

# **ASSIGNMENT OF DTNS**



## **DEPARTMENT OF STATISTICS**

**SUBMITTED BY:**

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## **RUN TEST**

**Q.1- The win-loss record of a certain basketball team for their 50 consecutive games was as follows :-**

**6W L 6W L W L 3W 2L 4W L 3W 2L 6W 2L 2W 3L 2W L 3W**

**Apply the run test to test that the sequence of win and loss is random.**

**Solution:-**

Here we set the null and alternative hypothesis as:

Ho: The sequence of win and loss is random

H1: The sequence of win and loss is not random.

**Observation:-**

<b>n1</b>	36
<b>n2</b>	14
<b>R</b>	19
<b><math>\mu</math></b>	21.16
<b><math>\sigma^2</math></b>	7.882971429
<b><math>\sigma</math></b>	2.807662983
<b>Zo</b>	-0.769323103
<b> Zo </b>	0.769323103
<b>F( Zo )</b>	0.779149237
<b>p*</b>	<b>0.441701526</b>
<b><math>\alpha</math></b>	<b>0.05</b>

**Conclusion:-**

Here, we can conclude that the value of p\* is greater than the value of  $\alpha$  ( $p^* > \alpha$ ). So, we accept the null hypothesis and we can say that the sequence of win and loss is random.

## SIGN TEST

**Q-2. The minimum temperature of a place has been recorded for fourteen successive days. Test if the successive changes (positive and negative) in the minimum temperatures are random.**

Day	1	2	3	4	5	6	7	8	9	10	11	12	13	14
Temperature(°C)	20.1	20.3	20.1	20.4	20.5	20.6	20.4	20.3	20.7	20.7	21.0	21.6	21.5	21.0

### **Solution:-**

Here we set the null and alternative hypothesis as

$H_0$ : The successive positive and negative changes in minimum temperatures are random.

$H_1$ : The successive positive and negative changes in minimum temperatures are not random.

We find the following positive and negative changes in temperatures with respect to day 1:

Day	1	2	3	4	5	6	7	8	9	10	11	12	13	14
Temperature(°C)	20.1	20.3	20.1	20.4	20.5	20.6	20.4	20.3	20.7	20.7	21.0	21.6	21.5	21.0
Difference		+	-	+	+	+	-	-	+	0	+	+	-	-

### **Observation:-**

<b>n</b>	12
<b>n1</b>	7
<b>n2</b>	5
<b>R</b>	6
<b>k</b>	3
<b>p*</b>	0.22727
<b><math>\alpha</math></b>	0.05

### **Conclusion:-**

Here, we can conclude that the value of  $p^*$  is greater than the value of  $\alpha$  ( $p^* > \alpha$ ). So, we accept the null hypothesis and we can say that the successive positive and negative changes in minimum temperatures are random.

### **SIGN TEST FOR LARGE SAMPLE**

**Q-3. Let us consider motor trained car road test data which consist 32 observations & 11 variables by using the dataset “mtcars” in R. Perform a sign test to test mileage (mileage per gallon) varies from its median mpg or not.**

**Solution:-**

**$H_0$ :** The population median is = 19.2

**$H_1$ :** The population median is not equal to 19.2

> x=mtcars

> x

	mpg	cyl	disp	hp	drat	wt	qsec	vs	am	gear	carb
Mazda RX4	21.0	6	160.0	110	3.90	2.620	16.46	0	1	4	4
Mazda RX4 Wag	21.0	6	160.0	110	3.90	2.875	17.02	0	1	4	4
Datsun 710	22.8	4	108.0	93	3.85	2.320	18.61	1	1	4	1
Hornet 4 Drive	21.4	6	258.0	110	3.08	3.215	19.44	1	0	3	1
Hornet Sportabout	18.7	8	360.0	175	3.15	3.440	17.02	0	0	3	2
Valiant	18.1	6	225.0	105	2.76	3.460	20.22	1	0	3	1
Duster 360	14.3	8	360.0	245	3.21	3.570	15.84	0	0	3	4
Merc 240D	24.4	4	146.7	62	3.69	3.190	20.00	1	0	4	2
Merc 230	22.8	4	140.8	95	3.92	3.150	22.90	1	0	4	2
Merc 280	19.2	6	167.6	123	3.92	3.440	18.30	1	0	4	4
Merc 280C	17.8	6	167.6	123	3.92	3.440	18.90	1	0	4	4
Merc 450SE	16.4	8	275.8	180	3.07	4.070	17.40	0	0	3	3
Merc 450SL	17.3	8	275.8	180	3.07	3.730	17.60	0	0	3	3
Merc 450SLC	15.2	8	275.8	180	3.07	3.780	18.00	0	0	3	3
Cadillac Fleetwood	10.4	8	472.0	205	2.93	5.250	17.98	0	0	3	4
Lincoln Continental	10.4	8	460.0	215	3.00	5.424	17.82	0	0	3	4
Chrysler Imperial	14.7	8	440.0	230	3.23	5.345	17.42	0	0	3	4
Fiat 128	32.4	4	78.7	66	4.08	2.200	19.47	1	1	4	1
Honda Civic	30.4	4	75.7	52	4.93	1.615	18.52	1	1	4	2
Toyota Corolla	33.9	4	71.1	65	4.22	1.835	19.90	1	1	4	1

```

Toyota Corona      21.5  4 120.1 97 3.70 2.465 20.01 1 0  3  1
Dodge Challenger   15.5  8 318.0 150 2.76 3.520 16.87 0 0  3  2
AMC Javelin        15.2  8 304.0 150 3.15 3.435 17.30 0 0  3  2
Camaro Z28         13.3  8 350.0 245 3.73 3.840 15.41 0 0  3  4
Pontiac Firebird   19.2  8 400.0 175 3.08 3.845 17.05 0 0  3  2
Fiat X1-9          27.3  4  79.0  66 4.08 1.935 18.90 1 1  4  1
Porsche 914-2      26.0  4 120.3  91 4.43 2.140 16.70 0 1  5  2
Lotus Europa       30.4  4  95.1 113 3.77 1.513 16.90 1 1  5  2
Ford Pantera L     15.8  8 351.0 264 4.22 3.170 14.50 0 1  5  4
Ferrari Dino       19.7  6 145.0 175 3.62 2.770 15.50 0 1  5  6
Maserati Bora      15.0  8 301.0 335 3.54 3.570 14.60 0 1  5  8
Volvo 142E         21.4  4 121.0 109 4.11 2.780 18.60 1 1  4  2
> length(x)
[1] 11
> y=x[,1]
> y
[1] 21.0 21.0 22.8 21.4 18.7 18.1 14.3 24.4 22.8 19.2 17.8 16.4 17.3 15.2 10.4
10.4 14.7
[18] 32.4 30.4 33.9 21.5 15.5 15.2 13.3 19.2 27.3 26.0 30.4 15.8 19.7 15.0 21.4
> median(y)
[1] 19.2
> z=y-median(y)
> z
[1]  1.8  1.8  3.6  2.2 -0.5 -1.1 -4.9  5.2  3.6  0.0 -1.4 -2.8 -1.9 -4.0 -8.8 -8.8 -4.5
[18] 13.2 11.2 14.7  2.3 -3.7 -4.0 -5.9  0.0  8.1  6.8 11.2 -3.4  0.5 -4.2  2.2
> which(y>19.2)
[1]  1  2  3  4  8  9 18 19 20 21 26 27 28 30 32
> q=which(y>19.2)
> length(q)
[1] 15
> w=which(y<19.2)
> length(w)
[1] 15
> r=which(y==19.2)
> length(r)
[1] 2
>
q=choose(30,0)+choose(30,1)+choose(30,2)+choose(30,3)+choose(30,4)+choos
e(30,5)+choose(30,6)+choose(30,7)+choose(30,8)+choose(30,9)+choose(30,10)
+choose(30,11)+choose(30,12)+choose(30,13)+choose(30,14)+choose(30,15)

```

```
> q
[1] 614429672
> l=2^30
> l
[1] 1073741824
> p=q/l
> p
[1] 0.5722322
```

**By 2<sup>nd</sup> method,**

```
> signtest(y,m=median(y),alternative="less")
```

Large Sample Approximation for the Sign Test

$H_0$ : The population median is = 19.2

$H_1$ : The population median is less than 19.2

$B = 15$

Significance Level = 0.05

**The p-value is 0.572433929707647**

There is not enough evidence to conclude that the population median is less than 19.2 at a significance level of 0.05

```
> signtest(y,m=median(y),alternative="greater")
```

Large Sample Approximation for the Sign Test

$H_0$ : The population median is = 19.2

$H_A$ : The population median is greater than 19.2

$B = 15$

Significance Level = 0.05

**The p-value is 0.572433929707647**

There is not enough evidence to conclude that the population median is greater than 19.2 at a significance level of 0.05

```
> signtest(y,m=median(y))
```

Large Sample Approximation for the Sign Test

$B = 15$

Significance Level = 0.05

**The p-value is 0.855132140584706**

**Conclusion:**

There is not enough evidence to conclude that the population median is different than 19.2 at a significance level of 0.05

**SIGN TEST FOR TWO SAMPLE**

**Q-4 A random sample of size 10 of diabetic patients is selected at random to see the effect of reducing blood sugar level of a new drug their fasting blood sugar level are measured before and after using the drug for 3 months. Perform a test at 1% level of significance to see if the drug is effective. given that  $S_{(L,\alpha)}=0$  for  $n=2$  and  $\alpha=0.01$ .**

**Solution:-**

**$H_0$ :After 3 month the drug is effective**

**$H_1$ :After 3 month the drug is not effective**

Time.1<-c(125,132,137,128,140,160,155,145,132,152)

Time.2<-c(110,135,125,132,140,142,148,127,132,140)

SIGN.test(x = Time.1,

+ y = Time.2,

+ alternative = "two.sided",

+ conf.level = 0.99)

Dependent-samples Sign-Test

data: Time.1 and Time.2

S = 6, p-value = 0.2891

alternative hypothesis: true median difference is not equal to 0

99 percent confidence interval:

-3.588 18.000

sample estimates:

median of x-y

9.5

Achieved and Interpolated Confidence Intervals:

Conf.Level L.E.pt U.E.pt

Lower Achieved CI 0.9785 -3.000 18

Interpolated CI 0.9900 -3.588 18

Upper Achieved CI 0.9980 -4.000 18

### Conclusion:

After compare with p value we conclude that after 3 month the drug is effective

### WILCOXON SIGN RANK TEST

**Q-5: In a certain experiment to compare two types of animal foods A and B the following results of increase in weights were observed in animals:**

Animal number		1	2	3	4	5	6	7	8	Total
Increase weight in lb	Food A	49	53	51	52	47	50	52	53	407
	Food B	52	55	52	53	50	54	54	53	423

( I ) Assuming that the two samples of animals are independent, can we conclude that food B is better than food A?

### Solution:

**$H_0$ : Food B is better than A**

**$H_1$ : Food B is not better than A**

**> x=c(49,53,51,52,47,50,52,53)**

**> x**

**[1] 49 53 51 52 47 50 52 53**

**> y=c(52,55,52,53,50,54,54,53)**

**> y**

**[1] 52 55 52 53 50 54 54 53**

**> z=x-y**

**> z**

**[1] -3 -2 -1 -1 -3 -4 -2 0**

**> wilcox.test(x,y,paired=T)**



## Wilcoxon signed rank test with continuity correction

data: x and y

**V = 0, p-value = 0.02178**

alternative hypothesis: true location shift is not equal to 0

Warning messages:

1: In wilcox.test.default(x, y, paired = T) :

cannot compute exact p-value with ties

2: In wilcox.test.default(x, y, paired = T) :

cannot compute exact p-value with zeroes

### **Conclusion:**

Here the p value is less than significance level so that the  $H_0$  hypothesis is accept.

## WILCOXON SIGN RANK TEST FOR SMALL SAMPLE

**Q-6.**Let us consider the following Data which shows IQ score of 15 employs selected at

random given as :

**99 ,100 ,90 ,94,135,108,107,111,119,104,127,109,117,105,125.**

**Solution:**

**H<sub>0</sub>:**The data of IQ score of employs selected at random

**H<sub>1</sub>:**The data of IQ score of employs not selected at random

```
> x=c(99,100,90,94,135,108,107,111,119,104,127,109,117,105,125) #assigned vector
> median(x)
[1] 108
> y=x-median(x)
> y
[1] -9 -8 -18 -14 27 0 -1 3 11 -4 19 1 9 -3 17
> 3
[1] 3
> a=which(y>0) #extract the y>0 values
> b=which(y<0) #extract the y<0 values
> a
[1] 5 8 9 11 12 13 15
> b
[1] 1 2 3 4 7 10 14
> length(a)
[1] 7
> length(b)
[1] 7

>
> t.test(x,mu=107)
```

One Sample t-test

data: x  
t = 0.92227, df = 14, p-value = 0.372  
alternative hypothesis: true mean is not equal to 107  
95 percent confidence interval:  
103.0234 116.9766  
sample estimates:  
mean of x  
110

```
> wilcox.test(x,mu=107,alternative ="less")
```

Wilcoxon signed rank test with continuity correction

data: x  
V = 64.5, p-value = 0.7837  
alternative hypothesis: true location is less than 107

Warning messages:

1: In wilcox.test.default(x, mu = 107, alternative = "less") :  
cannot compute exact p-value with ties  
2: In wilcox.test.default(x, mu = 107, alternative = "less") :  
cannot compute exact p-value with zeroes

```
> t.test(x,mu=107,alternative = "g")
```

One Sample t-test

data: x  
t = 0.92227, df = 14, p-value = 0.186  
alternative hypothesis: true mean is greater than 107  
95 percent confidence interval:  
104.2707 Inf  
sample estimates:  
mean of x  
110

**Conclusion:**

Here the value of p is greater than significance value so that the alternative

Hypothesis is accept so that here the data is not selected randomly.

### **WILCOXON TEST FOR LARGE SAMPLE**

**Q-7** Let us consider the following Data which shows IQ score of 15 employs selected at random given

**as:**99,100,90,94,135,108,107,111,119,104,127,109,117,105,125,98,112,85,92,140,111,115,137,117,123,132,83,120,82,106,147,110

**Solution:**

**H<sub>0</sub>:** The data of IQ score of employs selected at random

**H<sub>1</sub>:** The data of IQ score of employs not selected at random

```
>f=c(99,100,90,94,135,108,107,111,119,104,127,109,117,105,125,98,112,85,92,140,111,115,137,117,123,132,83,120,82,106,147,110)
```

```
> f
```

```
[1] 99 100 90 94 135 108 107 111 119 104 127 109 117 105 125 98 112 85 92
```

```
[20] 140 111 115 137 117 123 132 83 120 82 106 147 110
```

```
> length(f)
```

```
[1] 32
```

```
> x=median(f)
```

```
> x
```

```
[1] 110.5
```

```
> q=which(f>110.5)
```

```
> q
```

```
[1] 5 8 9 11 13 15 17 20 21 22 23 24 25 26 28 31
```

```
> length(q)
```

```
[1] 16
```

```
> p=which(f<110.5)
```

```
> p
```

```
[1] 1 2 3 4 6 7 10 12 14 16 18 19 27 29 30 32
```

```
> length(p)
```

```
[1] 16
```

```
> t=which(f==110.5)
```

```

> t
integer(0)
> length(t)
[1] 0
> sum_q=sum(q)
> sum_q
[1] 298
> n=32
> E_q=((n*(n+1))/4)
> E_q
[1] 264
> var_q=((n*(n+1)*(2*n+1))/24)
> var_q
[1] 2860
> sqrt_var_q=sqrt(var_q)
> sqrt_var_q
[1] 53.47897
> Zo=((sum_q-E_q)/sqrt_var_q)
> Zo
[1] 0.635764
> pnorm(Zo)
[1] 0.7375349
> P*=2*(1-pnorm(Zo))
> P*
[1] 0.5249303

```

### **Conclusion:**

Here p value is greater than significance value so that the  $H_0$  is rejected so the data are not selected randomly.

### WILCOXON SIGN RANK TEST

**Q-8 Apply the Wilcoxon Signed rank test for the “ToothGrowth” dataset in R, (which contain length (len), supplement (supp) and dose) to test significance difference of supplements for the growing length of the following:**

- 1. For the supplement “VC”,**
  - a. Dose 0.5 is effective to dose 1 or not**
  - b. Dose 1 is effective to dose 1.5 or not**
  - c. Dose 0.5 is effective to dose 1.5 or not**
- 2. For the supplement “OJ”,**
  - d. Dose 0.5 is effective to dose 1 or not**
  - e. Dose 1 is effective to dose 1.5 or not**
  - f. Dose 0.5 is effective to dose 1.5 or not**

**x=ToothGrowth**

**> x**

	len	supp	dose
1	4.2	VC	0.5
2	11.5	VC	0.5
3	7.3	VC	0.5
4	5.8	VC	0.5
5	6.4	VC	0.5
6	10.0	VC	0.5
7	11.2	VC	0.5
8	11.2	VC	0.5
9	5.2	VC	0.5
10	7.0	VC	0.5
11	16.5	VC	1.0
12	16.5	VC	1.0

13	15.2	VC	1.0
14	17.3	VC	1.0
15	22.5	VC	1.0
16	17.3	VC	1.0
17	13.6	VC	1.0
18	14.5	VC	1.0
19	18.8	VC	1.0
20	15.5	VC	1.0
21	23.6	VC	2.0
22	18.5	VC	2.0
23	33.9	VC	2.0
24	25.5	VC	2.0
25	26.4	VC	2.0
26	32.5	VC	2.0
27	26.7	VC	2.0
28	21.5	VC	2.0
29	23.3	VC	2.0
30	29.5	VC	2.0
31	15.2	OJ	0.5
32	21.5	OJ	0.5
33	17.6	OJ	0.5
34	9.7	OJ	0.5
35	14.5	OJ	0.5
36	10.0	OJ	0.5
37	8.2	OJ	0.5
38	9.4	OJ	0.5
39	16.5	OJ	0.5
40	9.7	OJ	0.5
41	19.7	OJ	1.0
42	23.3	OJ	1.0
43	23.6	OJ	1.0
44	26.4	OJ	1.0
45	20.0	OJ	1.0
46	25.2	OJ	1.0
47	25.8	OJ	1.0
48	21.2	OJ	1.0
49	14.5	OJ	1.0
50	27.3	OJ	1.0
51	25.5	OJ	2.0
52	26.4	OJ	2.0

53 22.4 OJ 2.0

54 24.5 OJ 2.0

55 24.8 OJ 2.0

56 30.9 OJ 2.0

57 26.4 OJ 2.0

58 27.3 OJ 2.0

59 29.4 OJ 2.0

60 23.0 OJ 2.0

> y<-subset(x,supp=="VC")

> y

len supp dose

1 4.2 VC 0.5

2 11.5 VC 0.5

3 7.3 VC 0.5

4 5.8 VC 0.5

5 6.4 VC 0.5

6 10.0 VC 0.5

7 11.2 VC 0.5

8 11.2 VC 0.5

9 5.2 VC 0.5

10 7.0 VC 0.5

11 16.5 VC 1.0

12 16.5 VC 1.0

13 15.2 VC 1.0

14 17.3 VC 1.0

15 22.5 VC 1.0

16 17.3 VC 1.0

17 13.6 VC 1.0

18 14.5 VC 1.0

19 18.8 VC 1.0

20 15.5 VC 1.0

21 23.6 VC 2.0

22 18.5 VC 2.0

23 33.9 VC 2.0

24 25.5 VC 2.0

25 26.4 VC 2.0

26 32.5 VC 2.0

27 26.7 VC 2.0

28 21.5 VC 2.0

29 23.3 VC 2.0



```
30 29.5 VC 2.0
> d1<-subset(y,dose==0.5)
> d1
```

```
  len supp dose
1  4.2  VC 0.5
2 11.5  VC 0.5
3  7.3  VC 0.5
4  5.8  VC 0.5
5  6.4  VC 0.5
6 10.0  VC 0.5
7 11.2  VC 0.5
8 11.2  VC 0.5
9  5.2  VC 0.5
10 7.0  VC 0.5
```

```
> d2<-subset(y,dose==1)
> d2
```

```
  len supp dose
11 16.5  VC  1
12 16.5  VC  1
13 15.2  VC  1
14 17.3  VC  1
15 22.5  VC  1
16 17.3  VC  1
17 13.6  VC  1
18 14.5  VC  1
19 18.8  VC  1
20 15.5  VC  1
```

```
> d3<-subset(y,dose==2)
> d3
```

```
  len supp dose
21 23.6  VC  2
22 18.5  VC  2
23 33.9  VC  2
24 25.5  VC  2
25 26.4  VC  2
26 32.5  VC  2
27 26.7  VC  2
28 21.5  VC  2
29 23.3  VC  2
30 29.5  VC  2
```

```

> l1<-d1$len
> l1
[1] 4.2 11.5 7.3 5.8 6.4 10.0 11.2 11.2 5.2 7.0
> l2<-d2$len
> l2
[1] 16.5 16.5 15.2 17.3 22.5 17.3 13.6 14.5 18.8 15.5
> l3<-d3$len
> l3
[1] 23.6 18.5 33.9 25.5 26.4 32.5 26.7 21.5 23.3 29.5
> q<-wilcox.test(l1,l2,paired = T)
> w<-wilcox.test(l1,l3,paired = T)
> e<-wilcox.test(l2,l3,paired = T)
> q

```

Wilcoxon signed rank exact test

data: l1 and l2

V = 0, p-value = 0.001953

alternative hypothesis: true location shift is not equal to 0

```
> w
```

Wilcoxon signed rank test with continuity correction

data: l1 and l3

V = 0, p-value = 0.005889

alternative hypothesis: true location shift is not equal to 0

```
> e
```

Wilcoxon signed rank exact test

data: l2 and l3

V = 0, p-value = 0.001953

alternative hypothesis: true location shift is not equal to 0

```

> p<-wilcox.test(l2,l1,paired = T)
> o<-wilcox.test(l3,l1,paired = T)
> i<-wilcox.test(l2,l3,paired = T)
> p

```

### Wilcoxon signed rank exact test

data: l2 and l1

V = 55, p-value = 0.001953

alternative hypothesis: true location shift is not equal to 0

> 0

### Wilcoxon signed rank test with continuity correction

data: l3 and l1

V = 55, p-value = 0.005889

alternative hypothesis: true location shift is not equal to 0

>

### MANN-WHITNEY U TEST

Q-9. Let the score of two groups of person out of which one under plesibo and other under a new drug given as

X= score under plesibo

Y=score under new drug

X	10,13,12,15,16,8,6
Y	20,14,7,9,17,18,19,25,24

Test whether the distribution of score of new drug is more suitable then under plesibo.

Given that  $u_{\alpha}=16$  for this data set,  $n_1=7$  &  $n_2=9$ .

Solution:

```
> x=c(10,13,12,15,16,8,6)
```

```
> y=c(20,14,7,9,17,18,19,25,24)
```

```
> x
```

```
[1] 10 13 12 15 16 8 6
```

```
> y
```

```
[1] 20 14 7 9 17 18 19 25 24
```

```
> n1=7
```

```
> n2=9
```

```
> n=n1+n2
```

```
> n
```

```
[1] 16
```

```
> m=n1*n2
```

```
> m
```

```
[1] 63
```

```
> z=c(x,y)
```

```
> z
```

```
[1] 10 13 12 15 16 8 6 20 14 7 9 17 18 19 25 24
```

```
> r=rank(z)
```

```
> r
```

```

[1] 5 7 6 9 10 3 1 14 8 2 4 11 12 13 16 15
> len=length(x)
> sum_rank=sum(r[1:len])
> sum_rank
[1] 41
> s=rank(x)
> s
[1] 3 5 4 6 7 2 1
> l=7
> sum_xrank=sum(s[1:l])
> sum_xrank
[1] 28
> u=sum_rank-sum_xrank
> u
[1] 13

```

### Conclusion:

The value of  $u_\alpha$  is greater than  $u$ , So the distribution of score of new drug is more suitable than under placebo.

```
> wilcox.test(x,y,alternative ="two.sided")
```

Wilcoxon rank sum exact test

data: x and y

W = 13, p-value = 0.0549

alternative hypothesis: true location shift is not equal to 0

### Key used:

```

x=c(10,13,12,15,16,8,6)
y=c(20,14,7,9,17,18,19,25,24)
x
y
n1=7
n2=9
n=n1+n2
n
m=n1*n2
m
z=c(x,y)
z
r=rank(z)
r
len=length(x)
sum_rank=sum(r[1:len])
sum_rank
s=rank(x)

```

```
s
l=7
sum_xrank=sum(s[1:l])
sum_xrank
u=sum_rank-sum_xrank
u

wilcox.test(x,y,alternative ="two.sided")
```