738	2.7401	2.7084	2.6762	2.6422	2.6003	2.5485	2.4831	2.3965
## 3	2.7436	2.7049	2.6659	2.6288	2.5885	2.5453	2.4986	2.4435	2.3701	2.2787
## 4	2.7500	2.7165	2.6809	2.6493	2.6203	2.5863	2.5449	2.4958	2.4319	2.3520
## 5	2.6586	2.6206	2.5821	2.5459	2.5079	2.4664	2.4211	2.3667	2.2955	2.2098
## 6	2.5470	2.5057	2.4638	2.4226	2.3782	2.3305	2.2813	2.2237	2.1507	2.0656
##	X.131	X.132	X.133	X.134	X.135	X.136	X.137	X.138	X.139	X.140
## 1	2.3520	2.2615	2.1826	2.1145	2.0522	1.9894	1.9224	1.8526	1.7848	1.7248
## 2	2.3020	2.2142	2.1365	2.0681	2.0062	1.9456	1.8817	1.8135	1.7458	1.6854
## 3	2.1837	2.0978	2.0253	1.9637	1.9081	1.8537	1.7964	1.7350	1.6739	1.6201
## 4	2.2644	2.1799	2.1048	2.0383	1.9779	1.9197	1.8590	1.7927	1.7250	1.6643
## 5	2.1224	2.0418	1.9710	1.9099	1.8552	1.8022	1.7473	1.6876	1.6259	1.5709
## 6	1.9802	1.9046	1.8417	1.7885	1.7406	1.6948	1.6474	1.5950	1.5400	1.4918
##	X.141	X.142	X.143	X.144	X.145	X.146	X.147	X.148	X.149	X.150
## 1	1.6786	1.6477	1.6238	1.5915	1.5351	1.4468	1.3313	1.2072	1.0979	1.02230
## 2	1.6382	1.6045	1.5763	1.5392	1.4800	1.3956	1.2905	1.1817	1.0890	1.02460
## 3	1.5811	1.5590	1.5449	1.5200	1.4682	1.3854	1.2793	1.1715	1.0839	1.02640
## 4	1.6158	1.5796	1.5477	1.5049	1.4414	1.3581	1.2609	1.1660	1.0887	1.03470

```

## 5 1.5303 1.5058 1.4880 1.4576 1.4014 1.3203 1.2235 1.1313 1.0605 1.01370
## 6 1.4601 1.4474 1.4428 1.4239 1.3743 1.2934 1.1933 1.0994 1.0309 0.98797
##      X.151      X.152      X.153      X.154      X.155      X.156      X.157      X.158      X.159
## 1 0.98235 0.96873 0.96943 0.97208 0.96891 0.95732 0.93511 0.90178 0.86321
## 2 0.98747 0.96987 0.96509 0.96567 0.96122 0.94465 0.91598 0.87981 0.84280
## 3 0.99373 0.97850 0.97561 0.97556 0.96874 0.95167 0.92322 0.88488 0.84445
## 4 0.99942 0.97663 0.96524 0.96287 0.95665 0.93488 0.89989 0.86141 0.82631
## 5 0.98274 0.96252 0.95336 0.95079 0.94237 0.91993 0.88499 0.84439 0.80603
## 6 0.95948 0.94130 0.93451 0.93157 0.92117 0.89906 0.86508 0.82293 0.78174
##      X.160      X.161      X.162      X.163      X.164      X.165      X.166      X.167      X.168
## 1 0.82637 0.79551 0.76532 0.72361 0.66715 0.60730 0.55709 0.51899 0.48948
## 2 0.80988 0.78279 0.75227 0.70640 0.64862 0.59210 0.54623 0.51148 0.48445
## 3 0.80914 0.78145 0.75225 0.70873 0.65284 0.59801 0.55402 0.52105 0.49547
## 4 0.79731 0.77390 0.74286 0.69303 0.63437 0.58120 0.53939 0.50764 0.48281
## 5 0.77479 0.75075 0.72096 0.67334 0.61650 0.56502 0.52510 0.49511 0.47173
## 6 0.74842 0.72380 0.69576 0.65136 0.59709 0.54753 0.50948 0.48122 0.45923
##      X.169      X.170      X.171      X.172      X.173      X.174      X.175      X.176      X.177
## 1 0.46553 0.44558 0.42895 0.41522 0.40396 0.39468 0.38694 0.38034 0.37455
## 2 0.46251 0.44425 0.42903 0.41642 0.40605 0.39750 0.39039 0.38431 0.37903
## 3 0.47468 0.45751 0.44333 0.43155 0.42185 0.41395 0.40746 0.40193 0.39712
## 4 0.46266 0.44592 0.43195 0.42036 0.41079 0.40287 0.39628 0.39071 0.38591
## 5 0.45281 0.43720 0.42423 0.41349 0.40468 0.39742 0.39141 0.38635 0.38202
## 6 0.44142 0.42680 0.41478 0.40487 0.39673 0.39006 0.38461 0.38008 0.37623
##      X.178      X.179      X.180      X.181      X.182      X.183      X.184      X.185      X.186
## 1 0.36945 0.36513 0.36175 0.35944 0.35826 0.35811 0.35872 0.35992 0.36171
## 2 0.37447 0.37064 0.36768 0.36577 0.36489 0.36488 0.36562 0.36691 0.36866
## 3 0.39299 0.38954 0.38688 0.38524 0.38457 0.38469 0.38553 0.38706 0.38911
## 4 0.38183 0.37850 0.37599 0.37442 0.37380 0.37399 0.37479 0.37609 0.37788
## 5 0.37833 0.37537 0.37320 0.37189 0.37146 0.37181 0.37278 0.37425 0.37624
## 6 0.37296 0.37035 0.36846 0.36737 0.36711 0.36756 0.36865 0.37033 0.37259
##      X.187      X.188      X.189      X.190      X.191      X.192      X.193      X.194      X.195
## 1 0.36414 0.36730 0.37142 0.37653 0.38241 0.38869 0.39506 0.40145 0.40788
## 2 0.37109 0.37435 0.37843 0.38336 0.38900 0.39496 0.40097 0.40699 0.41298
## 3 0.39170 0.39503 0.39923 0.40425 0.40998 0.41618 0.42245 0.42862 0.43474
## 4 0.38036 0.38365 0.38773 0.39255 0.39793 0.40360 0.40927 0.41487 0.42051
## 5 0.37888 0.38223 0.38637 0.39128 0.39680 0.40263 0.40853 0.41435 0.42003
## 6 0.37538 0.37875 0.38290 0.38784 0.39345 0.39948 0.40562 0.41163 0.41735
##      X.196      X.197      X.198      X.199      X.200      X.201      X.202      X.203      X.204
## 1 0.41431 0.42052 0.42601 0.42998 0.43154 0.43035 0.42697 0.42250 0.41825
## 2 0.41893 0.42476 0.43000 0.43390 0.43583 0.43557 0.43358 0.43088 0.42864
## 3 0.44069 0.44627 0.45129 0.45512 0.45696 0.45673 0.45504 0.45273 0.45082
## 4 0.42615 0.43160 0.43654 0.44044 0.44285 0.44362 0.44316 0.44230 0.44209
## 5 0.42554 0.43082 0.43559 0.43939 0.44176 0.44263 0.44242 0.44197 0.44220
## 6 0.42276 0.42787 0.43249 0.43615 0.43842 0.43932 0.43933 0.43920 0.43979
##      X.205      X.206      X.207      X.208      X.209      X.210      X.211      X.212      X.213
## 1 0.41552 0.41521 0.41748 0.42162 0.42685 0.43381 0.44468 0.46086 0.48147
## 2 0.42798 0.42972 0.43389 0.43978 0.44676 0.45544 0.46760 0.48429 0.50455
## 3 0.45056 0.45277 0.45739 0.46374 0.47114 0.48019 0.49281 0.51030 0.53191
## 4 0.44359 0.44740 0.45344 0.46113 0.46992 0.48031 0.49373 0.51083 0.53064
## 5 0.44408 0.44825 0.45475 0.46290 0.47207 0.48282 0.49665 0.51445 0.53545
## 6 0.44190 0.44624 0.45299 0.46139 0.47066 0.48146 0.49555 0.51388 0.53584
##      X.214      X.215      X.216      X.217      X.218      X.219      X.220      X.221      X.222
## 1 0.50364 0.52455 0.54347 0.56134 0.57942 0.60036 0.62888 0.66760 0.71161
## 2 0.52586 0.54594 0.56418 0.58120 0.59803 0.61709 0.64239 0.67568 0.71198

```

```

## 3 0.55492 0.57674 0.59689 0.61636 0.63667 0.66085 0.69249 0.73179 0.77257
## 4 0.55115 0.57042 0.58781 0.60388 0.61953 0.63663 0.65837 0.68569 0.71397
## 5 0.55755 0.57858 0.59783 0.61613 0.63512 0.65721 0.68492 0.71795 0.75057
## 6 0.55931 0.58178 0.60263 0.62315 0.64539 0.67225 0.70594 0.74466 0.78146
##      X.223      X.224      X.225      X.226      X.227      X.228      X.229      X.230      X.231
## 1 0.74849 0.76797 0.77298 0.77332 0.77099 0.76190 0.74646 0.73044 0.71725
## 2 0.74100 0.75583 0.76058 0.76421 0.76916 0.77233 0.77165 0.76655 0.75539
## 3 0.80412 0.81955 0.82342 0.82480 0.82465 0.81828 0.80530 0.78937 0.77181
## 4 0.73535 0.74604 0.75098 0.75804 0.77037 0.78596 0.79992 0.80539 0.79578
## 5 0.77430 0.78573 0.79025 0.79536 0.80285 0.80953 0.81181 0.80642 0.78996
## 6 0.80759 0.81979 0.82370 0.82656 0.82844 0.82490 0.81444 0.79810 0.77563
##      X.232      X.233      X.234      X.235      X.236      X.237      X.238      X.239      X.240
## 1 0.70428 0.68814 0.66857 0.64598 0.62081 0.59539 0.57245 0.55306 0.53702
## 2 0.73598 0.71012 0.68303 0.65673 0.63022 0.60396 0.57987 0.55894 0.54092
## 3 0.75052 0.72540 0.69906 0.67289 0.64675 0.62115 0.59753 0.57685 0.55948
## 4 0.76955 0.73416 0.70002 0.67048 0.64306 0.61620 0.59079 0.56798 0.54782
## 5 0.76176 0.72714 0.69369 0.66426 0.63709 0.61060 0.58543 0.56287 0.54336
## 6 0.74663 0.71376 0.68164 0.65282 0.62640 0.60074 0.57620 0.55427 0.53596
##      X.241      X.242      X.243      X.244      X.245      X.246      X.247      X.248      X.249
## 1 0.52345 0.51154 0.50104 0.49145 0.48152 0.47018 0.45690 0.44182 0.42590
## 2 0.52500 0.51086 0.49840 0.48715 0.47660 0.46636 0.45571 0.44413 0.43148
## 3 0.54519 0.53283 0.52160 0.51141 0.50170 0.49144 0.47986 0.46675 0.45222
## 4 0.52966 0.51330 0.49874 0.48584 0.47486 0.46588 0.45797 0.44993 0.44056
## 5 0.52657 0.51194 0.49879 0.48693 0.47670 0.46785 0.45916 0.44957 0.43827
## 6 0.52109 0.50842 0.49673 0.48596 0.47657 0.46788 0.45824 0.44694 0.43373
##      X.250      X.251      X.252      X.253      X.254      X.255      X.256      X.257      X.258
## 1 0.41059 0.39677 0.38416 0.37199 0.36018 0.34883 0.33808 0.32838 0.32001
## 2 0.41788 0.40389 0.39018 0.37691 0.36412 0.35194 0.34082 0.33122 0.32323
## 3 0.43679 0.42174 0.40788 0.39510 0.38335 0.37256 0.36253 0.35353 0.34613
## 4 0.42853 0.41426 0.39958 0.38519 0.37124 0.35833 0.34700 0.33749 0.32981
## 5 0.42443 0.40902 0.39406 0.38027 0.36746 0.35563 0.34515 0.33637 0.32935
## 6 0.41828 0.40202 0.38708 0.37409 0.36264 0.35224 0.34281 0.33475 0.32831
##      X.259      X.260      X.261      X.262      X.263      X.264      X.265      X.266      X.267
## 1 0.31308 0.30753 0.30313 0.29980 0.29748 0.29592 0.29489 0.29440 0.29432
## 2 0.31686 0.31195 0.30806 0.30499 0.30283 0.30149 0.30066 0.30031 0.30030
## 3 0.34026 0.33545 0.33164 0.32891 0.32703 0.32573 0.32493 0.32464 0.32460
## 4 0.32389 0.31947 0.31605 0.31332 0.31139 0.31023 0.30955 0.30928 0.30932
## 5 0.32394 0.31977 0.31649 0.31397 0.31218 0.31108 0.31046 0.31021 0.31026
## 6 0.32331 0.31934 0.31620 0.31389 0.31234 0.31140 0.31080 0.31052 0.31061
##      X.268      y train
## 1 0.29443 100.00      1
## 2 0.30041  80.22      1
## 3 0.32474  79.49      1
## 4 0.30947  60.80      1
## 5 0.31049  59.97      1
## 6 0.31093  60.48      1

```

EXERCISE 4

Let us take a look at data.

```
head(ozone)
```

```
##      Station  Av8stop      Lat      Lon
```

```
## 1      60 7.225806 34.13583 -117.9236
## 2      69 5.899194 34.17611 -118.3153
## 3      72 4.052885 33.82361 -118.1875
## 4      74 7.181452 34.19944 -118.5347
## 5      75 6.076613 34.06694 -117.7514
## 6      84 3.157258 33.92917 -118.2097
```

EXERCISE 5

Let us take a look at data.

```
head(pendigits)
```

```
##      V1 V2 V3 V4 V5 V6 V7 V8 V9 V10 V11 V12 V13 V14 V15 V16 V17 V18
## 1  47 100 27 81 57 37 26  0  0 23  56  53 100  90  40  98  8  1
## 2   0  89 27 100 42 75 29 45 15 15 37  0  69  2 100  6  2  2
## 3   0  57 31 68 72 90 100 100 76 75 50 51 28 25 16  0  1  3
## 4   0 100 7 92 5 68 19 45 86 34 100 45 74 23 67  0  4  6
## 5   0  67 49 83 100 100 81 80 60 60 40 40 33 20 47  0  1  3
## 6 100 100 88 99 49 74 17 47  0 16 37  0 73 16 20 20  6  4
##      V19 V20 V21 V22 V23 V24 V25 V26 V27 V28 V29 V30 V31 V32 V33 V34 V35 V36
## 1  47 100 27 81 57 37 26  0  0 23  56  53 100  90  40  98  8  1
## 2   0  89 27 100 42 75 29 45 15 15 37  0  69  2 100  6  2  2
## 3   0  57 31 68 72 90 100 100 76 75 50 51 28 25 16  0  1  3
## 4   0 100 7 92 5 68 19 45 86 34 100 45 74 23 67  0  4  6
## 5   0  67 49 83 100 100 81 80 60 60 40 40 33 20 47  0  1  3
## 6 100 100 88 99 49 74 17 47  0 16 37  0 73 16 20 20  6  4
```

EXERCISE 6

Let us take a look at data.

```
head(carmarks)
```

```
##      CARMARK ECONOMY SERVICE VALUE PRICE DESIGN SPORT SAFETY EASYINESS
## 1      A100      3.9      2.8  2.2  4.2      3.0  3.1      2.4      2.8
## 2      BMW3      4.8      1.6  1.9  5.0      2.0  2.5      1.6      2.8
## 3      CiAX      3.0      3.8  3.8  2.7      4.0  4.4      4.0      2.6
## 4      Ferr      5.3      2.9  2.2  5.9      1.7  1.1      3.3      4.3
## 5      FiUn      2.1      3.9  4.0  2.6      4.5  4.4      4.4      2.2
## 6      FoFi      2.3      3.1  3.4  2.6      3.2  3.3      3.6      2.8
```

EXERCISE 7

Let us take a look at data.

```
head(scapular)
```

```
##      genus AD.BD AD.CD EA.CD Dx.CD SH.ACR EAD beta gamma      class classdigit
## 1      54 65.56 166.0 50.55 12.80  70.3 115  14  45 Hylobates      1
## 2      54 50.91  93.9 61.82 13.09  75.0 121  20  54 Hylobates      1
## 3      54 46.15  80.8 64.10 11.80  70.0 120  25  61 Hylobates      1
## 4      54 70.29 220.5 50.00 12.75  61.1 113  12  45 Hylobates      1
## 5      54 63.16 144.0 57.89 12.98  64.9 115  14  46 Hylobates      1
```



```
## 6      54 50.72 134.6 56.23 11.88      52.6 136      14      46 Hylobates      1
```

EXERCISE 8

Let us take a look at data.

```
head(scapular)
```

```
##      genus AD.BD AD.CD EA.CD Dx.CD SH.ACR EAD beta gamma      class classdigit
## 1      54 65.56 166.0 50.55 12.80      70.3 115      14      45 Hylobates      1
## 2      54 50.91  93.9 61.82 13.09      75.0 121      20      54 Hylobates      1
## 3      54 46.15  80.8 64.10 11.80      70.0 120      25      61 Hylobates      1
## 4      54 70.29 220.5 50.00 12.75      61.1 113      12      45 Hylobates      1
## 5      54 63.16 144.0 57.89 12.98      64.9 115      14      46 Hylobates      1
## 6      54 50.72 134.6 56.23 11.88      52.6 136      14      46 Hylobates      1
```