

assignment9

2023-04-14

#HW09 #Use the R programs covered in the MultipleTesting module as a guide for HW09.

#Perform a multiple regression on all 100 predictors from the meatspec dataset in the #faraway package and extract the p-values for the coefficients of the 100 variables.

```
library(faraway)
data(meatspec, package="faraway") #the dataset meatspec is found in the faraway package

lmod=lm(fat ~ . , data=meatspec)
summary(lmod)
```

```
##
## Call:
## lm(formula = fat ~ ., data = meatspec)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.9833 -0.4982  0.0135  0.4864  3.1727
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      7.302      1.876   3.892 0.000168 ***
## V1             10898.047    3003.614   3.628 0.000428 ***
## V2             -12174.864    5520.233  -2.205 0.029426 *
## V3              -5953.285    8868.517  -0.671 0.503398
## V4              23229.862   15426.530   1.506 0.134875
## V5             -28386.219   19758.501  -1.437 0.153554
## V6              12748.270   17381.421   0.733 0.464794
## V7             -11422.335   11454.169  -0.997 0.320769
## V8               7102.332    7123.935   0.997 0.320892
## V9               783.655     5228.808   0.150 0.881130
## V10             3512.239     6787.803   0.517 0.605856
## V11            -10547.574   10580.407  -0.997 0.320926
## V12             34638.288   18344.772   1.888 0.061543 .
## V13            -38705.447   23098.395  -1.676 0.096542 .
## V14             28895.947   19952.355   1.448 0.150293
## V15            -13726.347   13312.307  -1.031 0.304676
## V16             -7062.769    8172.878  -0.864 0.389308
## V17              2571.597    6279.661   0.410 0.682932
## V18              5263.427    6183.397   0.851 0.396432
## V19              8860.827    8925.154   0.993 0.322914
## V20            -12149.937   15184.189  -0.800 0.425276
## V21            -19284.872   20536.132  -0.939 0.349680
## V22             36626.953   22847.592   1.603 0.111680
```

## V23	-11165.390	19302.712	-0.578	0.564111	
## V24	-15008.939	13616.072	-1.102	0.272655	
## V25	16698.992	8582.462	1.946	0.054151	.
## V26	-4891.852	5901.456	-0.829	0.408880	
## V27	-6334.752	6072.685	-1.043	0.299084	
## V28	24043.786	8144.906	2.952	0.003834	**
## V29	-39940.900	12335.575	-3.238	0.001578	**
## V30	33309.092	17674.622	1.885	0.062034	.
## V31	-23174.509	20974.708	-1.105	0.271539	
## V32	18764.305	18959.821	0.990	0.324423	
## V33	-3747.892	13458.994	-0.278	0.781158	
## V34	-6671.747	9353.448	-0.713	0.477122	
## V35	-5318.549	7534.861	-0.706	0.481716	
## V36	10488.898	5773.159	1.817	0.071869	.
## V37	-8410.539	5892.265	-1.427	0.156202	
## V38	-408.228	7970.269	-0.051	0.959241	
## V39	19815.971	11338.219	1.748	0.083206	.
## V40	-23690.179	15971.026	-1.483	0.140748	
## V41	29398.659	19340.032	1.520	0.131256	
## V42	-32055.252	20639.448	-1.553	0.123170	
## V43	11826.000	17491.895	0.676	0.500356	
## V44	-9994.257	11435.392	-0.874	0.383969	
## V45	23017.798	8927.175	2.578	0.011200	*
## V46	-9041.633	6218.630	-1.454	0.148705	
## V47	-4846.799	3520.124	-1.377	0.171246	
## V48	1536.042	4401.789	0.349	0.727764	
## V49	2188.418	7363.225	0.297	0.766848	
## V50	-13170.870	9829.843	-1.340	0.182947	
## V51	26420.737	13371.372	1.976	0.050580	.
## V52	-23565.834	16339.395	-1.442	0.151968	
## V53	-2005.210	16742.496	-0.120	0.904878	
## V54	30327.413	14023.378	2.163	0.032658	*
## V55	-31802.344	10650.780	-2.986	0.003461	**
## V56	12428.271	6395.916	1.943	0.054463	.
## V57	-102.107	4676.993	-0.022	0.982620	
## V58	210.251	4388.133	0.048	0.961869	
## V59	-7679.011	4511.526	-1.702	0.091465	.
## V60	11590.949	3967.244	2.922	0.004199	**
## V61	-6559.639	3756.703	-1.746	0.083485	.
## V62	2533.819	3939.248	0.643	0.521370	
## V63	11950.924	5296.267	2.256	0.025947	*
## V64	-18515.851	7070.171	-2.619	0.010021	*
## V65	4051.697	8539.248	0.474	0.636066	
## V66	222.861	9691.472	0.023	0.981694	
## V67	10439.030	10111.231	1.032	0.304061	
## V68	-22570.742	9493.417	-2.378	0.019094	*
## V69	17285.149	8168.742	2.116	0.036520	*
## V70	-45.036	7357.838	-0.006	0.995127	
## V71	-8134.714	6796.093	-1.197	0.233802	
## V72	-1768.780	6344.295	-0.279	0.780905	
## V73	15744.948	5531.706	2.846	0.005246	**
## V74	-11219.545	5666.910	-1.980	0.050132	.
## V75	5289.427	5067.718	1.044	0.298810	
## V76	-2454.612	4760.274	-0.516	0.607101	

```
## V77          740.608    4922.688    0.150 0.880677
## V78         -5730.806    5518.607   -1.038 0.301257
## V79          12166.493    6026.835    2.019 0.045863 *
## V80        -22688.979    7023.823   -3.230 0.001616 **
## V81          14991.763    8595.338    1.744 0.083824 .
## V82           3331.367    9984.910    0.334 0.739264
## V83         -6651.082   11358.746   -0.586 0.559337
## V84         -6752.949   12405.922   -0.544 0.587276
## V85          16271.066   12434.546    1.309 0.193323
## V86           5512.031   13689.180    0.403 0.687955
## V87        -21092.220   15770.171   -1.337 0.183730
## V88           9657.690   15143.593    0.638 0.524921
## V89           273.586   13103.448    0.021 0.983379
## V90         -5489.915   13927.199   -0.394 0.694180
## V91           2891.941   15479.740    0.187 0.852133
## V92          10160.850   14407.777    0.705 0.482103
## V93         -3183.235   11882.686   -0.268 0.789269
## V94         -7330.650   10959.287   -0.669 0.504913
## V95           5551.521    9450.485    0.587 0.558075
## V96         -3320.415    8349.562   -0.398 0.691613
## V97         -2512.787    7974.922   -0.315 0.753272
## V98         -5979.563    7355.289   -0.813 0.417935
## V99           8283.253    7911.765    1.047 0.297336
## V100         -101.926    3591.166   -0.028 0.977407
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.22 on 114 degrees of freedom
## Multiple R-squared:  0.9951, Adjusted R-squared:  0.9908
## F-statistic: 232 on 100 and 114 DF, p-value: < 2.2e-16
```

```
# Extract p-values for coefficients and excluding the intercept
p_values <- summary(lmod)$coefficients[-1, 4]
```

```
p_values
```

```
##          V1          V2          V3          V4          V5          V6
## 0.0004283297 0.0294256650 0.5033981889 0.1348745281 0.1535538037 0.4647943207
##          V7          V8          V9          V10         V11         V12
## 0.3207693528 0.3208916975 0.8811299832 0.6058559085 0.3209258938 0.0615433570
##          V13         V14         V15         V16         V17         V18
## 0.0965422164 0.1502926730 0.3046755629 0.3893083081 0.6829323019 0.3964319860
##          V19         V20         V21         V22         V23         V24
## 0.3229136323 0.4252764439 0.3496800608 0.1116799533 0.5641105785 0.2726547758
##          V25         V26         V27         V28         V29         V30
## 0.0541511081 0.4088802236 0.2990835887 0.0038340431 0.0015775341 0.0620343696
##          V31         V32         V33         V34         V35         V36
## 0.2715386246 0.3244226674 0.7811579873 0.4771222175 0.4817158475 0.0718690219
##          V37         V38         V39         V40         V41         V42
## 0.1562020058 0.9592407305 0.0832059828 0.1407482329 0.1312561915 0.1231696311
##          V43         V44         V45         V46         V47         V48
## 0.5003562519 0.3839687639 0.0111996624 0.1487052405 0.1712463541 0.7277641070
##          V49         V50         V51         V52         V53         V54
```

```
## 0.7668477428 0.1829470928 0.0505798445 0.1519677289 0.9048779908 0.0326584621
##          V55          V56          V57          V58          V59          V60
## 0.0034611034 0.0544625942 0.9826203770 0.9618689776 0.0914645627 0.0041986807
##          V61          V62          V63          V64          V65          V66
## 0.0834846286 0.5213704458 0.0259469095 0.0100214548 0.6360656884 0.9816939643
##          V67          V68          V69          V70          V71          V72
## 0.3040610504 0.0190942824 0.0365201162 0.9951269670 0.2338017348 0.7809045001
##          V73          V74          V75          V76          V77          V78
## 0.0052463432 0.0501317942 0.2988096151 0.6071008125 0.8806771054 0.3012574109
##          V79          V80          V81          V82          V83          V84
## 0.0458632490 0.0016162837 0.0838241617 0.7392639838 0.5593371748 0.5872756605
##          V85          V86          V87          V88          V89          V90
## 0.1933231459 0.6879553379 0.1837297663 0.5249211429 0.9833787242 0.6941796989
##          V91          V92          V93          V94          V95          V96
## 0.8521330321 0.4821032694 0.7892692561 0.5049125530 0.5580750742 0.6916131495
##          V97          V98          V99          V100
## 0.7532723377 0.4179350446 0.2973355717 0.9774068543
```

```
#Use the p-values vector to list the variables with coefficients that are significantly
#different from 0 with
#a) using no FWER adjustment (alpha=.05 for each test of hypothesis).
alpha <- 0.05
print(p_values[p_values < alpha])
```

```
##          V1          V2          V28          V29          V45          V54
## 0.0004283297 0.0294256650 0.0038340431 0.0015775341 0.0111996624 0.0326584621
##          V55          V60          V63          V64          V68          V69
## 0.0034611034 0.0041986807 0.0259469095 0.0100214548 0.0190942824 0.0365201162
##          V73          V79          V80
## 0.0052463432 0.0458632490 0.0016162837
```

```
names(p_values[p_values < alpha])
```

```
## [1] "V1" "V2" "V28" "V29" "V45" "V54" "V55" "V60" "V63" "V64" "V68" "V69"
## [13] "V73" "V79" "V80"
```

```
# We reject these 15 features based on the p values
```

```
#b) using the Bonferroni procedure to control the FWER alpha at 0.05.
#Get adjusted p value
p_values_bonfadjust <- p.adjust(p_values, method = "bonferroni")

# Identify variables with significant coefficients
adj_bonferroni <- p_values_bonfadjust[p_values_bonfadjust < alpha]

adj_bonferroni
```

```
##          V1
## 0.04283297
```

```
names(adj_bonferroni)
```

```
## [1] "V1"
```

```
# We reject only V1 based on the Bonferroni procedure
```

```
#c) using the Holm procedure to control the FWER alpha at 0.05.
```

```
# Get the adjusted p-val
```

```
p_values_holmadjust <- p.adjust(p_values, method = "holm")
```

```
# Identify variables with significant coefficients
```

```
adj_holm <- p_values_holmadjust[p_values_holmadjust < alpha]
```

```
adj_holm
```

```
##          V1
```

```
## 0.04283297
```

```
names(adj_holm)
```

```
## [1] "V1"
```

```
#We again reject only V1 based on Holm procedure as well
```