

```

> #read in the data from a csv file
> dat1=read.table("Data Sets/Chapter 2/Examples/ex-2-1.csv", header=T, sep=",")
> #if the data has no header and is separated by space in a txt file then use
> #dat1=read.table("Data Sets/Chapter 2/Examples/ex-2-1.txt", sep=" ")
> str(dat1) #check the data

```

```

'data.frame':  20 obs. of  3 variables:
 $ Observation..i      : int  1 2 3 4 5 6 7 8 9 10 ...
 $ Shear.Strength..yi..psi. : num  2159 1678 2316 2061 2208 ...
 $ Age.of.Propellant..xi..weeks.: num  15.5 23.8 8 17 5.5 ...

```

```

> dat1 #print out the dat

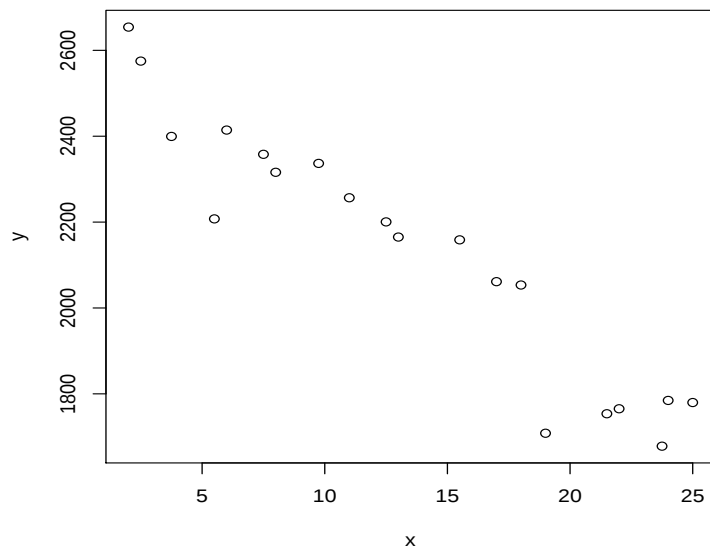
```

	Observation..i	Shear.Strength..yi..psi.	Age.of.Propellant..xi..weeks.
1	1	2158.70	15.50
2	2	1678.15	23.75
3	3	2316.00	8.00
4	4	2061.30	17.00
5	5	2207.50	5.50
6	6	1708.30	19.00
7	7	1784.70	24.00
8	8	2575.00	2.50
9	9	2357.90	7.50
10	10	2256.70	11.00
11	11	2165.20	13.00
12	12	2399.55	3.75
13	13	1779.80	25.00
14	14	2336.75	9.75
15	15	1765.30	22.00
16	16	2053.50	18.00
17	17	2414.40	6.00
18	18	2200.50	12.50
19	19	2654.20	2.00
20	20	1753.70	21.50

```

> y=dat1[,2] #assign the second column in dat1 to be variable y
> x=dat1[,3] #assign the third column in dat1 to be variable x
> plot(y~x) #plot the data

```



```
> fm1=lm(y~x) #fit a simple linear model with  $E(y)=\beta_0+\beta_1x_1$ 
> summary(fm1) # the properties of the coefficients for the fitting line
```

Call:

```
lm(formula = y ~ x)
```

Residuals:

Min	1Q	Median	3Q	Max
-215.98	-50.68	28.74	66.61	106.76

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	2627.822	44.184	59.48	< 2e-16
x	-37.154	2.889	-12.86	1.64e-10

Residual standard error: 96.11 on 18 degrees of freedom

Multiple R-squared: 0.9018, Adjusted R-squared: 0.8964

F-statistic: 165.4 on 1 and 18 DF, p-value: 1.643e-10

```
> anova(fm1) #the analysis of variance table
```

Analysis of Variance Table

Response: y

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
x	1	1527483	1527483	165.38	1.643e-10
Residuals	18	166255	9236		

```

> resid(fm1) # residuals, vector for e

      1      2      3      4      5      6
106.758301 -67.274574 -14.593631  65.088687 -215.977609 -213.604131
      7      8      9     10     11     12
 48.563824  40.061618   8.729573  37.567141  20.374323 -88.946393
     13     14     15     16     17     18
 80.817415  71.175153 -45.143358  94.442278   9.499187  37.097528
     19     20
100.684823 -75.320154

> df.residual(fm1) # degree of freedom for the residual sum of squares

[1] 18

> coef(fm1) # estimates for beta0 and beta1

(Intercept)          x
 2627.82236   -37.15359

> #assign some notations
> ybar=mean(y);ybar

[1] 2131.358

> xbar=mean(x);xbar

[1] 13.3625

> ei=resid(fm1)
> yhat=fitted(fm1) #y.hat
> df=df.residual(fm1)
> print(cbind(x,y,yhat, ei)) #cbind: combine the vectors by column

      x      y      yhat      ei
1  15.50 2158.70 2051.942 106.758301
2  23.75 1678.15 1745.425 -67.274574
3   8.00 2316.00 2330.594 -14.593631
4  17.00 2061.30 1996.211  65.088687
5   5.50 2207.50 2423.478 -215.977609
6  19.00 1708.30 1921.904 -213.604131
7  24.00 1784.70 1736.136  48.563824
8   2.50 2575.00 2534.938  40.061618
9   7.50 2357.90 2349.170   8.729573
10 11.00 2256.70 2219.133  37.567141
11 13.00 2165.20 2144.826  20.374323
12  3.75 2399.55 2488.496 -88.946393

```

```

13 25.00 1779.80 1698.983    80.817415
14  9.75 2336.75 2265.575    71.175153
15 22.00 1765.30 1810.443   -45.143358
16 18.00 2053.50 1959.058    94.442278
17  6.00 2414.40 2404.901     9.499187
18 12.50 2200.50 2163.402    37.097528
19  2.00 2654.20 2553.515   100.684823
20 21.50 1753.70 1829.020   -75.320154

```

```
> sum(ei^2)      # Residual sum of squares
```

```
[1] 166254.9
```

```
> sum(ei^2)/df    # Residual Mean Squares
```

```
[1] 9236.381
```

```
> s2.hat=sum(ei^2)/df
```

```
> sum((y-ybar)^2)
```

```
[1] 1693738
```

```
> sum((yhat-ybar)^2)
```

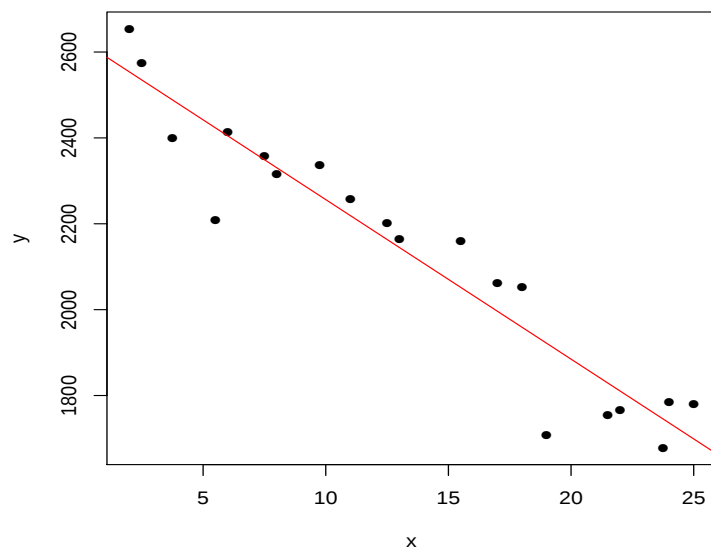
```
[1] 1527483
```

```
> #plot the data and the regression line
```

```
> fm1=lm(y~x)
```

```
> plot(y~x, pch=16) #pch: change the symbol from o to solid dot
```

```
> abline(coef(fm1), col=2)
```



Confident interval for β_0 and β_1 .

```

> coef(fm1)

(Intercept)          x
  2627.82236    -37.15359

> b0=coef(fm1)[1];b1=coef(fm1)[2]
> confint(fm1)

                2.5 %      97.5 %
(Intercept) 2534.99540 2720.64931
x           -43.22338  -31.08380

```

Done by hands

```

> Sxy=sum( (y-mean(y)) * (x-mean(x)) ); Sxy

[1] -41112.65

> Sxx=sum( (x-mean(x))^2 ); Sxx

[1] 1106.559

> se.b0=sqrt(s2.hat * (1/20+mean(x)^2/Sxx));se.b0

[1] 44.18391

> se.b1=sqrt(s2.hat/Sxx) ;se.b1

[1] 2.889107

> t.lim=qt(0.025,18) # .025 quantile for t(18)
> c(b1 - t.lim * se.b1, b1 + t.lim * se.b1)

                x          x
-31.08380 -43.22338

```

Confident interval and prediction interval. use the command **predict()**

```

> fm1=lm(y~x)
> plot(y~x, xlab="age of Propellant, x", ylab="Shear Strength, y")
> abline(coef(fm1), col=2) #col=2: change the colour of the line to be red

```

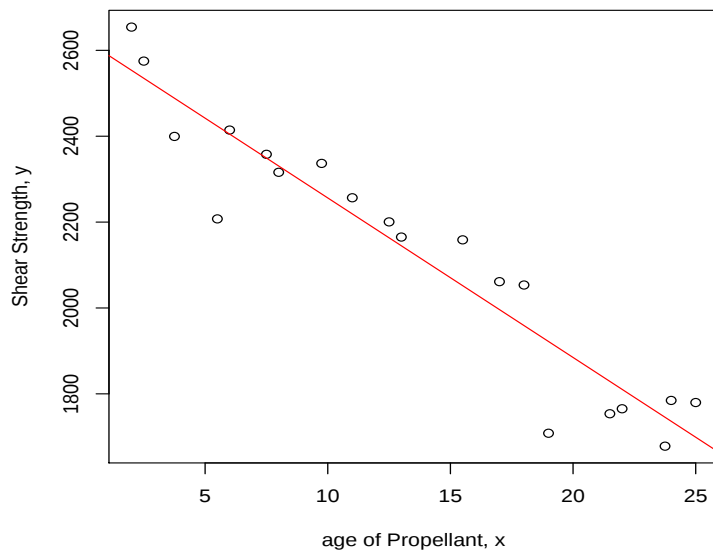


figure 2.4

`predict(fm1, new, interval="confidence")`: find the confidence interval for mean response for a set of given explanatory variables (say new).

`Y.CI`: matrix with 3 columns, one mean, lower limits and upper limits.

`matplot`: Plot the columns of one matrix against the columns of another. `x vs Y.CI[,1]`, `x vs Y.CI[,2]`, `x vs Y.CI[,3]`

`type="l"`, plot type =line, `lty`: line type, `col`: line color.

```
> new = data.frame(x = seq(3, 24, 3)) #table 2.6
> predict(fm1, new, interval="confidence")
```

	fit	lwr	upr
1	2516.362	2438.937	2593.786
2	2404.901	2341.375	2468.426
3	2293.440	2241.099	2345.781
4	2181.979	2136.079	2227.879
5	2070.518	2024.289	2116.748
6	1959.058	1905.853	2012.263
7	1847.597	1782.886	1912.308
8	1736.136	1657.349	1814.923

```
> Y.CI=predict(fm1, new, interval="confidence");Y.CI #assign the CIs to Y.CI
```

	fit	lwr	upr
1	2516.362	2438.937	2593.786
2	2404.901	2341.375	2468.426
3	2293.440	2241.099	2345.781
4	2181.979	2136.079	2227.879

```

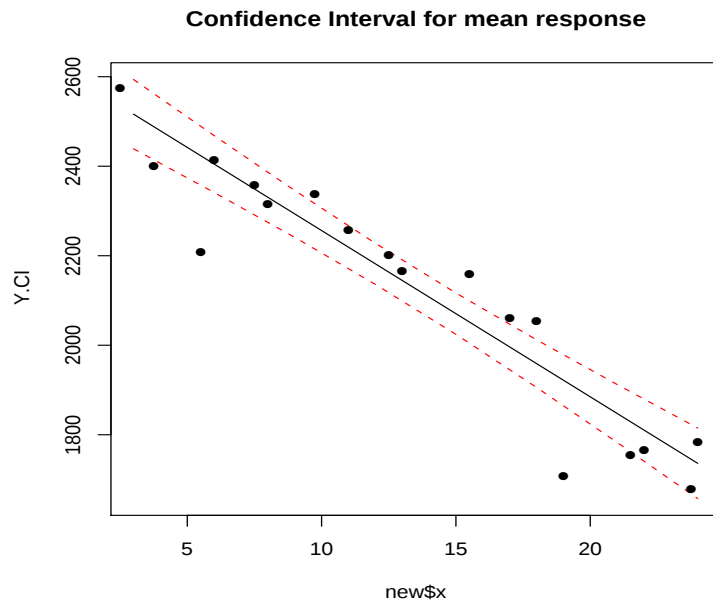
5 2070.518 2024.289 2116.748
6 1959.058 1905.853 2012.263
7 1847.597 1782.886 1912.308
8 1736.136 1657.349 1814.923

```

```

> matplot(new$x, Y.CI, col=c(1,2,2), lty=c(1,2,2), type="l",
+         main="Confidence Interval for mean response")
> points(x,y,pch=16) #add data to the plot

```



predict(fm1, new, interval="prediction"): find the prediction interval for a set of new given explanatory variables (say new).

Y.PI: matrix with 3 columns, one mean, lower limits and upper limits.

```

> Y.PI=predict(fm1, new, interval="prediction");Y.PI

```

	fit	lwr	upr
1	2516.362	2300.114	2732.609
2	2404.901	2193.232	2616.570
3	2293.440	2084.855	2502.025
4	2181.979	1974.916	2389.042
5	2070.518	1863.382	2277.655
6	1959.058	1750.254	2167.861
7	1847.597	1635.569	2059.625
8	1736.136	1519.398	1952.875

```

> matplot(new$x, Y.PI, col=c(1,4,4), lty=c(1,4,4), type="l",
+         main="Prediction Interval for new obs")
> points(x,y,pch=16) #add data to the plot

```

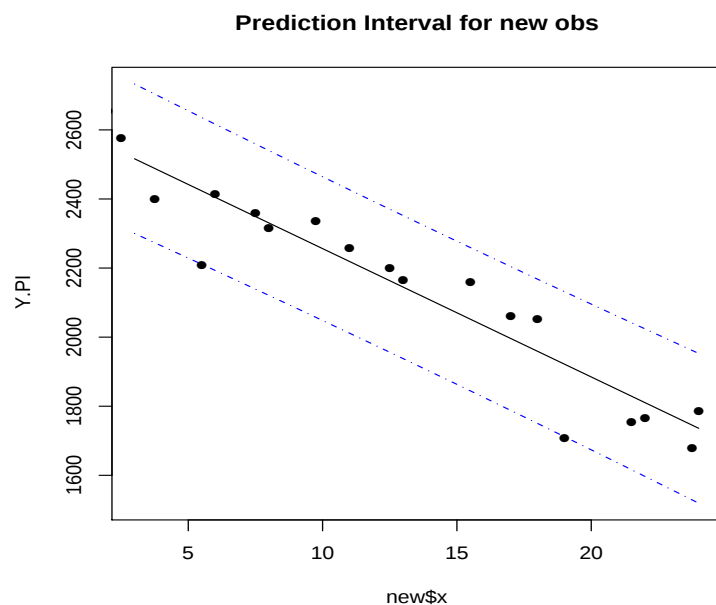


figure 2.5

add legend to the plot on the top-right corner.

```
> Y.mat=cbind(Y.CI, Y.PI[, -1]) #Y.mat, mean, lower.CI, upper.CI, lower.PI, upper.PI
> matplot(new$x, Y.mat, col=c(1,2,2,4,4), lty=c(1,2,2,4,4),
  type="l", xlab="x", ylab="response")
> points(x,y, pch=16) #add data to the plot
> legend("topright", legend=c("confint", "pred.int"), col=c(2,4), lty=c(2,4))
```

