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LAMPIRAN

Lampiran 1: Data Tuberkulosis Tahun 2021

| | y | x_1 | x_2 | x_3 | x_4 | ... | x_{21} |
|--------------------------|------|---------|--------|-------|-------|-----|----------|
| Kepulauan Selayar | 220 | 1180,52 | 146,18 | 34 | 8 | ... | 2986 |
| Bulukumba | 538 | 1170,71 | 10,84 | 104 | 3 | ... | 1183 |
| Bantaeng | 323 | 395,96 | 45,8 | 46 | 34 | ... | 36943 |
| Jeneponto | 562 | 797,5 | 86,2 | 87 | 50 | ... | 609 |
| Takalar | 635 | 554,78 | 25,16 | 73 | 0 | ... | 3246 |
| Gowa | 1334 | 1181,38 | 213,78 | 181 | 57 | ... | 115 |
| Sinjai | 352 | 864,72 | 24,66 | 54 | 21 | ... | 11412 |
| Maros | 533 | 1435,04 | 34,93 | 80 | 20 | ... | 6329 |
| Pangkajene Dan Kepulauan | 680 | 889,01 | 0 | 80 | 13 | ... | 1203 |
| Barru | 257 | 1202,16 | 0 | 43 | 7 | ... | 1970 |
| Bone | 1012 | 4571,03 | 67,61 | 174 | 15 | ... | 5434 |
| Soppeng | 208 | 1370,41 | 37,99 | 59 | 24 | ... | 1127 |
| Wajo | 697 | 2628,44 | 59,72 | 99 | 34 | ... | 705 |
| Sidenreng Rappang | 432 | 1790,31 | 36,27 | 73 | 37 | ... | 826 |
| Pinrang | 641 | 1884,36 | 53,4 | 90 | 9 | ... | 624 |
| Enrekang | 187 | 1821,53 | 15,62 | 47 | 10 | ... | 3431 |
| Luwu | 526 | 3048,49 | 10,43 | 80 | 2 | ... | 3112 |
| Tana Toraja | 170 | 2088,87 | 0 | 56 | 0 | ... | 1604 |
| Kota Makassar | 3908 | 177,96 | 77,92 | 361 | 380 | ... | 116838 |
| ... | ... | ... | ... | ... | ... | ... | 8812 |
| Kota Palopo | 388 | 250,64 | 95,03 | 42 | 32 | ... | 10370 |

Lampiran 2 : Sintaks dari Program R beserta output

```

#Input Data
qrdata<-read_excel("D:/DATA Thesis/AA.xlsx")
attach(qrdata)

#Deskripsi Data
summary(qrdata)
plot(qrdata)

#Mengidentifikasi Pencilan/Outlier
boxplot(qrdata)

#Uji multikolinearitas
datanew <- lm(formula =
y~x1+x2+x3+x4+x5+x6+x7+x8+x9+x10+x11+x12+x13+x14+x15+x16+x17+x18+x1
9+x20+x21, qrdata)
vif(datanew)

# Menentukan grup untuk penalti group lasso
Y=cbind(y)
X=cbind(x1,x2,x3,x4,x5,x6,x7,x8,x9,x10,x11,x12,x13,x14,x15,x16,x17,x18,x19,x20,
x21)
group=c(1,1,1,1,2,2,2,2,3,3,3,3,4,4,4,5,5,5,6,6,6)
group

# Jalankan model regresi kuantil dengan penalti group lasso dan cross-validation
fit_0.25 <- qr.glasso(X, Y, group, tau = 0.25, nfolds = 10)
fit_0.5 <- qr.glasso(X, Y, group, tau = 0.5, nfolds = 10)
fit_0.75 <- qr.glasso(X, Y, group, tau = 0.75, nfolds = 10)

# Mendapatkan koefisien untuk setiap kuantil
coef_0.25 <- coef(fit_0.25$final_fit)
coef_0.5 <- coef(fit_0.5$final_fit)
coef_0.75 <- coef(fit_0.75$final_fit)

# Membuat tabel koefisien
coef_table <- data.frame(
  Variable = names(coef_0.25),
  Q_0.25 = round(coef_0.25, 5),
  Q_0.5 = round(coef_0.5, 5),
  Q_0.75 = round(coef_0.75, 5)
)

```

```
#Mendapatkan Hasil Estimasi Regresi kuantil dengan Penalti Group LASSO
```

```
print(coef_table)
```

```
Output :
```

```
Variable   Q_0.25   Q_0.5   Q_0.75
(Intercept) (Intercept) -272.45192 -288.19502 -286.48254
x1          x1 -0.00943 -0.00950 -0.00949
x2          x2  0.62013  0.69314  0.68373
x3          x3  5.37901  4.57778  4.69162
x4          x4 -0.17273 -0.09341 -0.10787
x5          x5  0.00029  0.00033  0.00032
x6          x6  0.03471  0.03030  0.03097
x7          x7 11.18028 15.71492 15.34787
x8          x8 -1.78582 -1.56991 -1.58565
x9          x9 -0.45372  0.32498  0.24775
x10         x10  0.23090  0.19601  0.19810
x11         x11 -0.28305 -0.24799 -0.25070
x12         x12 -0.29292 -0.26168 -0.26394
x13         x13 -0.00026  0.00004  0.00003
x14         x14  0.00001  0.00020  0.00018
x15         x15  0.01188  0.01011  0.01029
x16         x16 -0.00300 -0.00268 -0.00272
x17         x17  0.00711  0.00680  0.00685
x18         x18  0.55221  0.55190  0.55741
x19         x19  0.16691  0.14977  0.14786
x20         x20  0.00111  0.00141  0.00136
x21         x21  0.00429  0.00424  0.00422
```

```
# Print lambda.min values
```

```
cat("Lambda min (0.25):", fit_0.25$cv_fit$lambda.min, "\n")
```

```
cat("Lambda min (0.5):", fit_0.5$cv_fit$lambda.min, "\n")
```

```
cat("Lambda min (0.75):", fit_0.75$cv_fit$lambda.min, "\n")
```

```
Output :
```

```
Lambda min (0.25): 1.188423
```

```
Lambda min (0.5): 3.013086
```

```
Lambda min (0.75): 2.745411
```

```
# Summary dari cross-validation  
summary(fit_0.25$cv_fit)
```

Nonzero coefficients: 21

Nonzero groups: 6

Maximum R-squared: 0.98

Maximum signal-to-noise ratio: 53.17

Scale estimate (sigma) at lambda.min: 100.061

```
summary(fit_0.5$cv_fit)
```

Nonzero coefficients: 21

Nonzero groups: 6

Maximum R-squared: 0.99

Maximum signal-to-noise ratio: 79.31

Scale estimate (sigma) at lambda.min: 82.179

```
summary(fit_0.75$cv_fit)
```

Nonzero coefficients: 21

Nonzero groups: 6

Maximum R-squared: 0.97

Maximum signal-to-noise ratio: 38.71

Scale estimate (sigma) at lambda.min: 116.864