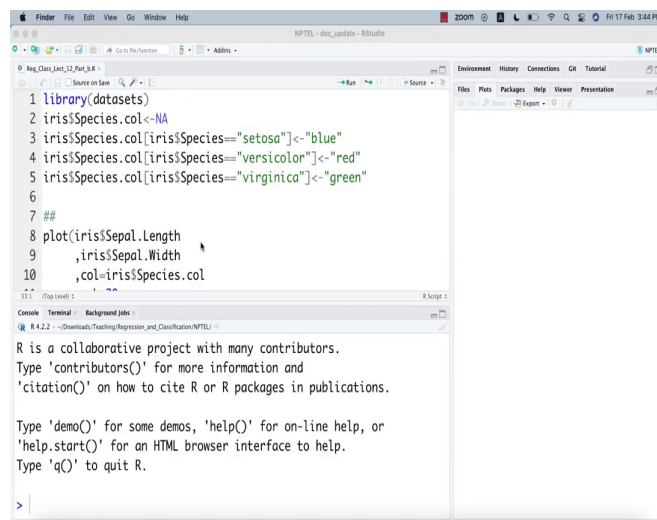


Predictive Analytics - Regression and Classification
Prof. Sourish Das
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Lecture - 41
Hands on with R with Iris dataset

Welcome to the part b of lecture 12. In this video, we are going to do some Hands on. So, in this hands on we are going to load the dataset library which has the; which has the iris dataset.

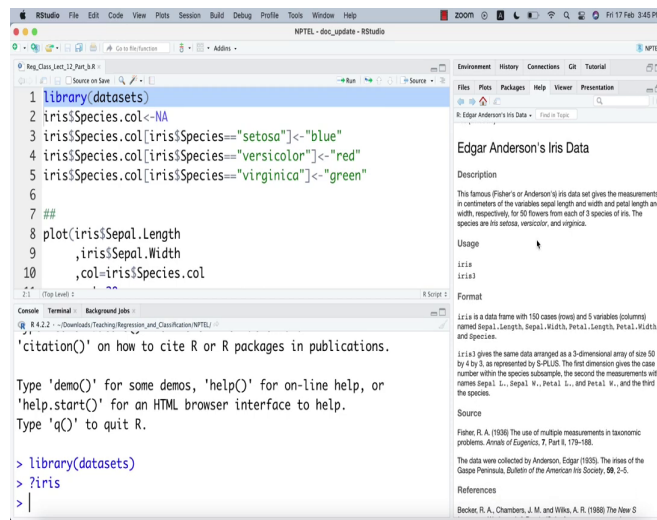
(Refer Slide Time: 00:23)



```
1 library(datasets)
2 iris$Species.col<-NA
3 iris$Species.col[iris$Species=="setosa"]<-"blue"
4 iris$Species.col[iris$Species=="versicolor"]<-"red"
5 iris$Species.col[iris$Species=="virginica"]<-"green"
6
7 ##
8 plot(iris$Sepal.Length
9      ,iris$Sepal.Width
10     ,col=iris$Species.col
```



(Refer Slide Time: 00:32)



```
1 library(datasets)
2 iris$Species.col<-NA
3 iris$Species.col[iris$Species=="setosa"]<-"blue"
4 iris$Species.col[iris$Species=="versicolor"]<-"red"
5 iris$Species.col[iris$Species=="virginica"]<-"green"
6
7 ##
8 plot(iris$Sepal.Length
9      ,iris$Sepal.Width
10     ,col=iris$Species.col
```

Console Terminal Background jobs

```
R 4.2.2 -> Downloads/Teaching/Regression_and_Classification/NPTEL/
> citation() on how to cite R or R packages in publications.
Type 'demo()' for some demos, 'help()' for on-line help, or
'help.start()' for an HTML browser interface to help.
Type 'q()' to quit R.
> library(datasets)
> ?iris
> |
```

Environment History Connectors CUI Tutorial

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Edgar Anderson's Iris Data

Description

This famous (Fisher's or Anderson's) iris data set gives the measurements in centimeters of the variables sepal length and width and petal length and width, respectively, for 50 flowers from each of 3 species of iris. The species are iris setosa, versicolor, and virginica.

Usage

```
iris
iris()
```

Format

iris is a data frame with 150 cases (rows) and 5 variables (columns) named Sepal.Length, Sepal.Width, Petal.Length, Petal.Width, and Species.

iris() gives the same data arranged as a 3-dimensional array of size 50 by 4 by 3, as represented by S-PLUS. The first dimension gives the case number within the species subsample, the second the measurements with names Sepal.Length, Sepal.Width, Petal.Length, and Petal.Width, and the third the species.

Source

Fisher, R. A. (1936) The use of multiple measurements in taxonomic problems. *Annals of Eugenics*, 7, Part II, 179-188.

The data were collected by Anderson, Edgar (1935). *The Irises of the Gaspe Peninsula*, *Bulletin of the American Iris Society*, 59, 2-5.

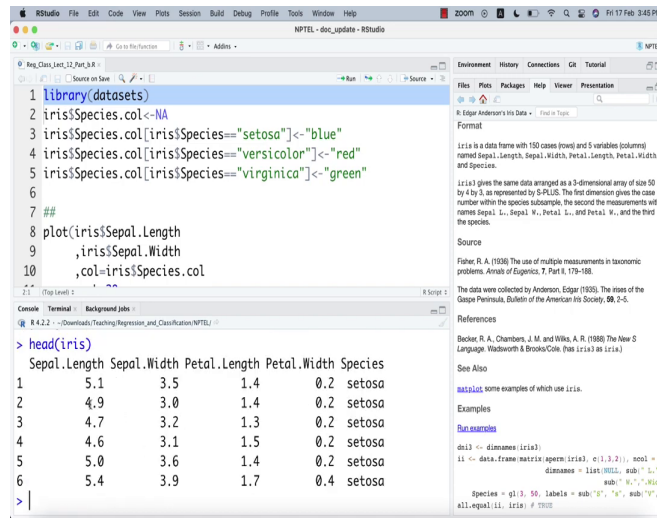
References

Becker, R. A., Chambers, J. M. and Wilks, A. R. (1988) *The New S*



So, if you go to the load the, you write iris. So, Edgar Anderson's iris datasets, it is also known as Fisher's iris dataset. So, it has iris is a English flower, which has three subspecies, one is called Setosa, one is called versicolor, and one is called virginica.

(Refer Slide Time: 01:02)



```
1 library(datasets)
2 iris$Species.col<-NA
3 iris$Species.col[iris$Species=="setosa"]<-"blue"
4 iris$Species.col[iris$Species=="versicolor"]<-"red"
5 iris$Species.col[iris$Species=="virginica"]<-"green"
6
7 ##
8 plot(iris$Sepal.Length
9      ,iris$Sepal.Width
10     ,col=iris$Species.col
```

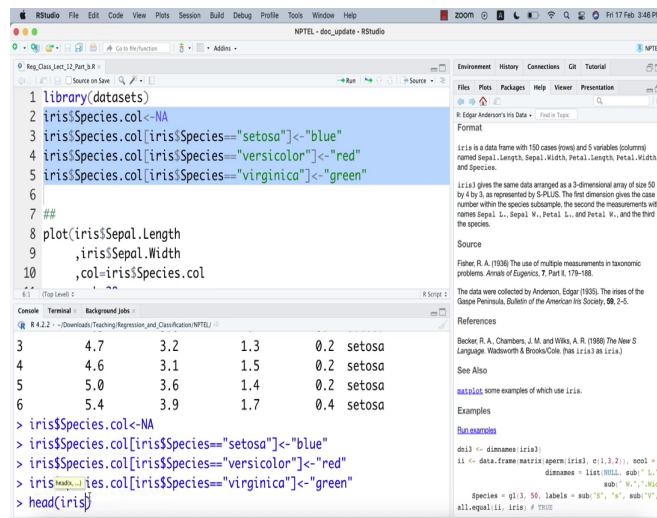
Console

```
> head(iris)
  Sepal.Length Sepal.Width Petal.Length Petal.Width Species
1          5.1         3.5          1.4         0.2  setosa
2          4.9         3.0          1.4         0.2  setosa
3          4.7         3.2          1.3         0.2  setosa
4          4.6         3.1          1.5         0.2  setosa
5          5.0         3.6          1.4         0.2  setosa
6          5.4         3.9          1.7         0.4  setosa
```



In this dataset, what happens is if you just say, let us say, iris, head iris. So, it has four predictors, sepal length, sepal width, petal length, and petal width. This four based on these four phenotype can you say which species the flower belongs to? Ok. So, what we I have done here, I for different colors setosa, versicolor, and virginica.

(Refer Slide Time: 01:48)



```
1 library(datasets)
2 iris$Species.col<-NA
3 iris$Species.col[iris$Species=="setosa"]<-"blue"
4 iris$Species.col[iris$Species=="versicolor"]<-"red"
5 iris$Species.col[iris$Species=="virginica"]<-"green"
6
7 ##
8 plot(iris$Sepal.Length
9      ,iris$Sepal.Width
10     ,col=iris$Species.col)

```

Console

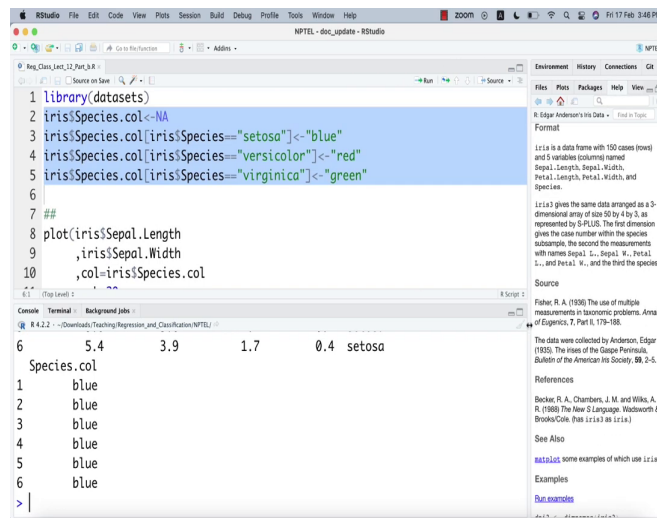
```
R 4.2.2 -> Downloads/Teaching/Regression_and_Classification/NPTEL/
3      4.7      3.2      1.3      0.2 setosa
4      4.6      3.1      1.5      0.2 setosa
5      5.0      3.6      1.4      0.2 setosa
6      5.4      3.9      1.7      0.4 setosa
> iris$Species.col<-NA
> iris$Species.col[iris$Species=="setosa"]<-"blue"
> iris$Species.col[iris$Species=="versicolor"]<-"red"
> iris$Species.col[iris$Species=="virginica"]<-"green"
> head(iris)

```



I have given different color, for different species, I have now had different color. So, now, if you go to head iris.

(Refer Slide Time: 01:55)



```
1 library(datasets)
2 iris$Species.col<-NA
3 iris$Species.col[iris$Species=="setosa"]<-"blue"
4 iris$Species.col[iris$Species=="versicolor"]<-"red"
5 iris$Species.col[iris$Species=="virginica"]<-"green"
6
7 ##
8 plot(iris$Sepal.Length
9       ,iris$Sepal.Width
10      ,col=iris$Species.col
```

Console Terminal | Background jobs

```
R 4.2.2 -> Downloads/Teaching/Regression_and_Classification/NPTEL/
6      5.4      3.9      1.7      0.4 setosa
Species.col
1      blue
2      blue
3      blue
4      blue
5      blue
6      blue
> |
```

Environment History Connections CR |

Files Plots Packages Help View |

R: Edge Anderson's Iris Data | Find in Topic

Format

iris is a data frame with 150 cases (rows) and 5 variables (columns) named Sepal.Length, Sepal.Width, Petal.Length, Petal.Width, and Species.

iris is the same data arranged as a 3-dimensional array of size 50 by 4 by 3, as represented by S-PLUS. The first dimension gives the case number within the species subsample, the second the measurements with names Sepal.L., Sepal.W., Petal.L., and Petal.W., and the third the species.

Source

Fisher, R. A. (1936) The use of multiple measurements in taxonomic problems. *Annals of Eugenics*, 7, Part II, 179-186.

The data were collected by Anderson, Edgar (1935). *The base of the Sierra Peninsula*. *Bulletin of the American Iris Society*, 88, 2-5.

References

Becker, R. A., Chambers, J. M. and Wilks, A. R. (1988) *The New S Language*. Wadsworth & Brooks/Cole (has `iris` as `iris`)

See Also

[iris3d](#) some examples of which use `iris`.

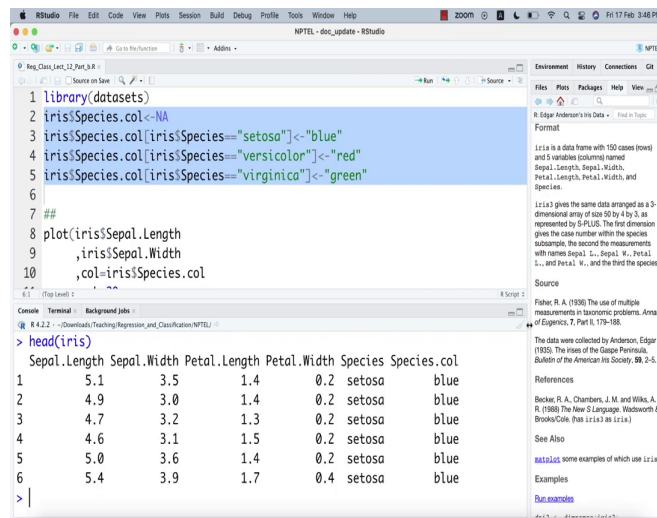
Examples

[Run examples](#)



So, you will see that in along with that, let me just.

(Refer Slide Time: 01:59)



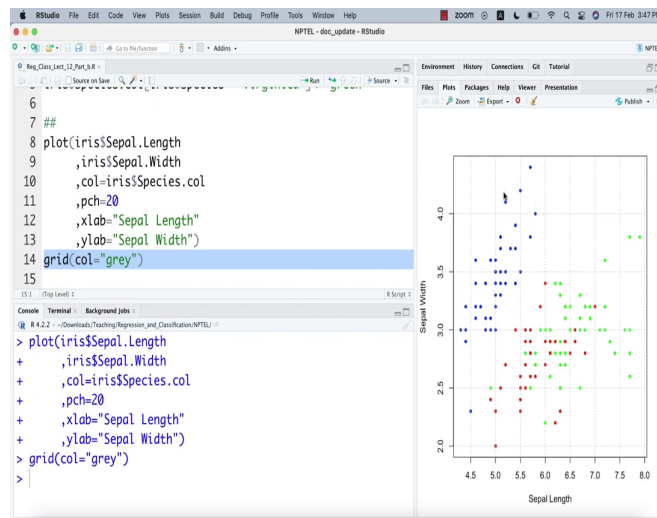
```
1 library(datasets)
2 iris$Species.col<-NA
3 iris$Species.col[iris$Species=="setosa"]<-"blue"
4 iris$Species.col[iris$Species=="versicolor"]<-"red"
5 iris$Species.col[iris$Species=="virginica"]<-"green"
6
7 ##
8 plot(iris$Sepal.Length
9      ,iris$Sepal.Width
10     ,col=iris$Species.col
11     )
```

```
> head(iris)
  Sepal.Length Sepal.Width Petal.Length Petal.Width Species Species.col
1          5.1         3.5          1.4          0.2 setosa      blue
2          4.9         3.0          1.4          0.2 setosa      blue
3          4.7         3.2          1.3          0.2 setosa      blue
4          4.6         3.1          1.5          0.2 setosa      blue
5          5.0         3.6          1.4          0.2 setosa      blue
6          5.4         3.9          1.7          0.4 setosa      blue
```



Yeah. So, now, this is the new column, that last column is a new column. For setosa, I had blue, versicolor I have given red, and virginica I given green.

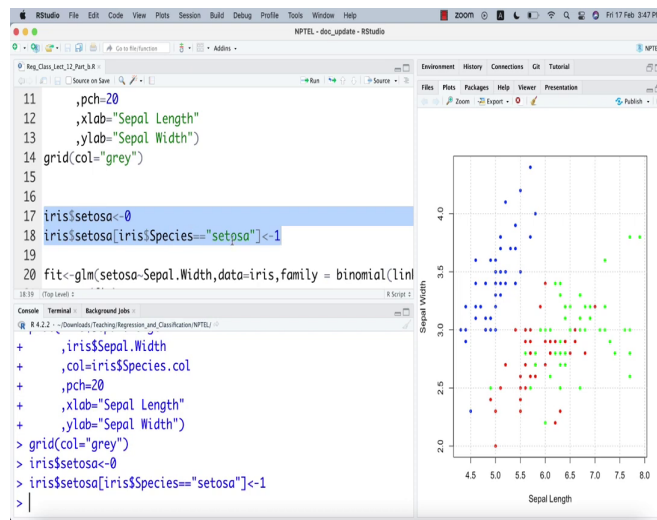
(Refer Slide Time: 02:18)



Now, if I now if you run this piece of code, I just plotting from iris dataset, I am just taking the sepal length and sepal width. I am just extracting it and giving it as a x value and y values. And if you run this plot, then you got this plot, you want, you can give a grid also.

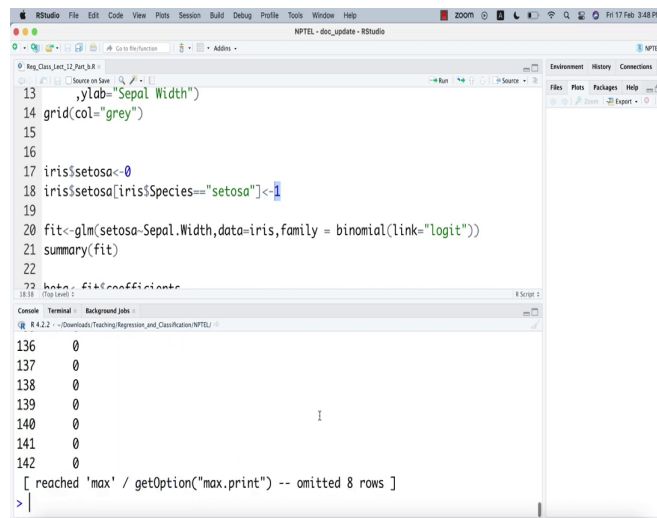
So, you can see these are like bluer setosa, we have given bluer setosa, reds are red points are all versicolor, and the green points are virginica.

(Refer Slide Time: 02:58)



Now, what I have done in that dataset, I have created a one-hot encoding for setosa. If it is setosa, then it will have a 0 or it is 1.

(Refer Slide Time: 03:21)



```
13   ,ylab="Sepal Width")
14   grid(col="grey")
15
16
17   iris$setosa<-0
18   iris$setosa[iris$Species=="setosa"]<-1
19
20   fit<-glm(setosa~Sepal.Width,data=iris,family = binomial(link="logit"))
21   summary(fit)
22
23   ##>>> fit$coefficients
```

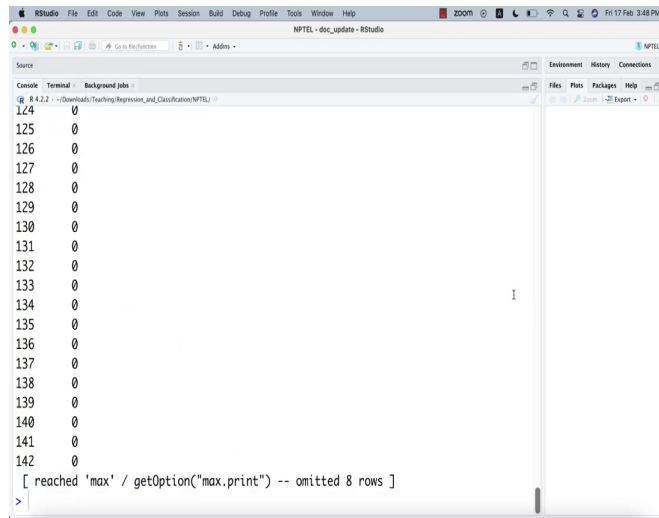
Console

```
136  0
137  0
138  0
139  0
140  0
141  0
142  0
[ reached 'max' / getOption("max.print") -- omitted 8 rows ]
>
```



So, for all it is 0, but whenever it will find setosa, it will get a 1 otherwise, it will be 0. So, now the dataset, let me just let me just show you how the dataset look likes now, iris, ok.

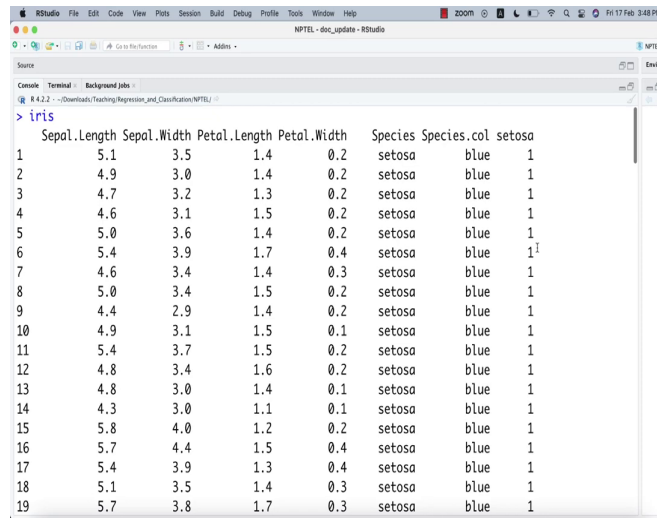
(Refer Slide Time: 03:36)



```
RStudio File Edit Code View Plots Session Build Debug Profile Tools Window Help
NPTEL - doc_update - RStudio
Source Environment History Connections
Console Terminal Background jobs
R 4.2.2 -> /Downloads/Teaching/Regression_and_Classification/NPTEL/
124 0
125 0
126 0
127 0
128 0
129 0
130 0
131 0
132 0
133 0
134 0
135 0
136 0
137 0
138 0
139 0
140 0
141 0
142 0
[reached 'max' /getOption('max.print')] -- omitted 8 rows ]
>
```



(Refer Slide Time: 03:40)



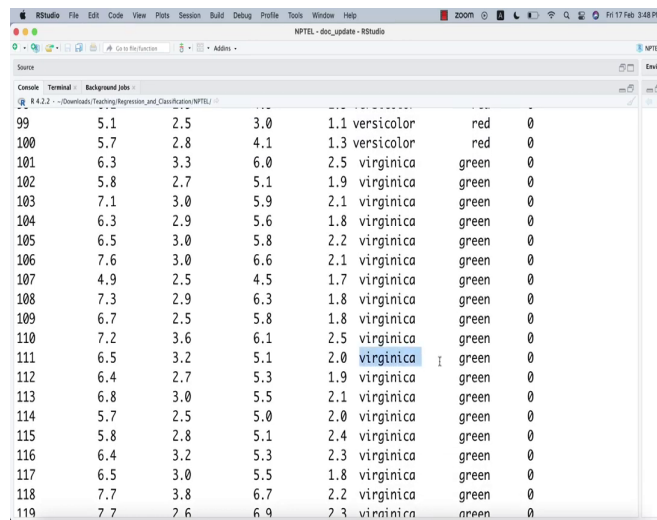
The screenshot shows the RStudio interface with the following content in the console:

```
> iris
```

| | Sepal.Length | Sepal.Width | Petal.Length | Petal.Width | Species | Species.col | setosa |
|----|--------------|-------------|--------------|-------------|---------|-------------|----------------|
| 1 | 5.1 | 3.5 | 1.4 | 0.2 | setosa | blue | 1 |
| 2 | 4.9 | 3.0 | 1.4 | 0.2 | setosa | blue | 1 |
| 3 | 4.7 | 3.2 | 1.3 | 0.2 | setosa | blue | 1 |
| 4 | 4.6 | 3.1 | 1.5 | 0.2 | setosa | blue | 1 |
| 5 | 5.0 | 3.6 | 1.4 | 0.2 | setosa | blue | 1 |
| 6 | 5.4 | 3.9 | 1.7 | 0.4 | setosa | blue | 1 ¹ |
| 7 | 4.6 | 3.4 | 1.4 | 0.3 | setosa | blue | 1 |
| 8 | 5.0 | 3.4 | 1.5 | 0.2 | setosa | blue | 1 |
| 9 | 4.4 | 2.9 | 1.4 | 0.2 | setosa | blue | 1 |
| 10 | 4.9 | 3.1 | 1.5 | 0.1 | setosa | blue | 1 |
| 11 | 5.4 | 3.7 | 1.5 | 0.2 | setosa | blue | 1 |
| 12 | 4.8 | 3.4 | 1.6 | 0.2 | setosa | blue | 1 |
| 13 | 4.8 | 3.0 | 1.4 | 0.1 | setosa | blue | 1 |
| 14 | 4.3 | 3.0 | 1.1 | 0.1 | setosa | blue | 1 |
| 15 | 5.8 | 4.0 | 1.2 | 0.2 | setosa | blue | 1 |
| 16 | 5.7 | 4.4 | 1.5 | 0.4 | setosa | blue | 1 |
| 17 | 5.4 | 3.9 | 1.3 | 0.4 | setosa | blue | 1 |
| 18 | 5.1 | 3.5 | 1.4 | 0.3 | setosa | blue | 1 |
| 19 | 5.7 | 3.8 | 1.7 | 0.3 | setosa | blue | 1 |



(Refer Slide Time: 03:42)

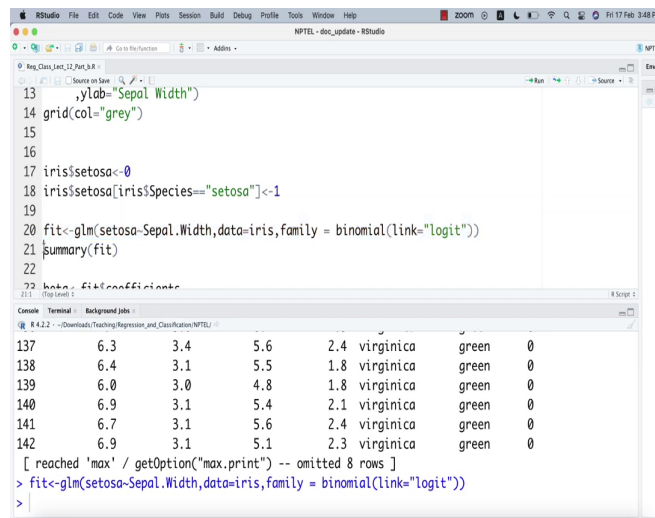


```
RStudio - doc_update - RStudio
Source
Console Terminal Background jobs
R 4.2.2 ~~/Downloads/Teaching/Regressor_and_Classification/NPTEL/
99      5.1      2.5      3.0      1.1 versicolor      red      0
100     5.7      2.8      4.1      1.3 versicolor      red      0
101     6.3      3.3      6.0      2.5 virginica        green    0
102     5.8      2.7      5.1      1.9 virginica        green    0
103     7.1      3.0      5.9      2.1 virginica        green    0
104     6.3      2.9      5.6      1.8 virginica        green    0
105     6.5      3.0      5.8      2.2 virginica        green    0
106     7.6      3.0      6.6      2.1 virginica        green    0
107     4.9      2.5      4.5      1.7 virginica        green    0
108     7.3      2.9      6.3      1.8 virginica        green    0
109     6.7      2.5      5.8      1.8 virginica        green    0
110     7.2      3.6      6.1      2.5 virginica        green    0
111     6.5      3.2      5.1      2.0 virginica        green    0
112     6.4      2.7      5.3      1.9 virginica        green    0
113     6.8      3.0      5.5      2.1 virginica        green    0
114     5.7      2.5      5.0      2.0 virginica        green    0
115     5.8      2.8      5.1      2.4 virginica        green    0
116     6.4      3.2      5.3      2.3 virginica        green    0
117     6.5      3.0      5.5      1.8 virginica        green    0
118     7.7      3.8      6.7      2.2 virginica        green    0
119     7.7      2.6      6.9      2.3 virginica        green    0
```



So, let me just, yeah. So, first few values are setosa and then when it is versicolor, it is coded as 0, when it is virginica, this is coded as 0, ok.

(Refer Slide Time: 04:25)



```
13     ,ylab="Sepal Width")
14 grid(col="grey")
15
16
17 iris$setosa<-0
18 iris$setosa[iris$Species=="setosa"]<-1
19
20 fit<-glm(setosa~Sepal.Width,data=iris,family = binomial(link="logit"))
21 summary(fit)
22
23 ##>>> fit$coefficients
```

```
R 4.2.2 -> Downloads/Teaching/Regression_and_Classification/NPTEL/
137      6.3      3.4      5.6      2.4 virginica green  0
138      6.4      3.1      5.5      1.8 virginica green  0
139      6.0      3.0      4.8      1.8 virginica green  0
140      6.9      3.1      5.4      2.1 virginica green  0
141      6.7      3.1      5.6      2.4 virginica green  0
142      6.9      3.1      5.1      2.3 virginica green  0
[ reached 'max' / getOption("max.print") -- omitted 8 rows ]
> fit<-glm(setosa~Sepal.Width,data=iris,family = binomial(link="logit"))
>
```



And the column name is setosa. So, that is how I created a one-hot encoding or binary class variable, sometime it is called indicator variable. So, let me run the model, glm, setosa as a function of Sepal Width, data equal to iris, you give family equal to binomial link equal to logit.

(Refer Slide Time: 04:29)

```
RStudio
NPTEL - doc_update - RStudio

13   ,ylab="Sepal Width")
14   grid(col="grey")
15
16
17   iris$setosa<-0
18   iris$setosa[iris$Species=="setosa"]<-1
19
20   fit<-glm(setosa~Sepal.Width,data=iris,family = binomial(link="logit"))
21   summary(fit)
22
23   ##>>> fit$coefficients

Console Terminal | Background jobs
R 4.2.2 -> Downloads/Teaching/Regression_and_Classification/NPTEL/
(Dispersion parameter for binomial family taken to be 1)

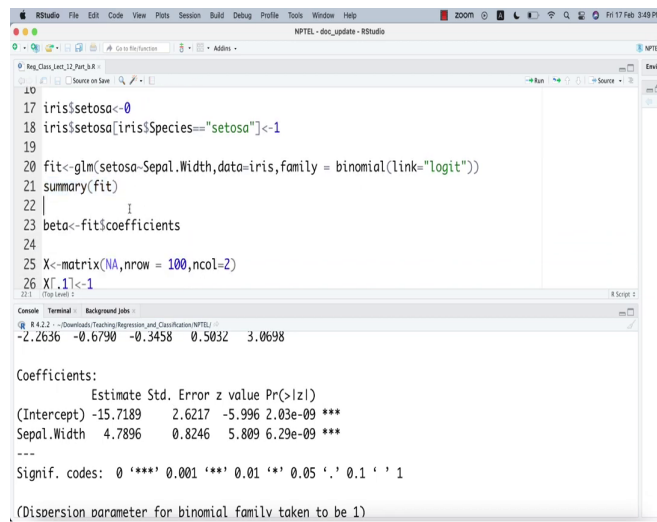
Null deviance: 190.95  on 149  degrees of freedom
Residual deviance: 123.83  on 148  degrees of freedom
AIC: 127.83

Number of Fisher Scoring iterations: 5

>
```



(Refer Slide Time: 04:32)



```
10
17 iris$setosa<-0
18 iris$setosa[iris$Species=="setosa"]<-1
19
20 fit<-glm(setosa~Sepal.Width,data=iris,family = binomial(link="logit"))
21 summary(fit)
22
23 beta<-fit$coefficients
24
25 X<-matrix(NA,nrow = 100,ncol=2)
26 X[,1]<-1
```

Console Terminal Background jobs

```
R 4.2.2 ->Downloads/TeachingRegression_and_Classification/NPTEL/
-2.2636 -0.6790 -0.3458 0.5032 3.0698
```

Coefficients:

| | Estimate | Std. Error | z value | Pr(> z) |
|-------------|----------|------------|---------|--------------|
| (Intercept) | -15.7189 | 2.6217 | -5.996 | 2.03e-09 *** |
| Sepal.Width | 4.7896 | 0.8246 | 5.809 | 6.29e-09 *** |

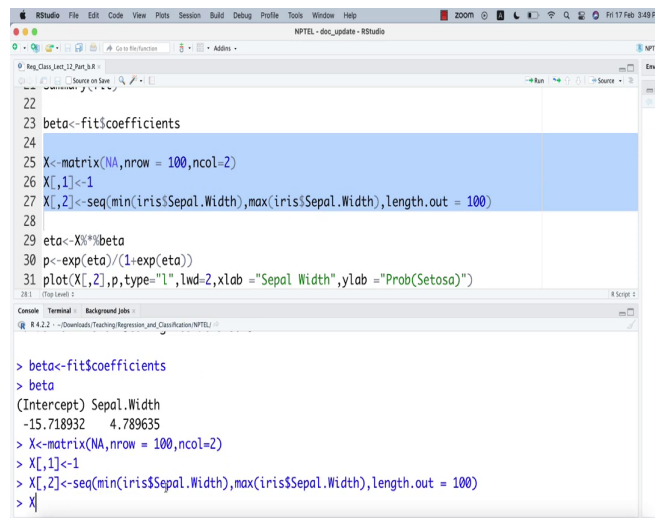
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for binomial family taken to be 1)



And if you run summary fit, then this is the fit that you will get.

(Refer Slide Time: 04:35)



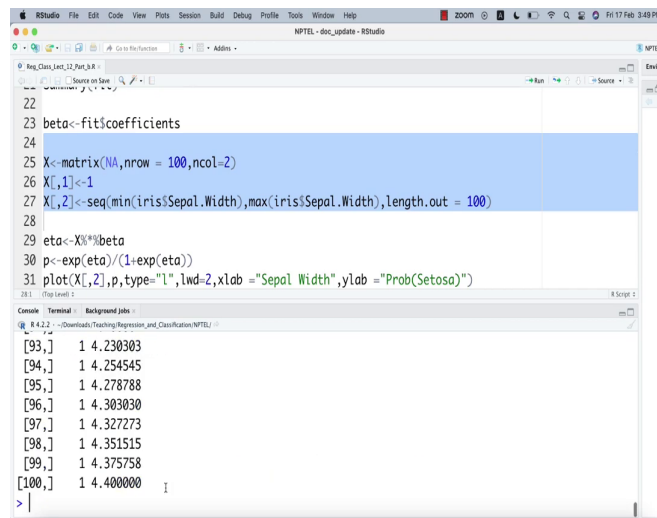
```
22
23 beta<-fit$coefficients
24
25 X<-matrix(NA,nrow = 100,ncol=2)
26 X[,1]<-1
27 X[,2]<-seq(min(iris$Sepal.Width),max(iris$Sepal.Width),length.out = 100)
28
29 eta<-X%*%beta
30 p<-exp(eta)/(1+exp(eta))
31 plot(X[,2],p,type="l",lwd=2,xlab="Sepal Width",ylab="Prob(Setosa)")

> beta<-fit$coefficients
> beta
(Intercept) Sepal.Width
-15.718932    4.789635
> X<-matrix(NA,nrow = 100,ncol=2)
> X[,1]<-1
> X[,2]<-seq(min(iris$Sepal.Width),max(iris$Sepal.Width),length.out = 100)
> X
```



If you run the beta, you can from the fit, you can extract the coefficients. Now, in the beta, I have the coefficient. Now, what I am going to do, I am going to calculate eta X matrix with 100 values with which X 2 takes values with minimum value of Sepal Width to maximum value of Sepal Width.

(Refer Slide Time: 05:04)



```
22
23 beta<-fit$coefficients
24
25 X<-matrix(NA,nrow = 100,ncol=2)
26 X[,1]<-1
27 X[,2]<-seq(min(iris$Sepal.Width),max(iris$Sepal.Width),length.out = 100)
28
29 eta<-X%*%beta
30 p<-exp(eta)/(1+exp(eta))
31 plot(X[,2],p,type="l",lwd=2,xlab="Sepal Width",ylab="Prob(Setosa)")
```

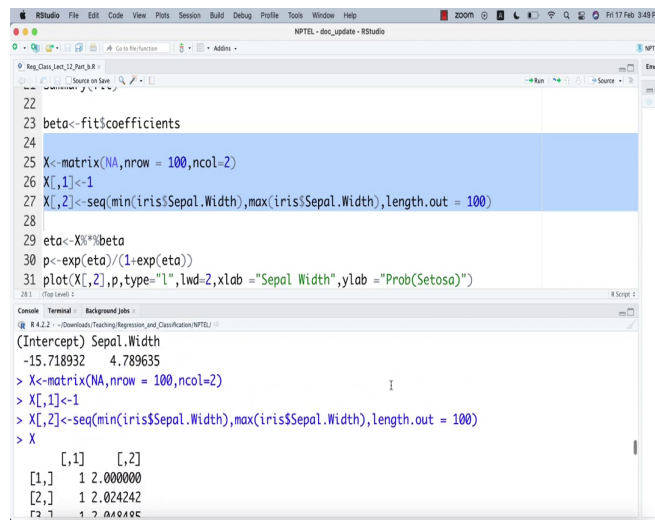
Console

```
[93,] 1 4.230303
[94,] 1 4.254545
[95,] 1 4.278788
[96,] 1 4.303030
[97,] 1 4.327273
[98,] 1 4.351515
[99,] 1 4.375758
[100,] 1 4.400000
```



So, if you just run it. So, minimum is I think 2.2 is the minimum.

(Refer Slide Time: 05:08)



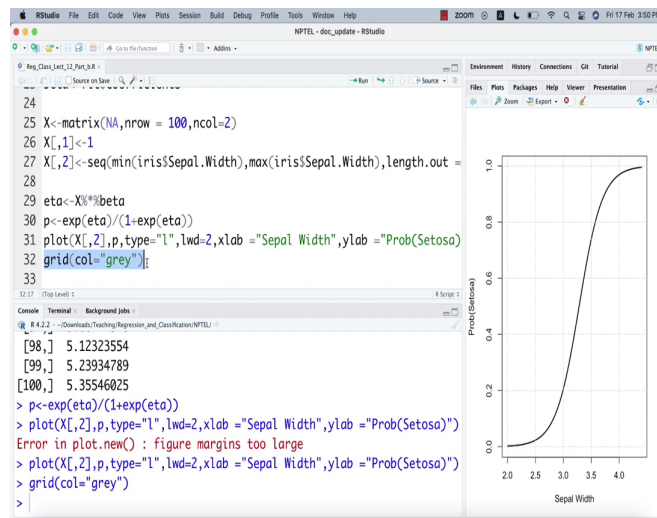
```
22
23 beta<-fit$coefficients
24
25 X<-matrix(NA,nrow = 100,ncol=2)
26 X[,1]<-1
27 X[,2]<-seq(min(iris$Sepal.Width),max(iris$Sepal.Width),length.out = 100)
28
29 eta<-X%*%beta
30 p<-exp(eta)/(1+exp(eta))
31 plot(X[,2],p,type="l",lwd=2,xlab="Sepal Width",ylab="Prob(Setosa)")
```

```
R 4.2.2 ->Downloads\Teaching\Regressor_and_Classification\NPTEL\
(Intercept) Sepal.Width
-15.718932    4.789635
> X<-matrix(NA,nrow = 100,ncol=2)
> X[,1]<-1
> X[,2]<-seq(min(iris$Sepal.Width),max(iris$Sepal.Width),length.out = 100)
> X
      [,1] [,2]
[1,]  1 2.000000
[2,]  1 2.024242
[3,]  1 2.048485
```



And max is the 4.4. And in between, it just fill up with some values in a equal width. And then I calculate the eta, these are the eta values or z values, latent variable values. And then I calculate the probability.

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And then I plot the X 2 versus Sepal Width versus sorry, but that sepal width versus the probability of setosa. So, if the Sepal Width increases, clearly probability that the flower is setosa increases. So, you can put a grid also in this. So, that is how we, this is that is how we can, you can draw probability of p, you can plot p against some predictor values.

So, I will stop here, but this iris data set shows you that your target variable could be not necessarily has to be binary class. It could be multi class variable. And then in that case, you have to you, you have a multi class classification. Because you have three class here. Remember that you have to do three class, not binary class.

You have a setosa, versicolor and virginica, three sub species are there. When you have more than two class, basically it is a k class problem or multi class problem. The most popular and the old and tested method is linear discriminant analysis.

So, in the next video, we are going to start linear discriminant analysis or Fisher's linear discriminant analysis. So, for now, thank you very much for your attention. See you in the next video.