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In [79]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from scipy import stats
from scipy.stats import norm
from scipy.stats import linregress
from sklearn.utils import resample
import seaborn as sns
import statsmodels.api as sm
import pylab

In [80]: raw=np.loadtxt('babies.txt', skiprows=1)

In [81]: tai=pd.DataFrame(raw)

In [82]: tai.columns=['bwt', 'gestation', 'parity', 'age', 'height', 'weight', 'smoke']

In [83]: #Detect outliers and delete the data
tai=tai.drop(tai[tai['weight']>=500].index)
tai=tai.drop(tai[tai['smoke']==8].index)
tai=tai.drop(tai[tai['gestation']>=99].index)
tai=tai.drop(tai[tai['height']>=99].index)
tai=tai.drop(tai[tai['age']>=99].index)

In [84]: #Add New column BMI
bmi=tai['weight']/4.45/(tai['height']**2)

In [85]: tai['BMI index']=bmi

In [86]: #Drop BMI index > 18.5
tai=tai[tai['BMI index']>18.5]
tai

Out[87]:

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	bwt	gestation
count	1130.000000	1130.000000
mean	119.620009	279.212099
std	18.346586	16.087643
min	65.000000	148.000000
25%	109.000000	272.000000
50%	120.000000	280.000000
75%	131.000000	288.000000
max	176.000000	353.000000

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In [88]: t=tai.drop(['smoke','parity'],1)

In [89]: cor=t.corr()
sns.heatmap(cor,square=True,annot=True)

Out[89]:

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Pearson correlation formula:
$$r = \frac{N \sum xy - \sum x \sum y}{\sqrt{(N \sum x^2 - (\sum x)^2)(N \sum y^2 - (\sum y)^2)}}$$

```

In [90]: plt.scatter(tai['gestation'],tai['bwt'])
plt.xlabel('gestation')
plt.ylabel('baby birth weight')
#linregress(tai['gestation'],tai['bwt'])
x=np.linspace(280,350)
z=np.slope
bwt_intercept
plt.plot(x,*z,color='black')
plt.savefig('gpc-bwt-jpg')

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In [91]: #Boxplot of relation between smoke and bwt
sns.set(style='whitegrid')

In [92]: tai_c=tai[tai['age']<=40]
tai_c=tai_c[tai_c['age']>=20]

In [93]: #Histogram and distribution simulation of birth weight, smoke=1
smoke=tai_c[tai_c['smoke']==1]
Non_smoke=tai_c[tai_c['smoke']==0]
smoke_bwt=smoke['bwt']
Non_smoke_bwt=Non_smoke['bwt']

In [94]: print(smoke.count())
print(Non_smoke.count())

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```

bwt      400
gestation 400
parity    400
age       400
height    400
weight    400
smoke     400
BMI index 400
dtype: int64

bwt      638
gestation 638
parity    638
age       638
height    638
weight    638
smoke     638
BMI index 638
dtype: int64

In [10]: print('smoke:',stats.mode(smoke_bwt))
print('Non-smoke:',stats.mode(Non_smoke_bwt))

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smoke: ModeResult(mode=array([115.]), count=array([16]))
Non-smoke: ModeResult(mode=array([125.]), count=array([21]))

In [73]: lbw_s=smoke[smoke['bwt']<88.2]['bwt'].count()
bw_s=smoke[smoke['bwt']<88.2]['bwt'].count()

lbw_ns=Non_smoke[Non_smoke['bwt']<88.2]['bwt'].count()
bw_ns=Non_smoke[Non_smoke['bwt']<88.2]['bwt'].count()

q1=[]
q1.append(lbw_s)
q1.append(lbw_ns)

q2=[]
q2.append(bw_s)
q2.append(bw_ns)

width=0.2
index=np.arange(2)

In [75]:

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fig, ax = plt.subplots()
ax.bar(index,q1,width,label='Low birth weight')
ax.bar(index,q2,width,label='Normal birth weight')
ax.set_xticks(index + width / 2)
ax.set_xticklabels(['Smoke', 'Non-Smoke'])
ax.legend()
plt.savefig('Low birth weight vs Normal birth weight')

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In [15]: #Histogram and distribution simulation of birth weight, smoke=0
fig=plt.figure(figsize=(10,10))

ax1=fig.add_subplot(221)
sns.distplot(smoke['bwt'],ax=ax1)
ax1.set_xlabel('birthweight')
ax1.set_title('smoking mother')

ax2=fig.add_subplot(222)
sns.distplot(Non_smoke['bwt'],ax=ax2,color='y')
ax2.set_xlabel('birthweight')
ax2.set_title('Non-smoking mother')

ax3=fig.add_subplot(223)
sns.bboxplot(x='smoke',y='bwt',data=ta1,ax=ax3,hue='smoke')
ax3.set_xlabel('birthweight')

ax4=fig.add_subplot(224)
sns.distplot(smoke_bwt,ax=ax4,label='smoke')
sns.distplot(Non_smoke_bwt,ax=ax4,label='Non-smoker')
ax4.set_xlabel('birthweight')
ax4.legend(loc='upper right')

fig.savefig('Boxplot and histogram')

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In [16]: #Bootstrap calculate kurtosis and skewness
kurt0=[]
skew0=[]
kurt1=[]
skew1=[]
for i in range(1000):
    boot_smoke_bwt=resample(smoke_bwt)
    boot_non_smoke_bwt=resample(Non_smoke_bwt)
    kurt0.append(stats.kurtosis(boot_smoke_bwt,fisher=True))
    skew0.append(stats.skew(boot_smoke_bwt))
    kurt1.append(stats.kurtosis(boot_non_smoke_bwt,fisher=True))
    skew1.append(stats.skew(boot_non_smoke_bwt))

print('kurtosis for smoke: ', np.mean(kurt0),'skewness for smoke: ',np.mean(skew0))
print('kurtosis for Non-smoke: ', np.mean(kurt1),'skewness for Non-smoke: ',np.mean(skew1))

kurtosis for smoke: -0.826788428828976843
skewness for smoke: -0.826935721678656877
kurtosis for Non-smoke: 0.8995389133808468
skewness for Non-smoke: -0.16884335769734115

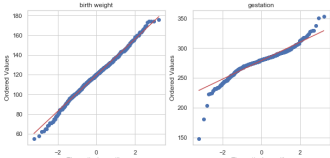
```

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In [19]: #fig=plt.figure(figsize=(10,10))
ax1=fig.add_subplot(221)
stats.probplot(ta1['bwt'],dist=norm,plot=ax1)
ax1.set_title('birth weight')
ax2=fig.add_subplot(222)
stats.probplot(ta1['gestation'],dist=norm,plot=ax2)
ax2.set_title('gestation')

Out[19]: Text(0.5, 1.0, 'gestation')

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In [20]: #The description of two groups
st=smoke.describe()
nst=Non_smoke.describe()

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In [21]: #Divide gestation group into 4 groups and compare the influence of smoke
#over birth weight
minimum=int(stat['gestation']['min'])
q1=int(stat['gestation']['Q1'])
mean=int(stat['gestation']['mean'])
q3=int(stat['gestation']['Q3'])
maximum=int(stat['gestation']['max'])

g1=ta1[ta1['gestation']<=q1]
g2=ta1[ta1['gestation']>q1]
g2=ta1[ta1['gestation']<=mean]
g3=ta1[ta1['gestation']>mean]
g4=ta1[ta1['gestation']>q3]

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In [22]: #And we form 4 groups. For each group, we separate again the data to smoke and non-smoke
s1=g1[g1['smoke']==1]
ns1=g1[g1['smoke']==0]
s2=g2[g2['smoke']==1]
ns2=g2[g2['smoke']==0]
s3=g3[g3['smoke']==1]
ns3=g3[g3['smoke']==0]
s4=g4[g4['smoke']==1]
ns4=g4[g4['smoke']==0]

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In [23]: #fig=plt.figure(figsize=(10,10))
ax1=fig.add_subplot(221)
sns.distplot(s1['bwt'],ax=ax1,label='smoke')
sns.distplot(ns1['bwt'],ax=ax1,label='non-smoke')
ax1.set_title('Group 1')
ax1.legend()

ax2=fig.add_subplot(222)
sns.distplot(s2['bwt'],ax=ax2,label='smoke')
sns.distplot(ns2['bwt'],ax=ax2,label='non-smoke')
ax2.set_title('Group 2')
ax2.legend()

ax3=fig.add_subplot(223)
sns.distplot(s3['bwt'],ax=ax3,label='smoke')
sns.distplot(ns3['bwt'],ax=ax3,label='non-smoke')
ax3.set_title('Group 3')
ax3.legend(loc='upper right')

ax4=fig.add_subplot(224)
sns.distplot(s4['bwt'],ax=ax4,label='smoke')
sns.distplot(ns4['bwt'],ax=ax4,label='non-smoke')
ax4.set_title('Group 4')
ax4.legend()

plt.savefig('Distribution over 4 groups.jpg')

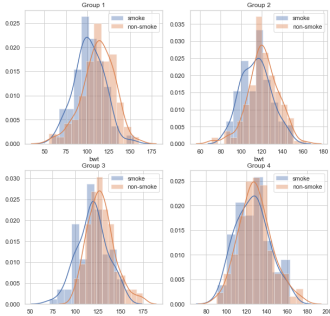
```

C:\Users\h131jan\Anaconda3\lib\site-packages\scipy\stats\stats.py:1713: FutureWarning: Using a non-tuple sequence for multidimensional indexing is deprecated; use 'arr[tuple(seq)]' instead of 'arr[seq]'. In the future this will be interpreted as an array index, 'arr[np.array(seq)]', which will result either in an error or a different result.

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return np.add.reduce(sorted[indices] * weights, axis=axis) / sumval

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In [24]: #Define a function to bootstrap data and calculate the skewness and kurtosis
#Define normalize function
def sta(x,ns):
    kurt_s=[]
    skew_s=[]
    kurt_ns=[]
    skew_ns=[]
    group_ns=[]
    for i in range(1000):
        x_s=resample(x['bwt'])
        x_ns=resample(x['bwt'])
        kurt_s.append(stats.kurtosis(x_s))
        kurt_ns.append(stats.kurtosis(x_ns))
        skew_s.append(stats.skew(x_s))
        skew_ns.append(stats.skew(x_ns))
    return pd.DataFrame([['kurtosis for smoke:', np.mean(kurt_s),'skewness for smoke:',np.mean(skew_s)],
                        ['kurtosis for Non-smoke:', np.mean(kurt_ns),'skewness for Non-smoke:',np.mean(skew_ns)]])
def normalize(x):
    return (x-np.mean(x))/np.std(x)

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In [25]: #Scroup1
sta(st,ns1)

Out[25]:

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	0	1	2	3
0	kurtosis for smoke: 0.057575	skewness for smoke: 0.037529		
1	kurtosis for Non-smoker: 0.357095	skewness for Non-smoker: -0.424144		

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In [26]: stats.kstest(normalize(st['bwt']), 'norm')

Out[26]: KstestResult(statistic=0.09195875619612467, pvalue=0.88567175693218855)

In [27]: stats.kstest(normalize(ns1['bwt']), 'norm')

Out[27]: KstestResult(statistic=0.06914297887844378, pvalue=0.4138208394658598)

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In [38]: #perform bootstrap
bt=[]
for i in range(1000):
    new1=resample(s1['but'])
    new2=resample(ns1['but'])
    f=stats.ttest_ind(new1,new2,equal_var=False)[1]
    bt.append(f)

In [31]: print('The mean p-value is: ', np.mean(bt))

The mean p-value is: 0.0006985700770165726

In [67]: #H0group?
sta(s2,ns2)

Out[67]:
      0      1      2      3
0  kurtosis for smoke: -0.459983  skewness for smoke: 0.151920
1  kurtosis for Non-smoke: 0.696749  skewness for Non-smoke: -0.309957

In [69]: stats.kstest(normalize(s2['but']), 'norm')

Out[69]: KstestResult(statistic=0.07160879598278153, pvalue=0.529534536559876)

In [70]: stats.kstest(normalize(ns2['but']), 'norm')

Out[70]: KstestResult(statistic=0.05850244374565028, pvalue=0.5978956437272418)

In [32]: bt2=[]
for i in range(1000):
    new1=resample(s2['but'])
    new2=resample(ns2['but'])
    f=stats.ttest_ind(new1,new2,equal_var=False)[1]
    bt2.append(f)

In [33]: print('The mean p-value is: ', np.mean(bt2))

The mean p-value is: 0.0059093854720566955

In [128]: #H0group?
sta(s3,ns3)

Out[128]:
      0      1      2      3
0  kurtosis for smoke: -0.176585  skewness for smoke: -0.124142
1  kurtosis for Non-smoke: 0.424900  skewness for Non-smoke: 0.509720

In [79]: print(stats.kstest(normalize(s3['but']), 'norm'))
print(stats.kstest(normalize(ns3['but']), 'norm'))

KstestResult(statistic=0.060510232897182326, pvalue=0.853532474725926)
KstestResult(statistic=0.066253176084308209, pvalue=0.1074800406720906)

In [34]: bt3=[]
for i in range(1000):
    new1=resample(s3['but'])
    new2=resample(ns3['but'])
    f=stats.ttest_ind(new1,new2,equal_var=False)[1]
    bt3.append(f)

In [35]: print('The mean p-value is: ', np.mean(bt3))

The mean p-value is: 0.0006449392814525881

In [129]: #H0group?
sta(s4,ns4)

Out[129]:
      0      1      2      3
0  kurtosis for smoke: -0.438254  skewness for smoke: 0.255058
1  kurtosis for Non-smoke: 0.122453  skewness for Non-smoke: 0.349934

In [82]: print(stats.kstest(normalize(s4['but']), 'norm'))
print(stats.kstest(normalize(ns4['but']), 'norm'))

KstestResult(statistic=0.05580091386345749, pvalue=0.9515974993124998)
KstestResult(statistic=0.05190208287427051, pvalue=0.7347264120079529)

In [36]: bt4=[]
for i in range(1000):
    new1=resample(s4['but'])
    new2=resample(ns4['but'])
    f=stats.ttest_ind(new1,new2,equal_var=False)[1]
    bt4.append(f)

In [37]: print('The mean p-value is: ', np.mean(bt4))

The mean p-value is: 0.268468187550974

In [27]: #Outcome?
#emp_logdetxt('baby123.txt',skiprows=1)

In [28]: ta2=pd.DataFrame(rw)
ta2.columns=['id', 'plurality', 'outcome', 'date', 'gestation', 'sex', 'wt', 'parity',
              'race', 'age', 'ed', 'ht', 'wgt', 'drac', 'dage', 'ded', 'dht', 'dwt', 'marital',
              'inc', 'smoke', 'time', 'number']

In [54]: ta2=ta2.drop(ta2[ta2['race']==99].index)
ta2=ta2.drop(ta2[ta2['gestation']==999].index)
ta2=ta2.drop(ta2[ta2['age']==99].index)
ta2=ta2.drop(ta2[ta2['drac']==99].index)
ta2=ta2.drop(ta2[ta2['dage']==99].index)
ta2=ta2.drop(ta2[ta2['ded']==99].index)
ta2=ta2.drop(ta2[ta2['dht']==99].index)
ta2=ta2.drop(ta2[ta2['dwt']==99].index)
ta2=ta2.drop(ta2[ta2['inc']==98].index)
ta2=ta2.drop(ta2[ta2['smoke']==9].index)
ta2=ta2.drop(ta2[ta2['time']==9].index)
ta2=ta2.drop(ta2[ta2['drac']==98].index)
ta2=ta2.drop(ta2[ta2['dht']==9].index)
ta2=ta2.drop(ta2[ta2['ed']==9].index)

In [43]: ta2=ta2.drop(ta2[ta2['smoke']==9].index)
ta2=ta2.drop(ta2[ta2['gestation']==999].index)

In [44]: ns2=ta2[ta2['smoke']==0]

In [47]: #rui contains biological info of babies
#rui contains mother's info
r0=Non_smoke['gestation']
r1=ns2[ns2['smoke']==1]['gestation']
r2=ns2[ns2['smoke']==2]['gestation']
r3=ns2[ns2['smoke']==3]['gestation']

data=[r0,r1,r2,r3]
fig7,ax7=plt.subplots()
ax7.set_title('Smoke vs Non-smoke boxplot')
ax7.boxplot(data,showfliers=False)
plt.xticks([1, 2, 3, 4], ['1','2','3'])

Out[47]: [(,
  <matplotlib.axis.XTick at 0x24f71df3cf0>,
  <matplotlib.axis.XTick at 0x24f71df3a50>,
  <matplotlib.axis.XTick at 0x24f71df3400>),
  <a list of 4 Text ticklabel objects>]

Smoke vs Non-smoke boxplot
250 300
280 290 300 310
0 1 2 3

In [46]: sns.boxplot(x='smoke',y='gestation',data=ns2)

Out[46]: <matplotlib.axes._subplots.AxesSubplot at 0x24f73cc00f0>

Smoke vs Non-smoke boxplot
340 320 300 280 260
10 20 30
smoke

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964 7689      5      1 1485      999  1 107      1  0 19  1
60 118      5 22  1
972 7711      5      1 1507      999  1 136      13  0 36  2
66 135      0 39  5
1193 8499      5      1 1680      999  1 124      0  7 39  2
65 228      7 38  1

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      dht dwt marital inc smoke time number      BMI
4      68 197      1  8      3  5      5 18.601134
90      74 170      1  6      1  1      5 18.284696
94      99 999      1  5      3  9      1 14.756070
99      71 168      1  9      3  3      1 23.975026
155     99 999      1  4      1  1      9  9.476584
243     99 999      1  4      1  1      3 20.136054
651     70 180      1  5      0  0      0 18.840830
707     99 999      1  3      1  1      2 20.680272
740     66 160      1  1      0  0      0 12.480469
880     67 156      1  0      0  0      0 24.961032
964     99 999      1  1      0  0      0 21.400000
972     72 185      1  7      0  0      0 22.479339
1193    70 220      1  4      0  0      0 21.131361

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> normal.index <- which (Data2$gestation != 999)
> Data2.normal <- Data2[normal.index,]
> boxplot(gestation~smoke, Data2.normal)
>
> #subset smoker/nonsmoker
> Data2.smoke <- subset(Data2.normal, Data2.normal$smoke == 1)
> Data2.nonsmoke <- subset(Data2.normal,
+                           (Data2.normal$smoke == 0 |
+                           Data2.normal$smoke == 2 |
+                           Data2.normal$smoke == 3 ))
> Data2.both <- subset(Data2.normal, (Data2.normal$smoke != 9))
> #total number of sample size without outliers/smoke=9
> nrow(Data2.smoke)+nrow(Data2.nonsmoke)
[1] 1205
>
> #we first want to find the mean and variance of gestation age
> summary(Data2.smoke$gestation)
  Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
223.0  271.0   279.0   277.9   286.0   330.0
> summary(Data2.nonsmoke$gestation)
  Min. 1st Qu.  Median    Mean 3rd Qu.    Max.

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148.0  273.0  281.0  280.1  289.0  353.0
> var(Data2.smoke$gestation)
[1] 226.7328
> var(Data2.nonsmoke$gestation)
[1] 277.5445
> sd(Data2.smoke$gestation)
[1] 15.05765
> sd(Data2.nonsmoke$gestation)
[1] 16.65967
>
> #we then find the histogram of gestation age in order to find the
distribution
> hist(Data2.smoke$gestation,
+       breaks = 100,
+       probability = TRUE,
+       col = rgb(1, 0, 0, 0.5),
+       xlab = "Gestational Age",
+       main = "Histogram of Gestational Age")
>
> lines(density(Data2.smoke$gestation), col = "red", lwd = 2)
>
> hist(Data2.nonsmoke$gestation,
+       breaks = 100,
+       probability = TRUE,
+       col = rgb(0, 0, 1, 0.5),
+       add = T)
>
> lines(density(Data2.nonsmoke$gestation), col = "blue", lwd = 2)
> legend("topright", c("smoke", "nonsmoke"), fill = c("red", "blue"))
>
>
>
> #boxplot
> boxplot(gestation~smoke, Data2.both,
+         xlab = "smoke level",
+         ylab = "gestational age",
+         main = "Boxplot of Gestational Age for Different Smoke Level",
+         cex.main = 1,
+         col = c("darkred", "#E69F00", "#56B4E9", "yellow"))
>
>

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>
> #frequency of different weeks
> Data2.smoke.35weeks <- subset(Data2.smoke,
+                               Data2.smoke$gestation < 245)
> nrow(Data2.smoke.35weeks)
[1] 14
> Data2.smoke.36weeks <- subset(Data2.smoke,
+                               Data2.smoke$gestation < 252)
> nrow(Data2.smoke.36weeks)
[1] 25
> Data2.smoke.37weeks <- subset(Data2.smoke,
+                               Data2.smoke$gestation < 259)
> nrow(Data2.smoke.37weeks)
[1] 41
>
> Data2.smoke.38weeks <- subset(Data2.smoke,
+                               Data2.smoke$gestation < 266)
> nrow(Data2.smoke.38weeks)
[1] 66
> Data2.smoke.39weeks <- subset(Data2.smoke,
+                               Data2.smoke$gestation < 273)
> nrow(Data2.smoke.39weeks)
[1] 141
>
> #non smoker
> Data2.nonsmoke.35weeks <- subset(Data2.nonsmoke,
+                                  Data2.nonsmoke$gestation < 245)
> nrow(Data2.nonsmoke.35weeks)
[1] 19
> Data2.nonsmoke.36weeks <- subset(Data2.nonsmoke,
+                                  Data2.nonsmoke$gestation < 252)
> nrow(Data2.nonsmoke.36weeks)
[1] 34
> Data2.nonsmoke.37weeks <- subset(Data2.nonsmoke,
+                                  Data2.nonsmoke$gestation < 259)
> nrow(Data2.nonsmoke.37weeks)
[1] 56
>
> Data2.nonsmoke.38weeks <- subset(Data2.nonsmoke,
+                                  Data2.nonsmoke$gestation < 266)
> nrow(Data2.nonsmoke.38weeks)

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[1] 86
> Data2.nonsmoke.39weeks <- subset(Data2.nonsmoke,
+                                Data2.nonsmoke$gestation < 273)
> nrow(Data2.nonsmoke.39weeks)
[1] 168
>
> #qqplot
> qqnorm(Data2.smoke$gestation, main = "Normal Q-Q Plot of smoking
group")
> qqline(Data2.smoke$gestation)
> qqnorm(Data2.nonsmoke$gestation,
+        main = "Normal Q-Q Plot of Nonsmoking Group")
> qqline(Data2.nonsmoke$gestation)
> qqplot(Data2.smoke$gestation, Data2.nonsmoke$gestation,
+        main = "Quantile-Quantile Plot Comparison for two distribution",
+        font.main = 8, cex.main = 1,
+        xlab = "smoking group",
+        ylab = "nonsmoking group")
> abline (c(0,1))
>
> #caculation of t-test
> mean(Data2.nonsmoke$gestation)-mean(Data2.smoke$gestation)
[1] 2.173941
> sqrt(((sd(Data2.nonsmoke$gestation))^2/nrow(Data2.nonsmoke))+
+      ((sd(Data2.smoke$gestation))^2/nrow(Data2.smoke)))
[1] 0.9255124
> t = 2.173941/0.9255124
>
>
>
>
> #calculate the kurtosis and skewness
> kurtosis(Data2.smoke$gestation)
[1] 5.083699
> kurtosis(Data2.nonsmoke$gestation)
[1] 11.73105
> skewness(Data2.smoke$gestation)
[1] -0.2293753
> skewness(Data2.nonsmoke$gestation)
[1] -1.071203

```