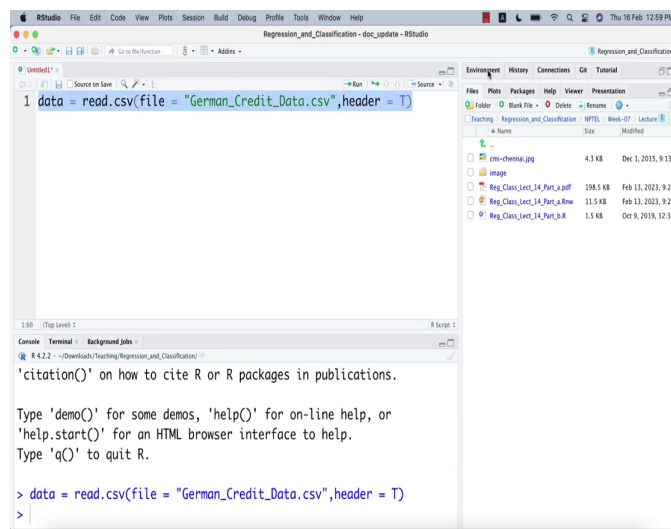


Predictive Analytics - Regression and Classification
Prof. Sourish Das
Department of Mathematics
Chennai Mathematical Institute

Lecture - 38
Hands on with R for Logistic Regression

Hello all, welcome back to the part C of lecture 11. In this part, we are going to do some Hands-on.

(Refer Slide Time: 00:23)



```
1 data = read.csv(file = "German_Credit_Data.csv", header = T)
```

Console Terminal Background Jobs

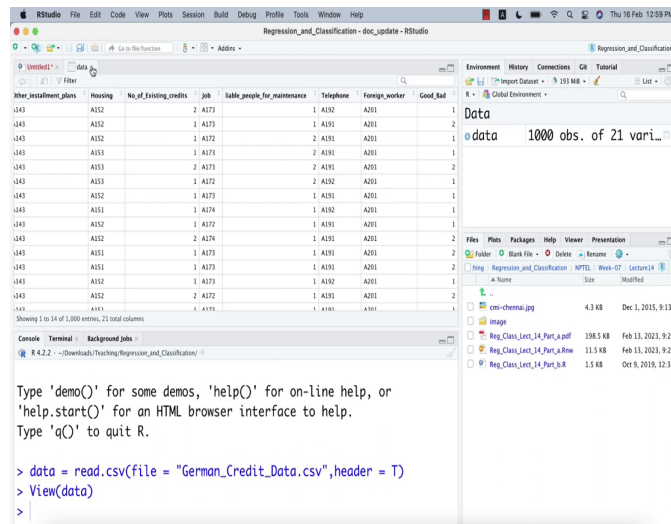
```
R 4.2.2 - ~/Downloads/Teaching/Regression_and_Classification >  
'citation()' on how to cite R or R packages in publications.  
  
Type 'demo()' for some demos, 'help()' for on-line help, or  
'help.start()' for an HTML browser interface to help.  
Type 'q()' to quit R.  
  
> data = read.csv(file = "German_Credit_Data.csv", header = T)  
>
```



So, we are going to call the and this hands-on will be based on R. So, first we are going to read the CSV file. So, we are going to read the German credit code data. So, I have already shared this on the NPTEL platform, Credit also if you do search German credit code data on

internet, you will get this data same data set its a slightly old, but reasonably good data to get your hands dirty with, you know, the first logistic regression model kind of thing.

(Refer Slide Time: 01:36)



The screenshot shows the RStudio interface with a data table and a console window. The data table has the following columns: `file_installment_plan`, `housing`, `no_of_existing_credits`, `job`, `table_people_for_maintenance`, `telephone`, `foreign_worker`, and `good_bud`. The console shows the following code and output:

```
> data = read.csv(file = "German_Credit_Data.csv", header = T)
> View(data)
>
```

The console also displays instructions: "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."



(Refer Slide Time: 01:54)

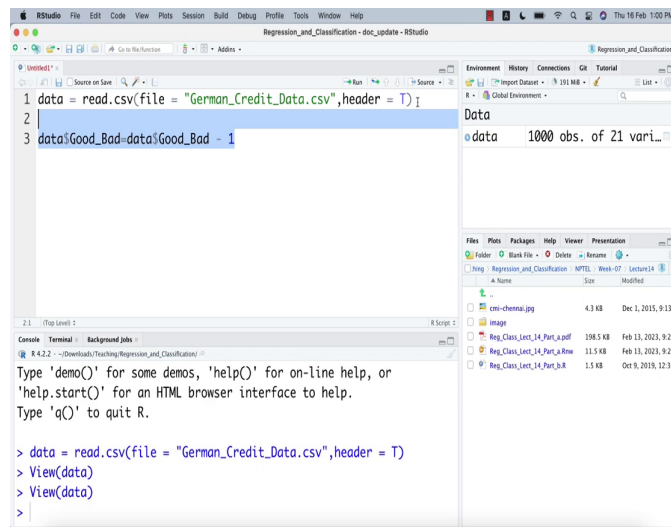
The screenshot displays the RStudio interface. The main window shows a data frame with the following columns: 'Residence', 'No_of_Existing_credits', 'job', 'labile_people_for_maintenance', 'telephone', 'Foreign_worker', and 'Good_loan'. The data is presented in a table format with rows numbered 1443 to 1449. The console window shows the following R code being executed:

```
> data = read.csv(file = "German_Credit_Data.csv", header = T)
> View(data)
> View(data)
>
```



So, yeah, if you just come see there are about 1000 values and there are some status are given and if you look into the status, if you go to and then the target here is good or bad, good loan or bad loan and then essentially I, we have to, it is recorded as 1 or 2.

(Refer Slide Time: 02:02)



The screenshot shows the RStudio interface. The source editor contains the following R code:

```
1 data = read.csv(file = "German_Credit_Data.csv", header = T)
2
3 data$Good_Bad = data$Good_Bad - 1
```

The Environment pane on the right shows a variable named 'data' with 1000 observations and 21 variables.

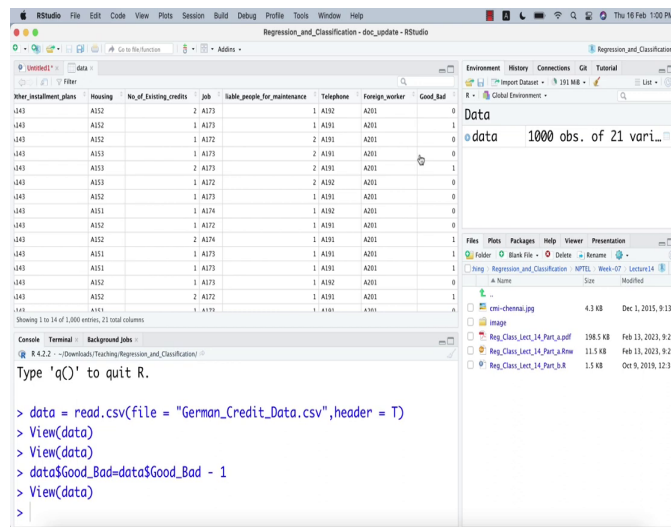
The console shows the following output:

```
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.

> data = read.csv(file = "German_Credit_Data.csv", header = T)
> View(data)
> View(data)
>
```



(Refer Slide Time: 02:29)



The screenshot shows the RStudio interface. The main window displays a data table with columns: `Year`, `Installation_plan`, `Shooting`, `No_of_Existing_credits`, `Job`, `Table_people_for_maintenance`, `Telephone`, `Foreign_worker`, and `Good_Bad`. The console shows the following R code:

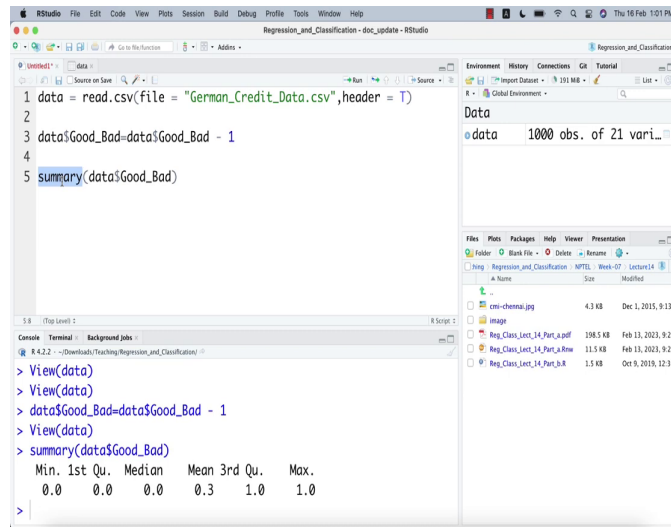
```
> data = read.csv(file = "German_Credit_Data.csv",header = T)
> View(data)
> View(data)
> data$Good_Bad=data$Good_Bad - 1
> View(data)
>
```

The environment pane on the right shows a data object named `data` with 1000 observations and 21 variables. The file explorer on the right shows a folder named `Regression_and_Classification` containing several files.



So, what we will do is data dollar Good Bad equal to, if we just say data dollar Good Bad minus 1. So, that will give us, you see good or bad, if it is 1, then it is bad, 0 means good. So, we want to predict accordingly.

(Refer Slide Time: 02:40)



The screenshot shows the RStudio interface with the following code in the script editor:

```
1 data = read.csv(file = "German_Credit_Data.csv",header = T)
2
3 data$Good_Bad=data$Good_Bad - 1
4
5 summary(data$Good_Bad)
```

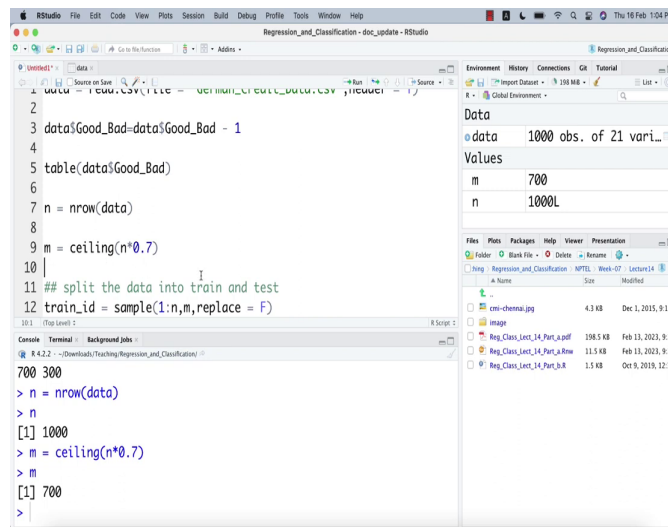
The console output shows the following summary statistics for the variable Good_Bad:

```
> View(data)
> View(data)
> data$Good_Bad=data$Good_Bad - 1
> View(data)
> summary(data$Good_Bad)
  Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
   0.0   0.0   0.0   0.3   1.0   1.0
```

The Environment pane shows the variable 'data' with 1000 observations and 21 variables. The Files pane shows a list of files in the current directory, including 'cni-chemai.jpg', 'image', 'Reg_Class_Lect_14_Part_a.pdf', 'Reg_Class_Lect_14_Part_a.Rnw', and 'Reg_Class_Lect_14_Part_b.R'.



(Refer Slide Time: 03:07)



```
1 data = read.csv(file = "credit_data.csv", header = 1)
2
3 data$Good_Bad = data$Good_Bad - 1
4
5 table(data$Good_Bad)
6
7 n = nrow(data)
8
9 m = ceiling(n*0.7)
10
11 ## split the data into train and test
12 train_id = sample(1:n,m,replace = F)
```

700 300
> n = nrow(data)
> n
[1] 1000
> m = ceiling(n*0.7)
> m
[1] 700
>



Now, if we just do summary, summary of good or bad. So, about 30 percent is bad actually. So, just a minute, I think if I just do table instead of summary, the if I just do a table. Yeah. So, 300 out of 1000 cases, they have 300, they have given as bad cases and 700 given as a good cases. Now, typically in real life, you will not have default more than 2, 3 percent of the entire credit customers.

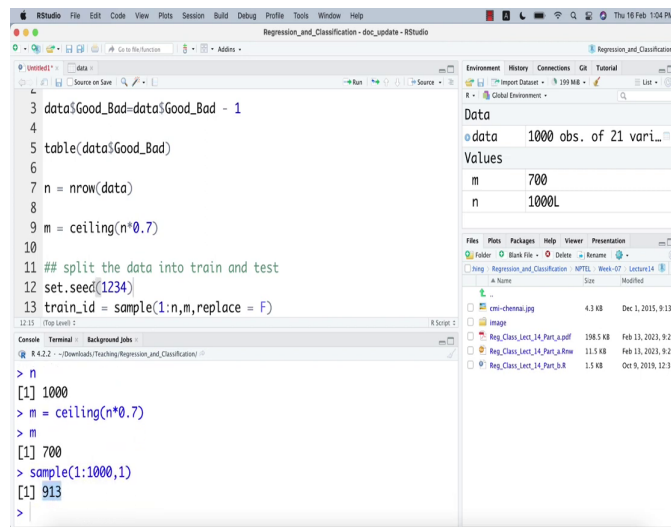
So, this is slightly a artificial in that sense because you cannot, if you are, if your 30 percent of customer is this defaulting, Then you will not be able to you know survive. But anyway, but if you have in your data set, if you have only 2 percent ones and 98 percent zeros, those kind of data sets are called imbalance data sets and in those kind of data sets doing prediction is extremely difficult.

And because most of the time, what happens that any standard model will just say everybody is kind of, you know, good, then no matter what, what is your, what happens 98 percent of the time you are correct. So, and in only 2 percent time you are wrong. So, a naive model will do correct, will be correct 98 percent of time. So, in real.

So, those kind of naive model, we have to be very careful about. So, what I will do first, I will just say number of rows in the data is 1000, I believe yeah. And then I will take m is the testing number of testing, like, you know, n into 0.7, maybe 70 percent of the data we will consider for the testing.

So, 700 out of 1000 700 data points will be considering for testing. So, for we will do a split the data into train and test. First, what we will do? we will choose the train id's sample 1, is to, n out m many samples you will we will draw and replace equal to false. Ok, we have to set up a seed.

(Refer Slide Time: 06:06)



```
3 data$Good_Bad=data$Good_Bad - 1
4
5 table(data$Good_Bad)
6
7 n = nrow(data)
8
9 m = ceiling(n*0.7)
10
11 ## split the data into train and test
12 set.seed(1234)
13 train_id = sample(1:n,m,replace = F)
```

Environment History Connections GR Tutorial
R 4.2.2 939 MB Global Environment

Data
data 1000 obs. of 21 vari_

Values
m 700
n 1000L

Files Packages Help Viewer Presentation
Folder Blank File Delete Rename
Reg_Regression_and_Classification NPTEL Week-07 Lecture14

File Name Size Modified
cmi-chenai.jpg 4.3 KB Dec 1, 2015, 9:13 A
image
Reg_Class_Lect_14_Part_a.pdf 108.5 KB Feb 11, 2023, 9:26
Reg_Class_Lect_14_Part_a.Rnw 11.5 KB Feb 11, 2023, 9:26
Reg_Class_Lect_14_Part_b.R 1.4 KB Oct 9, 2019, 12:39 P

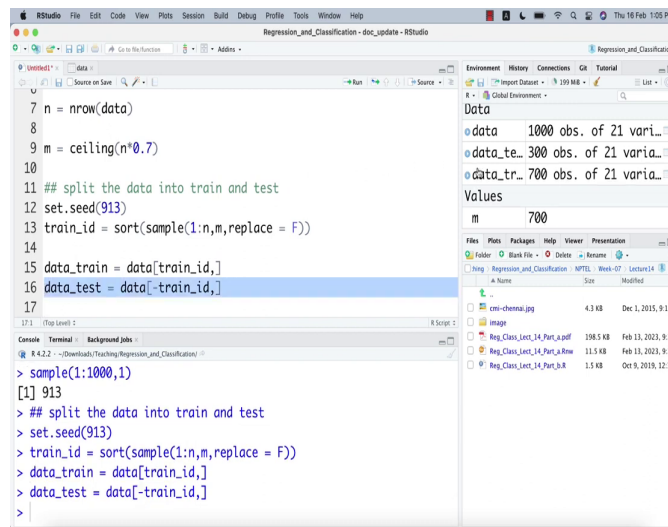
Console Terminal Background Jobs
R 4.2.2 - (Downloads/Teaching/Regression_and_Classification)

```
> n
[1] 1000
> m = ceiling(n*0.7)
> m
[1] 700
> sample(1:1000,1)
[1] 913
>
```



Set dot, seed is 1234. So, the advantage of using seed is if you set up a seed and if you use exactly this number 1234 actually you can do any number, I can just draw a sample between 1, is to 1000, one sample and 913, I am let me just use 913 as my seed.

(Refer Slide Time: 06:34)



```
7 n = nrow(data)
8
9 m = ceiling(n*0.7)
10
11 ## split the data into train and test
12 set.seed(913)
13 train_id = sort(sample(1:n,m,replace = F))
14
15 data_train = data[train_id,]
16 data_test = data[-train_id,]
17
```

Environment: Global Environment

Data

- data: 1000 obs. of 21 vari...
- data_te: 300 obs. of 21 varia...
- data_tr: 700 obs. of 21 varia...

Values

m: 700

Files: cmi-chenai.jpg (4.3 KB, Dec 1, 2015, 9:13 A), image, Reg_Class_Lect_14_Part_a.pdf (108.5 KB, Feb 11, 2022, 9:26), Reg_Class_Lect_14_Part_a.Rnw (11.5 KB, Feb 11, 2022, 9:26), Reg_Class_Lect_14_Part_b.R (1.4 KB, Oct 9, 2019, 12:39)

Console:

```
> sample(1:1000,1)
[1] 913
> ## split the data into train and test
> set.seed(913)
> train_id = sort(sample(1:n,m,replace = F))
> data_train = data[train_id,]
> data_test = data[-train_id,]
>
```



Now, that means, if you use this seed number as your seed, then what will happen is you, you my result and your result will exactly match ok. So, this ensures because we are going to split the data set randomly 70 percent data we will take as a training data, ok. So, data underscore train equal to what I am going to do data. Just I will provide the training id comma maybe I will just do a sorting here ok.

Let me just take this and then I will just data train and data test equal to what I will do. I will just copy this and instead of train id, I will just say minus train id. So, just take all the ids, but do not take the training ids.

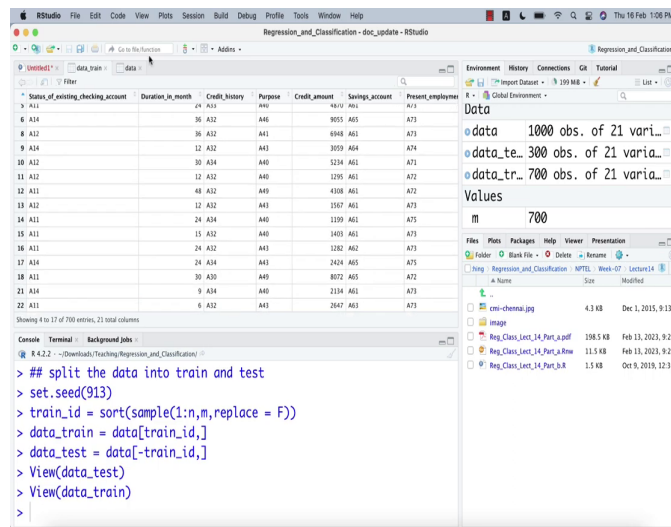
(Refer Slide Time: 07:40)

The screenshot shows the RStudio interface. The main window displays a data table with the following columns: `Status_of_existing_checking_account`, `Duration_in_months`, `Credit_history`, `Purpose`, `Credit_Lamount`, `Savings_account`, and `Present_employm`. The data is sorted by `Present_employm`. The environment pane on the right shows the `data` object with 1000 observations of 21 variables. The console at the bottom contains the following R code:

```
> ## split the data into train and test
> set.seed(913)
> train_id = sort(sample(1:n,m,replace = F))
> data_train = data[train_id,]
> data_test = data[-train_id,]
> View(data_test)
> View(data_train)
>
```



(Refer Slide Time: 07:57)



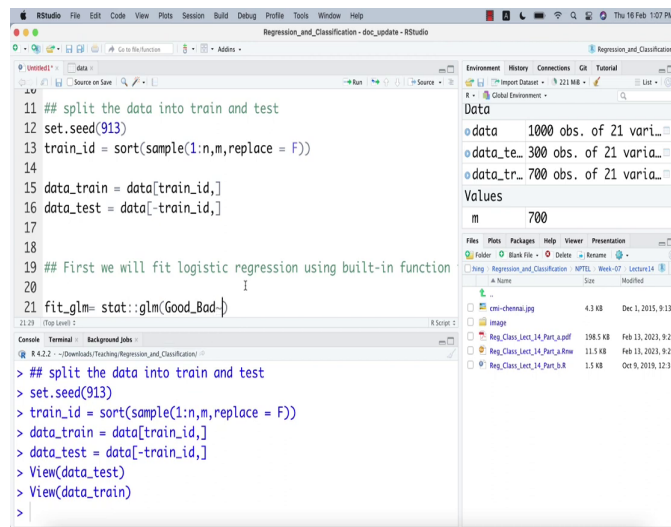
The screenshot shows the RStudio interface. The top pane displays a data table with columns: Status_of_existing_checking_account, Duration_in_months, Credit_history, Purpose, Credit_amount, Savings_account, and Present_employment. The bottom pane shows the following R code:

```
> ## split the data into train and test
> set.seed(913)
> train_id = sort(sample(1:n,m,replace = F))
> data_train = data[train_id,]
> data_test = data[-train_id,]
> View(data_test)
> View(data_train)
>
```



So, that will automatically give me the testing data. And so, row number 1, 7, 19, 20, 26, this came to the test data. And if you go to the training data, 2, 3, 4, 5, 6, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17, 18, then 19 missing 21. So, those are the train data, ok. So, now we have splitted the data. So, first we will fit a logistic regression using built-in function in R ok.

(Refer Slide Time: 08:19)



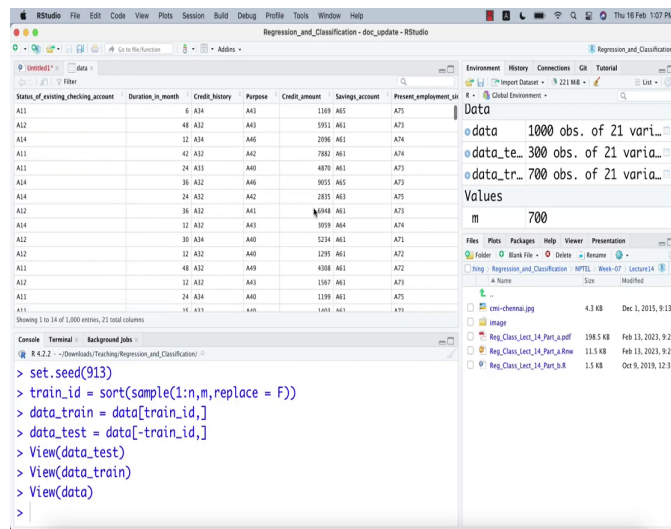
```
11 ## split the data into train and test
12 set.seed(913)
13 train_id = sort(sample(1:n,m,replace = F))
14
15 data_train = data[train_id,]
16 data_test = data[-train_id,]
17
18
19 ## First we will fit logistic regression using built-in function
20
21 fit_glm= stat::glm(Good_Bad~
```

The screenshot shows the RStudio interface. The main editor window contains R code for splitting data into training and testing sets and fitting a logistic regression model. The console window shows the execution of these commands. The Environment pane on the right shows the objects created: 'data' (1000 obs. of 21 variables), 'data_train' (300 obs. of 21 variables), and 'data_test' (700 obs. of 21 variables). The 'Values' pane shows the value of 'm' as 700. The file explorer shows a folder named 'Regression_and_Classification' with several files.



First we will fit logistic regression using built-in function built-in function in R ok. Now, first fit glm. So, from the stats package, you extract the glm function, you call the glm function. So, you just say Good Bad and dollar.

(Refer Slide Time: 09:14)



The screenshot shows the RStudio interface. The main window displays a data table with columns: State_of_existing_checking_account, Duration_in_month, Credit_history, Purpose, Credit_amount, Savings_account, and Present_employment_status. The console shows the following R code:

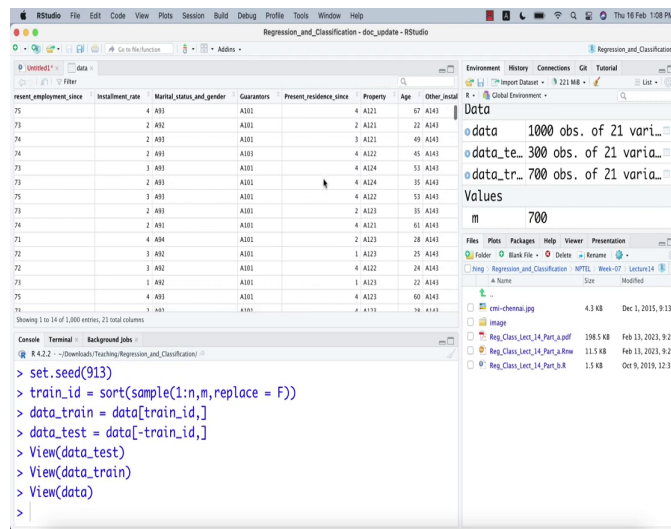
```
> set.seed(913)
> train_id = sort(sample(1:n,m,replace = F))
> data_train = data[train_id,]
> data_test = data[-train_id,]
> View(data_test)
> View(data_train)
> View(data)
>
```

The Environment pane on the right shows the 'data' object with 1000 observations of 21 variables. The 'Values' pane shows the variable 'm' with the value 700. The File pane shows a folder named 'Regression_and_Classification' containing several files.



And I am going to take two in this data. There are most of these data are categorical, but I just want to keep it little simple. So, that, you know, you understand it more care more easily.

(Refer Slide Time: 09:40)



The screenshot displays the RStudio environment. The main window shows a data table with columns: `heart_employment_size`, `localism_rate`, `marital_status_and_gender`, `coastlines`, `present_residence_size`, `property`, `age`, and `other_income`. The console shows the following R code:

```
> set.seed(913)
> train_id = sort(sample(1:n,m,replace = F))
> data_train = data[train_id,]
> data_test = data[-train_id,]
> View(data_test)
> View(data_train)
> View(data)
>
```

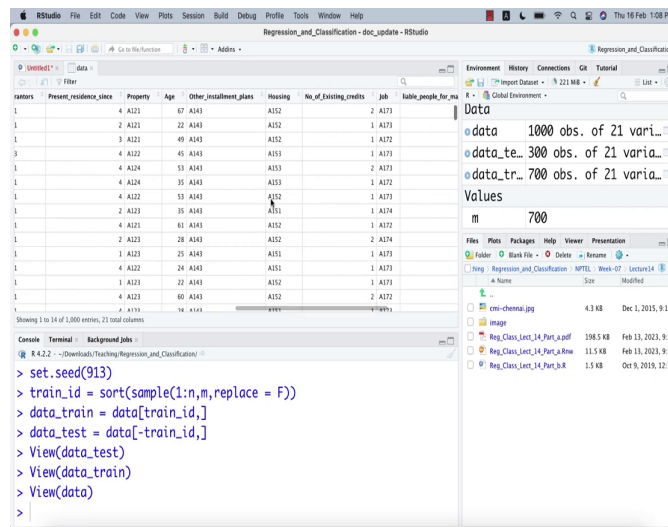
The Environment pane on the right shows the following data objects:

- `data`: 1000 obs. of 21 variables
- `data_train`: 300 obs. of 21 variables
- `data_test`: 700 obs. of 21 variables
- `m`: 700

The Files pane shows a folder named `Regression_and_Classification` containing several files, including `cmi-chennai.jpg`, `image`, `Reg_Class_Lect_14_Part_a.pdf`, `Reg_Class_Lect_14_Part_a.Rnw`, and `Reg_Class_Lect_14_Part_b.R`.



(Refer Slide Time: 09:44)



The screenshot shows the RStudio interface. The top pane displays a data frame with columns: Present_residence_months, Property_Age, Other_installment_months, Housing, No_of_existing_credits, job, and label_people_for_mortgage. The middle pane shows the R console with the following code:

```
> set.seed(913)
> train_id = sort(sample(1:n,m,replace = F))
> data_train = data[train_id,]
> data_test = data[-train_id,]
> View(data_test)
> View(data_train)
> View(data)
>
```

The right pane shows the Environment tab with the following details:

- data: 1000 obs. of 21 variables
- data_train: 300 obs. of 21 variables
- data_test: 700 obs. of 21 variables
- Values: m = 700

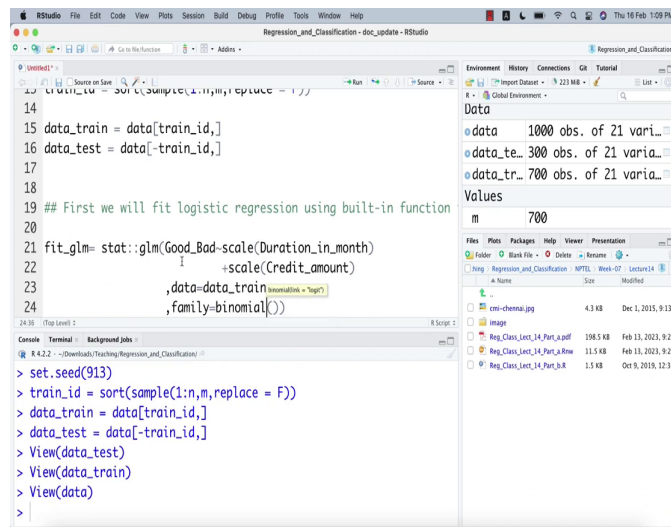
The bottom right pane shows a file explorer with files like cmi-chennai.jpg, image, Reg_Class_Lect_14_Part_a.pdf, Reg_Class_Lect_14_Part_a.Rnw, and Reg_Class_Lect_14_Part_b.R.



The duration in month is a sort of a numeric value, you know, yeah, credit amount and duration in month. These are the 2 columns, which are in numeric. There are some other numeric also, installment rate, present, residence, since age. So, these also can be, I can take, but just to be, my purpose of this demo is to give you how the logistic regression works, ok.

You can add as many predictors as you want. There are about 20, 19 or 20 predictors are there in this data set. But you can add as many predictors as you want. My purpose is to just give you how the whole logistic regression works in R.

(Refer Slide Time: 10:20)



```
14
15 data_train = data[train_id,]
16 data_test = data[-train_id,]
17
18
19 ## First we will fit logistic regression using built-in function
20
21 fit_glm= stat::glm(Good_Bad~scale(Duration_in_month)
22                   +scale(Credit_amount)
23                   ,data=data_train,method="logit")
24                   ,family=binomial())
```

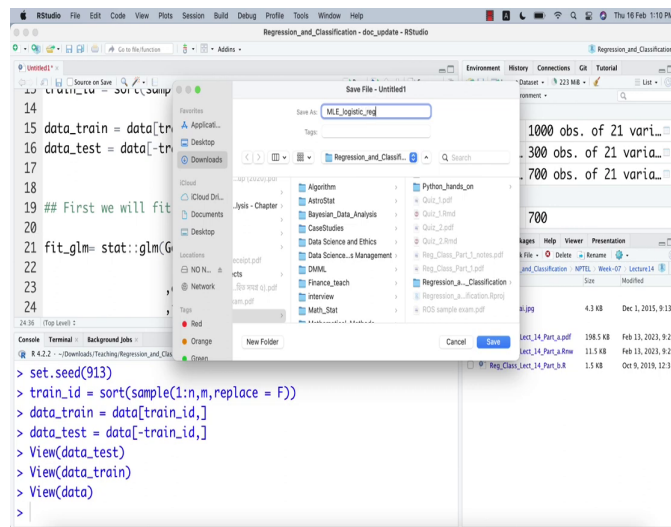
```
> set.seed(913)
> train_id = sort(sample(1:n,m,replace = F))
> data_train = data[train_id,]
> data_test = data[-train_id,]
> View(data_test)
> View(data_train)
> View(data)
>
```

The screenshot shows the RStudio interface. The main editor window contains R code for data splitting and fitting a logistic regression model. The console window shows the execution of these commands. The Environment pane on the right shows the objects 'data' (1000 obs. of 21 variables), 'data_train' (700 obs. of 21 variables), and 'data_test' (300 obs. of 21 variables). The 'Values' pane shows the value of 'm' as 700. The Files pane shows a list of files in the current directory.



And what I am going to do, I am just going to take 2 predictor, which are numeric predictor. And what I am going to do, I am going to just scale. This scale always helps Duration in month and plus scale Credit amount credit amount and then data equal to data train.

(Refer Slide Time: 11:26)



```
14
15 data_train = data[train_id,]
16 data_test = data[-train_id,]
17
18
19 ## First we will fit
20
21 fit_glm = stat::glm(G)
22
23
24
```

```
> set.seed(913)
> train_id = sort(sample(1:n,m,replace = F))
> data_train = data[train_id,]
> data_test = data[-train_id,]
> View(data_test)
> View(data_train)
> View(data)
>
```



And now here is the important part that you have to mention family, family equal to binomial, because remember that it is a binary classification model. So, it raised into binomial I have to save it somewhere. Let me just call it. So, I will just call it maybe, MLE logistic regression logistic regression, ok binomial. So, since it is a binary classification problem, in the binary classification problem, the model automatically raise up what it comes is a binomial model or Bernoulli model.

(Refer Slide Time: 12:07)

```
15 data_train = data[train_id,]
16 data_test = data[-train_id,]
17
18
19 # First we will fit logistic regression using built-in function
20
21 fit_glm= stats::glm(Good_Bad~scale(Duration_in_month)
22                    +scale(Credit_amount)
23                    ,data=data_train
24                    ,family=binomial(link = "logit"))
25
```

```
> fit_glm= stats::glm(Good_Bad~scale(Duration_in_month)
+                    +scale(Credit_amount)
+                    ,data=data_train
+                    ,family=binomial(link = "logit"))
Error in loadNamespace(x) : there is no package called 'stat'
>
```



(Refer Slide Time: 12:39)

```
16 data_test = data[-train_id,]
17
18
19 # First we will fit logistic regression using built-in function
20
21 fit_glm= stats::glm(Good_Bad~scale(Duration_in_month)
22                   +scale(Credit_amount)
23                   ,data=data_train
24                   ,family=binomial(link = "logit"))
25
26 summary(fit_glm)
```

Null deviance: 855.21 on 699 degrees of freedom
Residual deviance: 817.75 on 697 degrees of freedom
AIC: 823.75

Number of Fisher Scoring iterations: 4



(Refer Slide Time: 12:49)

```
Call:
stats::glm(formula = Good_Bad ~ scale(Duration_in_month) + scale(Credit_amount),
            family = binomial(link = "logit"), data = data_train)

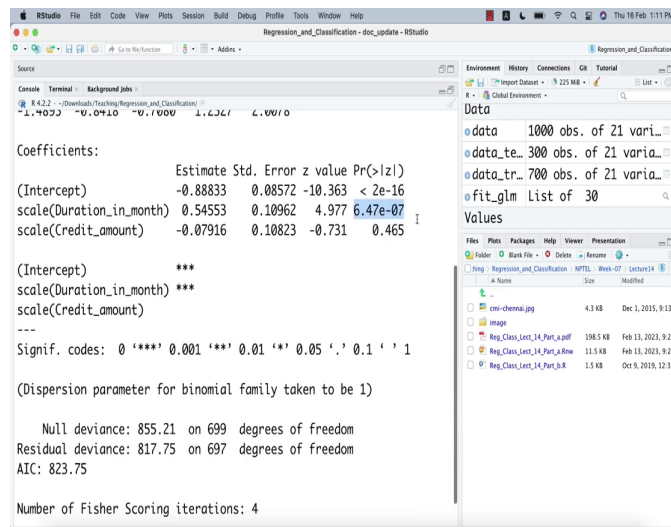
Deviance Residuals:
    Min       1Q   Median       3Q      Max
-1.4895  -0.8418  -0.7080   1.2527   2.0078

Coefficients:
            Estimate Std. Error z value Pr(>|z|)
(Intercept)  -0.88833    0.08572  -10.363 < 2e-16
scale(Duration_in_month)  0.54553    0.10962   4.977 6.47e-07
scale(Credit_amount)    -0.107916    0.10823  -0.731  0.465

(Intercept) ***
scale(Duration_in_month) ***
scale(Credit_amount)
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```



(Refer Slide Time: 12:52)



```
RStudio File Edit Code View Plots Session Build Debug Profile Tools Window Help
Regression_and_Classification - dsc_update - RStudio

Source
Console Terminal Background jobs
R 4.2.2 -> (Downloads) TeachingRegression_and_Classification
~1.4032 ~0.0+10 ~0.1000 1.232/ 2.00/0

Coefficients:
              Estimate Std. Error z value Pr(>|z|)
(Intercept)   -0.88833    0.08572 -10.363 < 2e-16
scale(Duration_in_month)  0.54553    0.10962   4.977 6.47e-07
scale(credit_amount) -0.07916    0.10823  -0.731  0.465

(Intercept)      ***
scale(Duration_in_month) ***
scale(credit_amount)
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for binomial family taken to be 1)

Null deviance: 855.21 on 699 degrees of freedom
Residual deviance: 817.75 on 697 degrees of freedom
AIC: 823.75

Number of Fisher Scoring iterations: 4
```

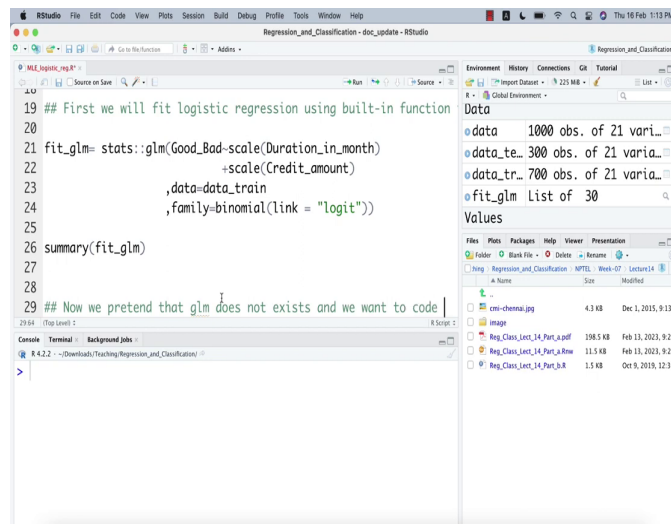


Now, I am going to say summary, fit glm ok. So, it what it says that duration in month and credit amount, both of duration in month has a statistically significant m effect in the, in the whether a probe, customer is going to be a good customer or bad customer.

And turns out credit amount is not that important. You see, p value is very large, but credit amount is important. So, this is the estimate, the coefficient. These are the standard error and z value. So, you can have these, you can create these information's. And now what I am going to do is now we are going to pretend that as if glm function in our does not exist and we want to code it from scratch ok. So, we are going to write down the negative log likelihood function.

And then we will use the optimization subroutine technique and we will find what estimates of the coefficients are. And what these estimates are going to be match going to match with built in glm function ok. So, we will first write negative log likelihood function ok, alright.

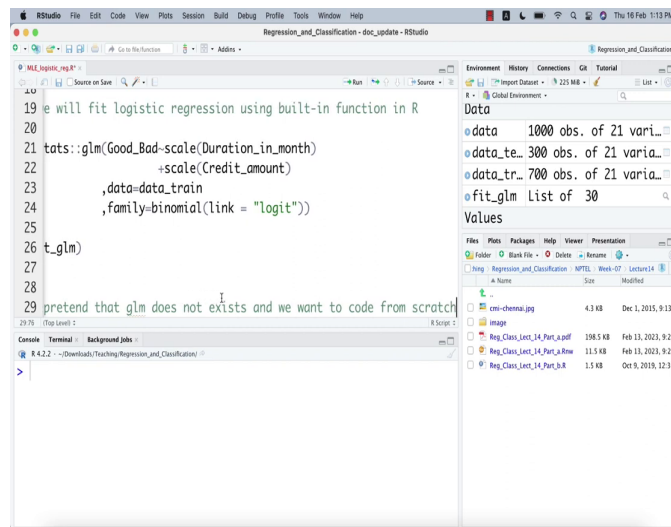
(Refer Slide Time: 14:21)



```
19 # First we will fit logistic regression using built-in function
20
21 fit_glm= stats::glm(Good_Bad~scale(Duration_in_month)
22                   +scale(Credit_amount)
23                   ,data=data_train
24                   ,family=binomial(link = "logit"))
25
26 summary(fit_glm)
27
28
29 ## Now we pretend that glm does not exists and we want to code
```



(Refer Slide Time: 14:54)



```
19 # we will fit logistic regression using built-in function in R
20
21 # data: glm(Good_Bad~scale(Duration_in_month)
22 #         +scale(Credit_amount)
23 #         ,data=data_train
24 #         ,family=binomial(link = "logit"))
25
26 # fit_glm
27
28
29 # pretend that glm does not exist and we want to code from scratch
```

Environment: R 4.2.2 (64-bit) x86_64-pc-linux-gnu

Data:

- data: 1000 obs. of 21 variables
- data_train: 300 obs. of 21 variables
- data_test: 700 obs. of 21 variables
- fit_glm: List of 30

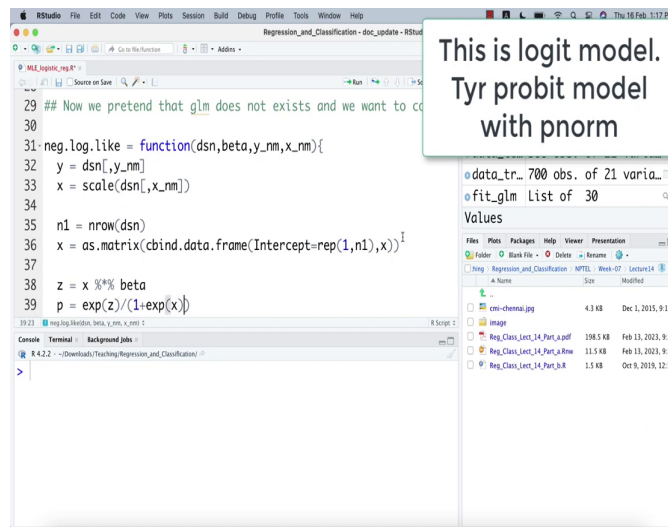
Files:

- cmi-chennai.jpg: 4.3 KB, Dec 1, 2015, 9:13 A
- image
- Reg_Class_Lect_14_Part_a.pdf: 108.5 KB, Feb 11, 2022, 9:26
- Reg_Class_Lect_14_Part_a.Rnw: 11.5 KB, Feb 11, 2022, 9:26
- Reg_Class_Lect_14_Part_b.R: 1.4 KB, Oct 9, 2019, 12:39



Let me first, let me just write a point here that: Now we will we pretend that glm does not exist exists and we want to code from scratch ok.

(Refer Slide Time: 15:10)



```
29 ## Now we pretend that glm does not exist and we want to code it from scratch
30
31 neg.log.like = function(dsn,beta,y_nm,x_nm){
32   y = dsn[,y_nm]
33   x = scale(dsn[,x_nm])
34
35   n1 = nrow(dsn)
36   x = as.matrix(cbind(data.frame(Intercept=rep(1,n1),x))
37
38   z = x %*% beta
39   p = exp(z)/(1+exp(x))
40 }
```

data_tr... 700 obs. of 21 varia...
fit_glm List of 30
Values



Now, we want to code from scratch. Now, first, how do you do that? Negative log likelihood you have to write the negative log likelihood function. So, there will be all the statements inside and then everything is first dataset name we have to give then beta values we have to give then y name we have to give and the x names we have to give correct, alright. So, the first from the we have to extract y from the dsn y name and keep it in x and then x equal to scale dsn x name.

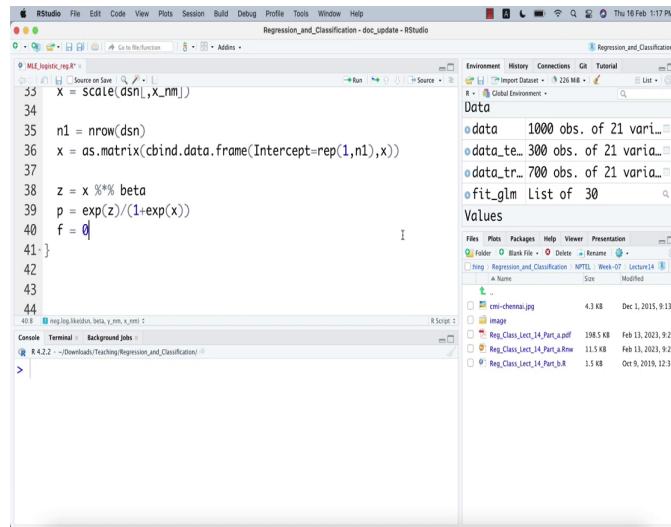
So, the way I am doing it at this moment it will not work for categorical variable. So, it will work only for the numerical variable when writing the full blown code is not my purpose. And my purpose is to give you a demo that how if require you can write your own model you can develop your own model write down the likelihood function, pass it through the

optimization sub routine and feed your own model and hey, here is your own model isn't it fun?

So, what I am going to do is first is $n1$ is $nrow$ of dsn and then I am going to define it as remember that most likely dsn is going to be a data frame not most likely we are going to pass it as a data frame, but we need to do a matrix operation. Since we need to do a matrix operation what we need to do we have to call it as matrix this x as dot matrix. And we have to do this $cbind$ dot data dot frame with x and before that Intercept equal to replicate of 1 comma $n1$.

So, we have to create a x matrix with all the x columns that has a name of the x columns with all the predictors name and a column, which will be intercept right. Now then once I have the x what I have to do I have to calculate for z remember that in the theory part we have a z value latent variable z . So, this is the latent variable I am going to introduce here. Percentage star percentage beta and then p equal to e to the power z divided by 1 plus e to the power z .

(Refer Slide Time: 18:50)



```
33 x = scale(dsn[,x_nm])
34
35 n1 = nrow(dsn)
36 x = as.matrix(cbind.data.frame(Intercept=rep(1,n1),x))
37
38 z = x %*% beta
39 p = exp(z)/(1+exp(x))
40 f = 0
41 }
42
43
44
```

Environment History Connections GR Tutorial
Global Environment

Data

- data 1000 obs. of 21 vari...
- data_te_ 300 obs. of 21 varia...
- data_tr_ 700 obs. of 21 varia...
- fit_glm List of 30

Values

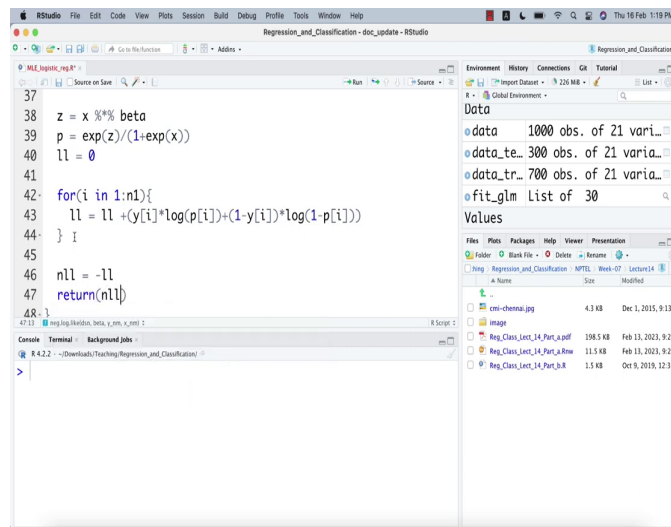
File Edit Packages Help View Presentation
Folder Blank File Delete Rename
Reg Regression_and_Classification NPTEL Week-07 Lecture14

A Name	Size	Modified
..		
cmi-chenai.jpg	4.3 KB	Dec 1, 2015, 9:13 A
image		
Reg_Class_Lect_14_Part_a.pdf	108.5 KB	Feb 11, 2022, 9:26
Reg_Class_Lect_14_Part_a.Rnw	11.5 KB	Feb 11, 2022, 9:26
Reg_Class_Lect_14_Part_b.R	1.4 KB	Oct 9, 2019, 12:39

Console Terminal Background Jobs
R 4.2.2 - [Download/Teaching/Regression_and_Classification]



(Refer Slide Time: 18:55)



```
37
38 z = x %>% beta
39 p = exp(z)/(1+exp(x))
40 ll = 0
41
42 for(i in 1:n1){
43   ll = ll +(y[i]*log(p[i])+(1-y[i])*log(1-p[i]))
44 } i
45
46 nll = -ll
47 return(nll)
48 }
```



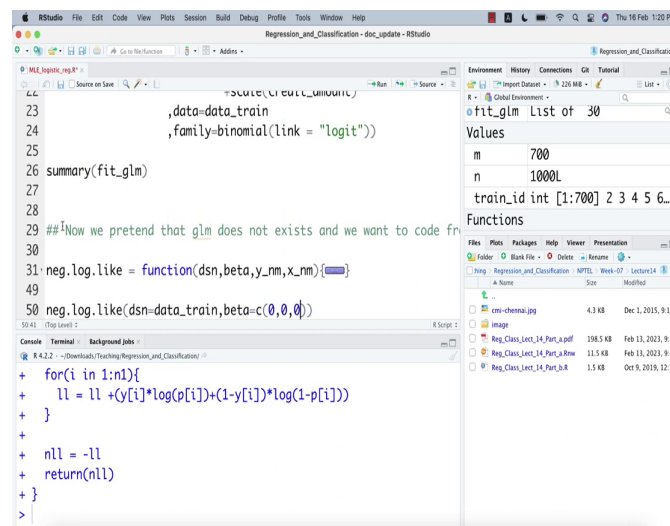
Now, I have the p wonderful now what I have to do is f equal to 0 I have to calculate the log likelihood actually I should say log likelihood ll equal to 0. For I could have done it slightly better way in a some kind of you know more computed efficient way, but my purpose is to show you not a write a write an extremely efficient way of writing the code when my purpose to show you how the concept of writing negative log likelihood.

And because the best thing is if I could have write this three lines of code in a vectorized form, but instead of writing it in a vectorized form I am purposefully writing it as in a loop so, that you can understand what I am doing. So, plus now y i times log p i plus 1 minus y i into log of 1 minus p i ok.

Now, here is a here is a task for you ok. Write this code three lines of code without for loop, can you do that and check if do the, if you do that whether your overall optimization time is

going to be reduced or not. So, let me just let me just come finally, write negative log likelihood will be just minus of log likelihood. So, this is my log likelihood and just and then return negative log likelihood ok.

(Refer Slide Time: 21:34)



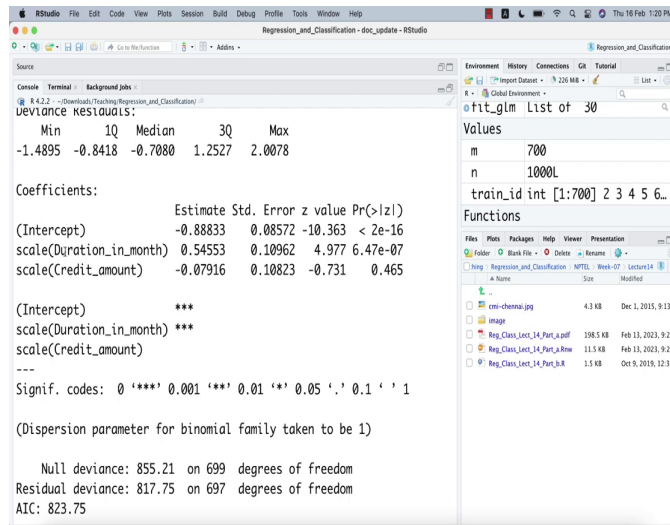
```
23     ,data=data_train
24     ,family=binomial(link = "logit"))
25
26 summary(fit_glm)
27
28
29 ##!Now we pretend that glm does not exists and we want to code fr
30
31 neg.log.like = function(dsn,beta,y_nm,x_nm){
49
50 neg.log.like(dsn=data_train,beta=c(0,0,0))
```

The screenshot shows the RStudio interface. The main editor window contains R code for fitting a binomial GLM and defining a custom negative log likelihood function. The console window shows the execution of the function with initial parameters. The environment pane shows the 'fit_glm' object with values for 'm' (700), 'n' (1000L), and 'train_id' (int [1:700] 2 3 4 5 6...). The file explorer shows a folder named 'Regression_and_Classification' with several files.



Now, once I have written it once I have written the negative log likelihood at least we have to check whether the value the function works for some data set or not. So, some value some initial value dsn will be data underscore train then beta will be 0 comma 0 comma 0. So, why is 0 comma 0 comma 0? Because you see I have if I have in the model that I am trying to fit this model I am going to fit.

(Refer Slide Time: 22:08)



```
fit_glm List of 30
Values
 m      700
 n     1000L
train_id int [1:700] 2 3 4 5 6...
Functions
```

Deviance residuals:

Min	1Q	Median	3Q	Max
-1.4895	-0.8418	-0.7080	1.2527	2.0078

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)
(Intercept)	-0.88833	0.08572	-10.363	< 2e-16
scale(Duration_in_month)	0.54553	0.10962	4.977	6.47e-07
scale(Credit_amount)	-0.07916	0.10823	-0.731	0.465

(Intercept) ***
scale(Duration_in_month) ***
scale(Credit_amount)

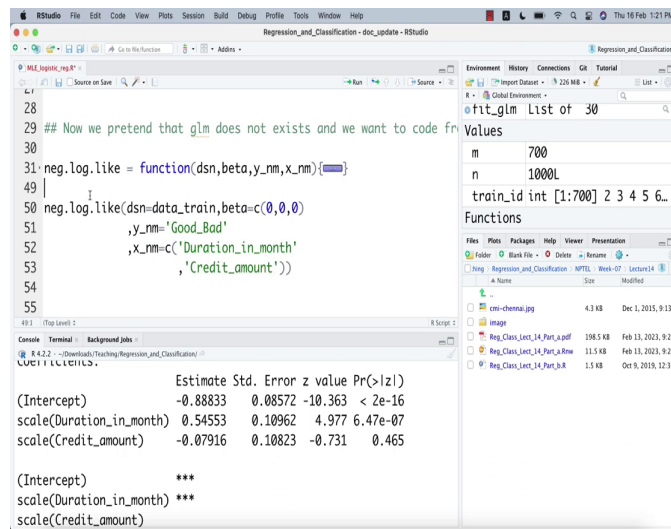
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for binomial family taken to be 1)

Null deviance: 855.21 on 699 degrees of freedom
Residual deviance: 817.75 on 697 degrees of freedom
AIC: 823.75



(Refer Slide Time: 22:24)



```
28
29 ## Now we pretend that glm does not exist and we want to code fr
30
31 neg.log.like = function(dsn,beta,y_nm,x_nm){
49
50 neg.log.like(dsn=data_train,beta=c(0,0,0)
51             ,y_nm='Good_Bad'
52             ,x_nm=c('Duration_in_month'
53                   ,'credit_amount'))
54
55
```

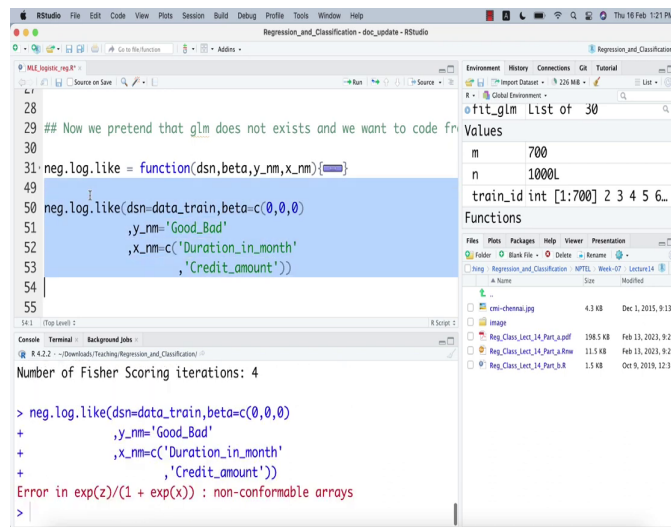
```
Estimate Std. Error z value Pr(>|z|)
(Intercept) -0.88833 0.08572 -10.363 < 2e-16
scale(Duration_in_month) 0.54553 0.10962 4.977 6.47e-07
scale(credit_amount) -0.07916 0.10823 -0.731 0.465

(Intercept) ***
scale(Duration_in_month) ***
scale(credit_amount)
```



So, intercept one coefficient for Duration in month another coefficient Credit in amount. So, these are three coefficients for all three coefficients I am going to give initial value of 0. And then I have to give y name y name y name is in a vector form sorry y name is good or bad Good Bad and x name actually I can just copy these names Duration in month. Because that is the number exactly what we have and Credit amount ok.

(Refer Slide Time: 23:09)



```
28
29 ## Now we pretend that glm does not exist and we want to code fr
30
31 neg.log.like = function(dsn,beta,y_nm,x_nm){
49
50 neg.log.like(dsn=data_train,beta=c(0,0,0)
51             ,y_nm='Good_Bad'
52             ,x_nm=c('Duration_in_month'
53                   ,'Credit_amount'))
54
55
```

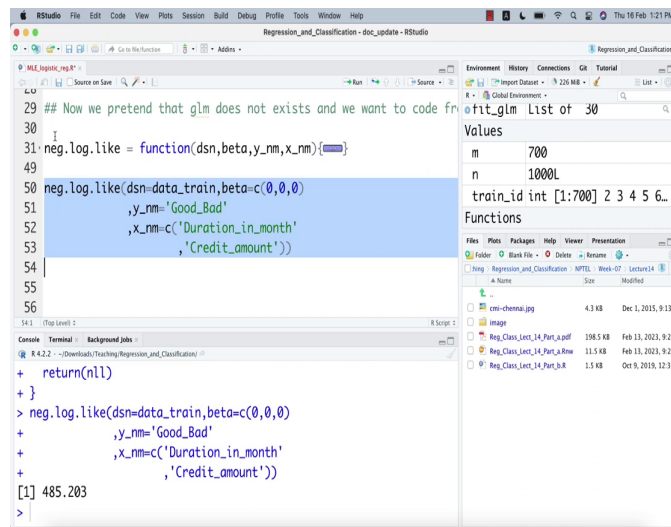
```
Number of Fisher Scoring iterations: 4

> neg.log.like(dsn=data_train,beta=c(0,0,0)
+             ,y_nm='Good_Bad'
+             ,x_nm=c('Duration_in_month'
+                   ,'Credit_amount'))
Error in exp(z)/(1 + exp(x)) : non-conformable arrays
>
```



So, if I just run this these are the initial values this are called non-conformable let me see of course, because it should be z. So, there was a mistake alright.

(Refer Slide Time: 23:27)

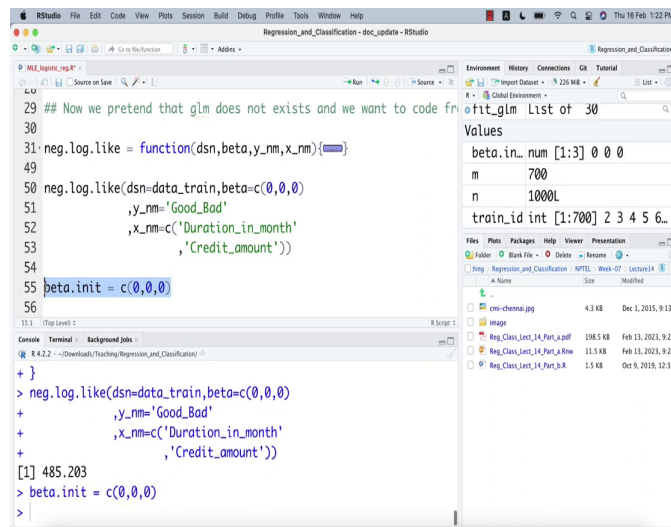


```
29 ## Now we pretend that glm does not exist and we want to code from scratch
30
31 neg.log.like = function(dsn,beta,y_nm,x_nm){
49
50 neg.log.like(dsn=data_train,beta=c(0,0,0)
51             ,y_nm='Good_Bad'
52             ,x_nm=c('Duration_in_month'
53                   ,'Credit_amount'))
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```

```
+ return(nll)
+ }
> neg.log.like(dsn=data_train,beta=c(0,0,0)
+             ,y_nm='Good_Bad'
+             ,x_nm=c('Duration_in_month'
+                   ,'Credit_amount'))
+
+ [1] 485.203
>
```



(Refer Slide Time: 23:44)



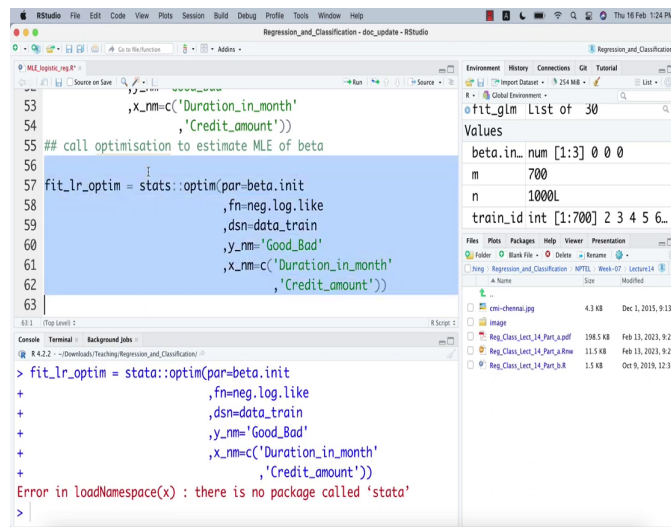
```
29 ## Now we pretend that glm does not exist and we want to code from scratch
30
31 neg.log.like = function(dsn,beta,y_nm,x_nm){
49
50 neg.log.like(dsn=data_train,beta=c(0,0,0)
51             ,y_nm='Good_Bad'
52             ,x_nm=c('Duration_in_month'
53                   , 'Credit_amount'))
54
55 beta.init = c(0,0,0)
56
57 }
58
59 > neg.log.like(dsn=data_train,beta=c(0,0,0)
60             ,y_nm='Good_Bad'
61             ,x_nm=c('Duration_in_month'
62                   , 'Credit_amount'))
[1] 485.203
63 > beta.init = c(0,0,0)
64 >
```

The screenshot shows the RStudio interface. The source editor contains R code defining a function `neg.log.like` and calling it with `beta=c(0,0,0)`. The console shows the output of the function call, which is `[1] 485.203`. The environment pane on the right shows the values of `beta.in_num` (a 1x3 matrix of zeros), `m` (700), `n` (1000L), and `train_id` (a vector of integers).



Let me run it once more yup. So, in the first initial value the negative log likelihood function runs and it gives me a value of the log negative log likelihood now I am going to initiate the initial value of beta I have to give initial value as 0, 0, 0 I have given here maybe I will just run it here and I will do it there.

(Refer Slide Time: 24:27)



```
53     ,x_nm=c('Duration_in_month'
54           , 'Credit_amount'))
55 # call optimisation to estimate MLE of beta
56
57 fit_lr_optim = stats::optim(par=beta.init
58                            ,fn=neg.log.like
59                            ,dsn=data_train
60                            ,y_nm='Good_Bad'
61                            ,x_nm=c('Duration_in_month'
62                                  , 'Credit_amount'))
```

Environment History Connections GR Tutorial
Global Environment
fit_glm List of 30
Values
beta.in_ num [1:3] 0 0 0
m 700
n 1000L
train_id int [1:700] 2 3 4 5 6...

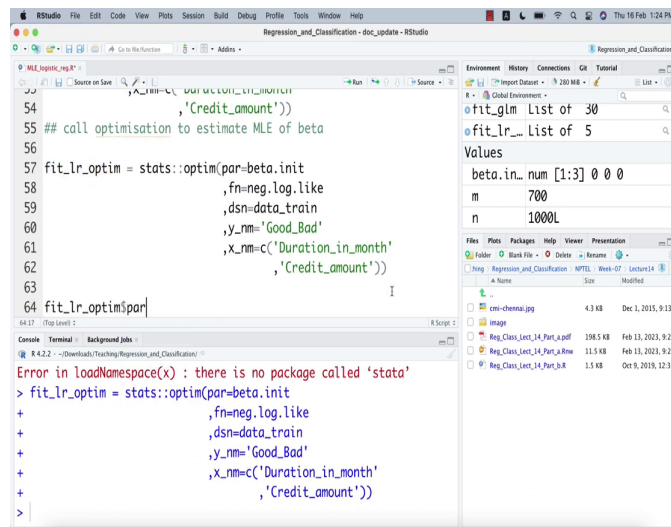
```
> fit_lr_optim = stats::optim(par=beta.init
+                             ,fn=neg.log.like
+                             ,dsn=data_train
+                             ,y_nm='Good_Bad'
+                             ,x_nm=c('Duration_in_month'
+                                     , 'Credit_amount'))
Error in loadNamespace(x) : there is no package called 'stats'
```



And if I just run it yeah it is working perfectly fine and then I am going to write optimization sub routine to get the maximum likelihood estimate write optimization or just call optimization sub routine optimization to estimate MLE of beta ok. So, what I will do, fit lr logistic regression with optim and from the stats package you just call optim ok the parameter first is you have to give the parameter which is initialize beta dot init.

Then you have to give the objective function on what is your objective function? This is your objective function negative log likelihood function. So, you have to give the objective function then you have to give some additional in the objective function you have dsn data dot train. Then in the objective function you have this y name and x name that you have to provide ok.

(Refer Slide Time: 25:55)



```
54 y_lm=c('Duration_in_month',  
55 'Credit_amount'))  
56 ## call optimisation to estimate MLE of beta  
57 fit_lr_optim = stats::optim(par=beta.init  
58 ,fn=neg.log.like  
59 ,dsn=data_train  
60 ,y_nm='Good_Bad'  
61 ,x_nm=c('Duration_in_month'  
62 , 'Credit_amount'))  
63  
64 fit_lr_optim$par
```

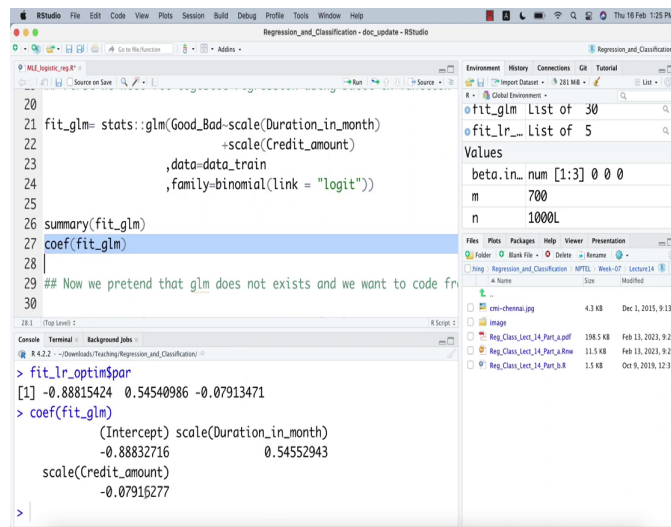
Environment History Connections GR Tutorial
Import Dataset 280 MB
Global Environment
fit_lm List of 30
fit_lr_ List of 5
Values
beta.in_ num [1:3] 0 0 0
m 700
n 1000L
File Packages Help Viewer Presentation
Blank File Delete Remove
Regression_and_Classification NPTEL Week-07 Lecture14
cmi-chennai.jpg 4.3 KB Dec 1, 2015, 9:13 A
image
Reg_Class_Lect_14_Part_a.pdf 108.5 KB Feb 11, 2022, 9:26
Reg_Class_Lect_14_Part_a.Rnw 11.5 KB Feb 11, 2022, 9:26
Reg_Class_Lect_14_Part_b.R 1.5 KB Oct 9, 2019, 12:39

```
R 4.2.2 - ~/Downloads/Teaching/Regression_and_Classification  
Error in loadNamespace(x) : there is no package called 'stata'  
> fit_lr_optim = stats::optim(par=beta.init  
+ ,fn=neg.log.like  
+ ,dsn=data_train  
+ ,y_nm='Good_Bad'  
+ ,x_nm=c('Duration_in_month'  
+ , 'Credit_amount'))  
>
```



So, and if you run this, I have made again a mistake it should be stats not stata and it has run and without any error. So, this is a good news.

(Refer Slide Time: 26:26)



```
20
21 fit_glm= stats::glm(Good_Bad~scale(Duration_in_month)
22                   +scale(Credit_amount)
23                   ,data=data_train
24                   ,family=binomial(link = "logit"))
25
26 summary(fit_glm)
27 coef(fit_glm)
28
29 ## Now we pretend that glm does not exists and we want to code fr
30
```

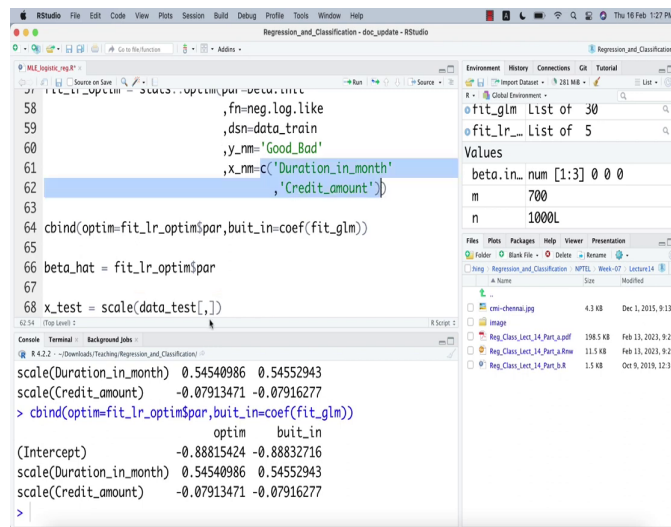
```
> fit_lr_optim$par
[1] -0.88815424  0.54540986 -0.07913471
> coef(fit_glm)
      (Intercept) scale(Duration_in_month)
      -0.88832716           0.54552943
      scale(Credit_amount)
      -0.07915277
```



And let me run from this let me call the parameter estimated parameter value. So, these are my estimated parameter values. So, these are my parameter values from manually estimated whereas, when we use the summary dot fit or maybe coefficients we just call coefficient of fit glm this is looks like these are exactly not exactly.

But at least up to three decimal places they are matching looks like 0791, 0791. So, up to 3 to 4 decimal places it is matching three 3 to 4 decimal places it is matching. There are ways to match it if I increase the tolerance level and all these things then it will match maybe up to a great extent.

(Refer Slide Time: 27:04)



```
58     ,fn=neg.log.like
59     ,dsn=data_train
60     ,y_nm='Good_Bad'
61     ,x_nm=c('Duration_in_month'
62           , 'Credit_amount'))
63
64 cbind(optim=fit_lr_optim$par,buit_in=coef(fit_glm))
65
66 beta_hat = fit_lr_optim$par
67
68 x_test = scale(data_test[,])
```

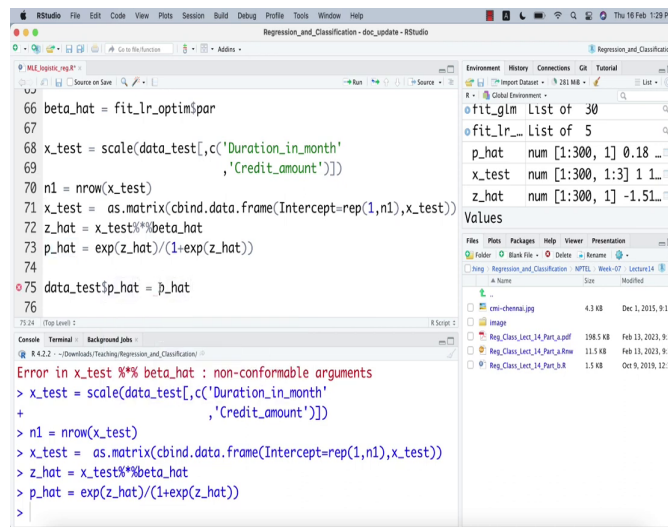
```
scale(Duration_in_month) 0.54540986 0.54552943
scale(Credit_amount)    -0.07913471 -0.07916277
> cbind(optim=fit_lr_optim$par,buit_in=coef(fit_glm))
      optim      buit_in
(Intercept) -0.88815424 -0.88832716
scale(Duration_in_month) 0.54540986 0.54552943
scale(Credit_amount)    -0.07913471 -0.07916277
```



So, what I will do if you. In fact, I can just say cbind and to just compare we can see. So, first one is from optim and second one is from the built in yeah. So, this is the when we use optim this is the value and when we use the built in r function glm this is what happens. So, they are almost matching up to the three decimal places and then they are deviating this one has matched up to four decimal places three to four decimal places they are matching.

If we increase the tolerance level probably, we will match even more. So, now, we have the let us write the beta hat so, since ok. So, beta hat equal to fit lr par and then x test x test we have to take from we have to scale it remember that we have to scale it data test comma these two value.

(Refer Slide Time: 28:58)



```
66 beta_hat = fit_lr_optim$par
67
68 x_test = scale(data_test[,c('Duration_in_month'
69                             , 'Credit_amount')])
70 n1 = nrow(x_test)
71 x_test = as.matrix(cbind.data.frame(Intercept=rep(1,n1),x_test))
72 z_hat = x_test%*%beta_hat
73 p_hat = exp(z_hat)/(1+exp(z_hat))
74
75 data_test$p_hat = p_hat
76
```

```
Environment:
fit_lm_... List of 30
fit_lr_... List of 5
p_hat   num [1:300, 1] 0.18 ...
x_test  num [1:300, 1:3] 1 1...
z_hat   num [1:300, 1] -1.51...
Values
```

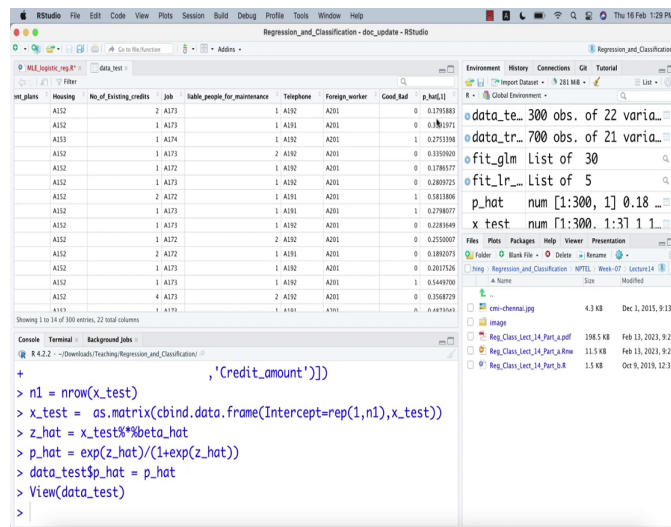
```
Console:
Error in x_test %*% beta_hat : non-conformable arguments
> x_test = scale(data_test[,c('Duration_in_month'
+                               , 'Credit_amount')])
+
> n1 = nrow(x_test)
> x_test = as.matrix(cbind.data.frame(Intercept=rep(1,n1),x_test))
> z_hat = x_test%*%beta_hat
> p_hat = exp(z_hat)/(1+exp(z_hat))
>
```



So, we have to take the test ok we got the test values and then what I am going to do is we are going to calculate this first the latent variable z hat which is x test percentage star percentage β hat. Sorry I have to call this guy and then yeah it is not confirmable. So, of course, because I have to do few more work in row of x test and then I have to call x test equal to you see I have done this line right.

So, I have to do the same kind of lining arrangement here also because that is not x test directly is not my you know my call. So, let me just run this and now probably it will be fine, yes now it is fine now it is fine. And now once I have the set test let me just have p hat p hat equals to e to the power z hat divided by 1 plus e to the power z hat ok. So, this is my p hat. Now, now in the data test data test I am going to p hat equals to p hat this is going to be a new column in the data test. See I have created a new column here alright.

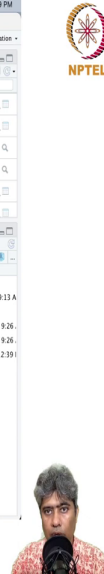
(Refer Slide Time: 31:17)



The screenshot shows the RStudio interface. The top pane displays a data table with columns: mt_class, Housing, No_of_Existing_credits, job, table_people_for_maintenance, Telephone, Foreign_worker, Good_Ead, and p_hat. The console shows the following R code and output:

```
R 4.2.2 - ~/Downloads/Teaching/Regression_and_Classification/
+                                     , 'Credit_amount'])
> n1 = nrow(x_test)
> x_test = as.matrix(cbind.data.frame(Intercept=rep(1,n1),x_test))
> z_hat = x_test%*%beta_hat
> p_hat = exp(z_hat)/(1+exp(z_hat))
> data_testp_hat = p_hat
> View(data_test)
>
```

The environment pane on the right shows several objects: data_te_ (300 obs. of 22 variables), data_tr_ (700 obs. of 21 variables), fit_glm (List of 30), fit_lr_ (List of 5), p_hat (num [1:300, 1] 0.18...), and x_test (num [1:300, 1:31] 1 1...).



Next what I am going to do I am not sure why it shows a p hat column 1 ok, we will see what can be done alright. Next is we have to predict.

(Refer Slide Time: 32:08)

```
67
68 x_test = scale(data_test[,c('Duration_in_month'
69                            , 'Credit_amount')])
70 n1 = nrow(x_test)
71 x_test = as.matrix(cbind.data.frame(Intercept=rep(1,n1),x_test))
72 z_hat = x_test%*%beta_hat
73 p_hat = exp(z_hat)/(1+exp(z_hat))
74
75 data_test$p_hat = p_hat
76 data_test$Good_Bad_pred = 0
77
```

```
> p_hat = exp(z_hat)/(1+exp(z_hat))
> data_test$p_hat = p_hat
> View(data_test)
> View(data_test)
> data_test$p_hat = p_hat
> View(data_test)
> data_test$Good_Bad_pred = 0
>
```



(Refer Slide Time: 32:24)

The screenshot displays the RStudio interface with the following components:

- Environment Pane:** Shows objects including `data_test` (300 obs. of 23 variables), `data_train` (700 obs. of 21 variables), `fit_glm` (List of 30), `fit_lr` (List of 5), `p_hat` (num [1:300, 1] 0.18...), and `x_test` (num [1:300, 1:3] 1 1...).
- Table:** A data frame with columns: `job`, `no_of_existing_credits`, `liable_people_for_maintenance`, `Telephone`, `Foreign_worker`, `Good_Bad`, `p_hat_1`, and `Good_Bad_pred`.
- Console:** Shows the execution of the following R commands:

```
> data_test$p_hat = p_hat
> View(data_test)
> View(data_test)
> data_test$p_hat = p_hat
> View(data_test)
> data_test$Good_Bad_pred = 0
> View(data_test)
>
```



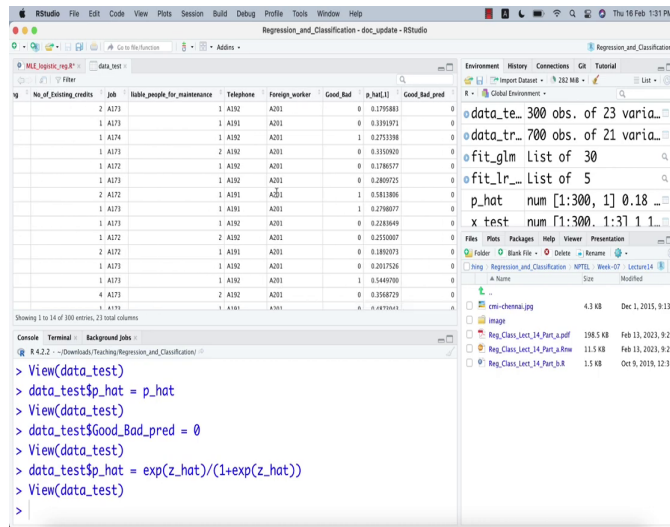
(Refer Slide Time: 32:32)

```
67
68 x_test = scale(data_test[,c('Duration_in_month'
69                            , 'Credit_amount')])
70 n1 = nrow(x_test)
71 x_test = as.matrix(cbind.data.frame(Intercept=rep(1,n1),x_test))
72 z_hat = x_test%*%beta_hat
73 p_hat = exp(z_hat)/(1+exp(z_hat))
74
75 data_test$p_hat = exp(z_hat)/(1+exp(z_hat))
76 data_test$Good_Bad_pred = 0
77
```

```
> View(data_test)
> View(data_test)
> data_test$p_hat = p_hat
> View(data_test)
> data_test$Good_Bad_pred = 0
> View(data_test)
> data_test$p_hat = exp(z_hat)/(1+exp(z_hat))
>
```



(Refer Slide Time: 32:38)

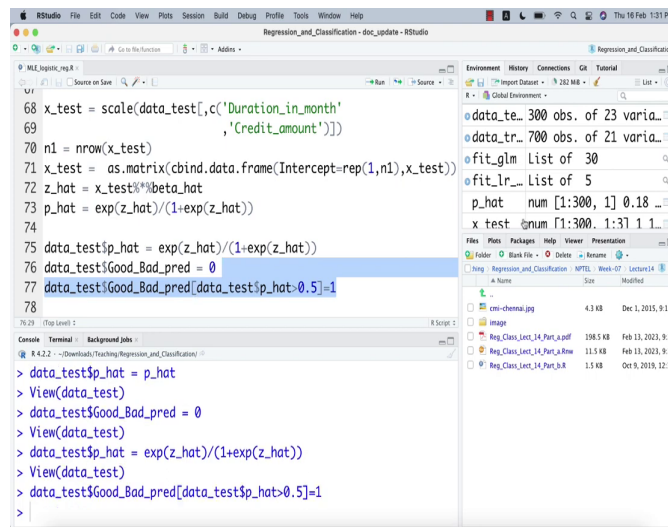


The screenshot displays the RStudio interface. The main editor shows a data frame with columns: job, liable_people_for_maintenance, Telephone, Foreign_worker, Good_Bad, p_hat_1, and Good_Bad_pred. The environment pane on the right lists objects: data_test (300 obs. of 23 variables), data_train (700 obs. of 21 variables), fit_glm (List of 30), fit_lr_ (List of 5), p_hat (num [1:300, 1] 0.18...), and x_test (num [1:300, 1:3] 1 1...). The console shows the following R commands and their outputs:

```
> View(data_test)
> data_test$p_hat = p_hat
> View(data_test)
> data_test$Good_Bad_pred = 0
> View(data_test)
> data_test$p_hat = exp(z_hat)/(1+exp(z_hat))
> View(data_test)
>
```



(Refer Slide Time: 32:45)



```
68 x_test = scale(data_test[,c('Duration_in_month',
69 'Credit_amount')])
70 n1 = nrow(x_test)
71 x_test = as.matrix(cbind.data.frame(Intercept=rep(1,n1),x_test))
72 z_hat = x_test%*%beta_hat
73 p_hat = exp(z_hat)/(1+exp(z_hat))
74
75 data_test$p_hat = exp(z_hat)/(1+exp(z_hat))
76 data_test$Good_Bad_pred = 0
77 data_test$Good_Bad_pred[data_test$p_hat>0.5]=1
78
```

```
> data_test$p_hat = p_hat
> View(data_test)
> data_test$Good_Bad_pred = 0
> View(data_test)
> data_test$p_hat = exp(z_hat)/(1+exp(z_hat))
> View(data_test)
> data_test$Good_Bad_pred[data_test$p_hat>0.5]=1
>
```



So, data test dollar Good Bad, but we have to do first prediction the prediction equal to first I am saying 0 and then first I am taking all the values of sort of a 0. Actually, what I can do? I can just take this yeah, I will see what (Refer Time: 32:44) can do and then I can just say that if p hat is greater than 1 greater than half. Then most likely its a going to be a bad customer.

(Refer Slide Time: 33:09)

The screenshot displays the RStudio environment. The main window shows a data table with columns: job, no_of_existing_credits, liable_people_for_maintenance, Telephone, Foreign_worker, Good_Bad, p_hat_01, and Good_Bad_pred. The console shows the following R commands and their output:

```
> View(data_test)
> data_test$Good_Bad_pred = 0
> View(data_test)
> data_test$p_hat = exp(z_hat)/(1+exp(z_hat))
> View(data_test)
> data_test$Good_Bad_pred[data_test$p_hat>0.5]=1
> View(data_test)
>
```

The Environment pane on the right shows the following objects:

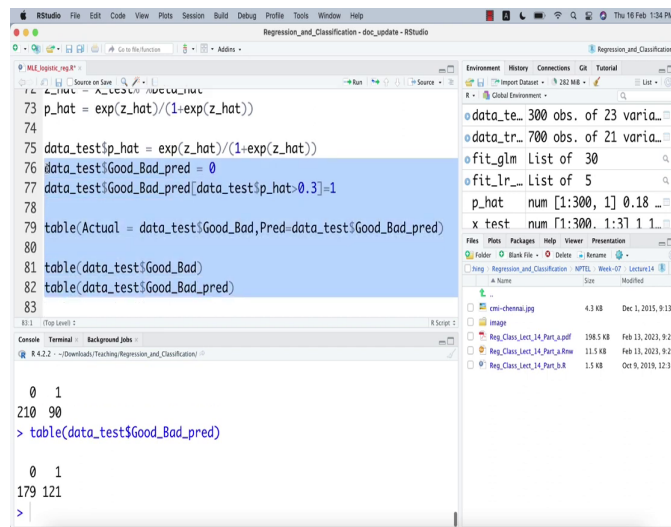
- data_te... 300 obs. of 23 varia...
- data_tr... 700 obs. of 21 varia...
- fit_glm List of 30
- fit_lr... List of 5
- p_hat num [1:300, 1] 0.18 ...
- x_test num [1:300, 1:3] 1 1 ...

The Files pane shows a folder named 'Regression_and_Classification' containing several files:

- cmi-chennai.jpg (4.3 KB, Dec 1, 2015, 9:13 A)
- image
- Reg_Class_Lect_14_Part_a.pdf (108.5 KB, Feb 11, 2022, 9:26)
- Reg_Class_Lect_14_Part_a.Rnw (11.5 KB, Feb 11, 2022, 9:26)
- Reg_Class_Lect_14_Part_b.R (1.4 KB, Oct 9, 2019, 12:39)



(Refer Slide Time: 33:21)



```
73 p_hat = exp(z_hat)/(1+exp(z_hat))
74
75 data_test$p_hat = exp(z_hat)/(1+exp(z_hat))
76 data_test$Good_Bad_pred = 0
77 data_test$Good_Bad_pred[data_test$p_hat>0.3]=1
78
79 table(Actual = data_test$Good_Bad,Pred=data_test$Good_Bad_pred)
80
81 table(data_test$Good_Bad)
82 table(data_test$Good_Bad_pred)
83
```

```
0 1
210 90
> table(data_test$Good_Bad_pred)

0 1
179 121
>
```

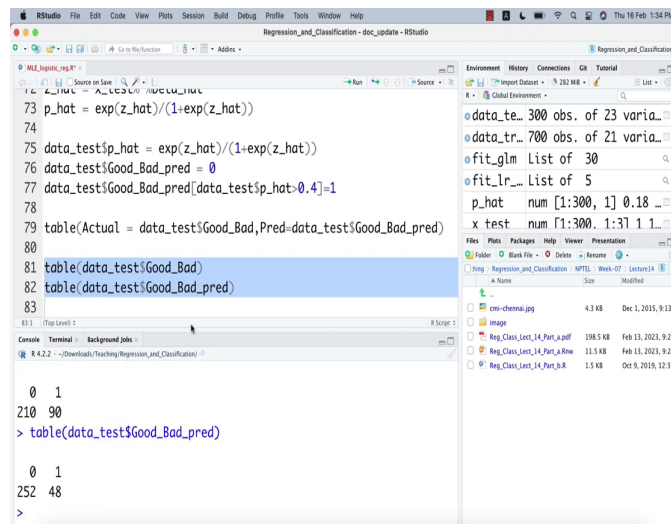


So, there are some prediction where it has said it is going to be a bad customer ok. Now I am going to do a table I call accuracy actual equals to data test dollar Good Bad and prediction predict equal to data test dollar Good Bad predicted values and this is the case. So, there are out of about 300, 212 cases where 202 plus 210 there are too many like bad customer is being missed and said that ok these are Good.

So, it is looks like the model is predicting most of the cases are as good 2000 out of 300 test cases. So, in the test cases test data set data test if I just do a simple table now data test comma good bad they are about 210 cases are good and 90 cases are bad. Now on the other hand if I just do a table of data test Good Bad pred. So, 282 cases are saying that they are Good and 18 cases they are saying Bad. So, this model is actually failing to you know failing to say who is Good who is Bad.

So, one thing people do try to understand that if I in case of 0.5 if I increase it to say 0.3 and will it be helpful. So, if you do that say probably yes, instead of 0.5 if you just reduce it now you have just too many probably saying or maybe 0.4. So, you have these many cases.

(Refer Slide Time: 36:07)



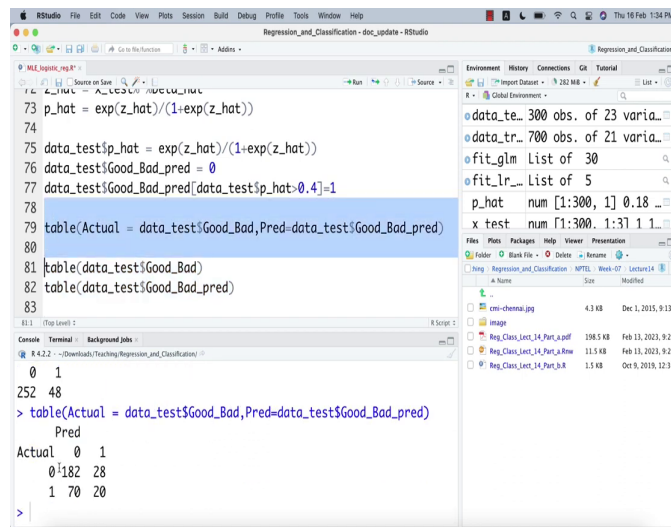
```
73 p_hat = exp(z_hat)/(1+exp(z_hat))
74
75 data_test$p_hat = exp(z_hat)/(1+exp(z_hat))
76 data_test$Good_Bad_pred = 0
77 data_test$Good_Bad_pred[data_test$p_hat>0.4]=1
78
79 table(Actual = data_test$Good_Bad, Pred=data_test$Good_Bad_pred)
80
81 table(data_test$Good_Bad)
82 table(data_test$Good_Bad_pred)
83
```

```
0 1
210 90
> table(data_test$Good_Bad_pred)

0 1
252 48
>
```



(Refer Slide Time: 36:30)



```
73 p_hat = exp(z_hat)/(1+exp(z_hat))
74
75 data_test$p_hat = exp(z_hat)/(1+exp(z_hat))
76 data_test$Good_Bad_pred = 0
77 data_test$Good_Bad_pred[data_test$p_hat>0.4]=1
78
79 table(Actual = data_test$Good_Bad,Pred=data_test$Good_Bad_pred)
80
81 table(data_test$Good_Bad)
82 table(data_test$Good_Bad_pred)
83
```

```
0 1
252 48
> table(Actual = data_test$Good_Bad,Pred=data_test$Good_Bad_pred)
      Pred
Actual 0  1
      0:182 28
      1  70 20
```



So, if you have that now you have kind of 90 cases in actual cases these are the actual. So, 90 cases good and now it is better slightly better you can try out with this cut off marks. But the problem is another level now problem is there are two type of error one is truly there is a Good cases, which is being predicted as bad and there are Bad cases actually bad cases which is predicted as Good.

So, this is like actual bad cases which is predicted as good, which is actually bad thing you do not want this to happen if you as a bank manager you are giving loan to somebody and he or she is not returning in the money that is a really bad thing that is a real loss. This error on the other hand for a bank where actually the person is good, but bank thought ok if they will be bad customer and you are not giving loan.

So, in that case it is going to be a real problem. So, there are we are seeing in binary classification there are two kinds of problems. So, and one problem is typically called false positive another called false negative in statistical language it is called type 1 error and type 2 error. So, we are going to talk about that which one is good and which one is bad in the next lecture we are going to talk about that so, for now.

Thank you very much.