Statistical Modelling Techniques ~ Elif Kartal

Create a single script with the appropriate and different built-in datasets for Regression, ANOVA, ANCOVA, Generalized Linear Model with any specific link function, Linear Probability Model, Logit OR Probit Model, Truncated Regression, Censored Regression, Poisson Model, Negative-Binomial Model, Zero-inflated Model, Quantile Regression. Briefly write dataset properties (Which properties of this dataset cause you to use X analysis method?) and comment on all results (with#), separately.

Install and load necessary libraries:

```
install.packages("datasets")
Installing package into '/cloud/lib/x86_64-pc-linux-gnu-library/4.4'
(as 'lib' is unspecified)
Warning: package 'datasets' is a base package, and should not be updated
install.packages("car")
Installing package into '/cloud/lib/x86_64-pc-linux-gnu-library/4.4'
(as 'lib' is unspecified)
install.packages("lmtest")
Installing package into '/cloud/lib/x86_64-pc-linux-gnu-library/4.4'
(as 'lib' is unspecified)
install.packages("sandwich")
Installing package into '/cloud/lib/x86_64-pc-linux-gnu-library/4.4'
(as 'lib' is unspecified)
install.packages("survival")
Installing package into '/cloud/lib/x86_64-pc-linux-gnu-library/4.4'
(as 'lib' is unspecified)
```

```
install.packages("MASS")
Installing package into '/cloud/lib/x86_64-pc-linux-gnu-library/4.4'
```

```
install.packages("pscl")
```

(as 'lib' is unspecified)

Installing package into '/cloud/lib/x86_64-pc-linux-gnu-library/4.4'
(as 'lib' is unspecified)

```
install.packages("quantreg")
```

Installing package into '/cloud/lib/x86_64-pc-linux-gnu-library/4.4' (as 'lib' is unspecified)

```
install.packages("VGAM")
```

Installing package into '/cloud/lib/x86_64-pc-linux-gnu-library/4.4'
(as 'lib' is unspecified)

```
install.packages("AER")
```

Installing package into '/cloud/lib/x86_64-pc-linux-gnu-library/4.4'
(as 'lib' is unspecified)

```
install.packages("truncreg")
```

Installing package into '/cloud/lib/x86_64-pc-linux-gnu-library/4.4'
(as 'lib' is unspecified)

```
install.packages("censReg")
```

Installing package into '/cloud/lib/x86_64-pc-linux-gnu-library/4.4'
(as 'lib' is unspecified)

library(AER)

```
Loading required package: car
Loading required package: carData
Loading required package: lmtest
Loading required package: zoo
Attaching package: 'zoo'
The following objects are masked from 'package:base':
    as.Date, as.Date.numeric
Loading required package: sandwich
Loading required package: survival
library(datasets)
library(car)
library(lmtest)
library(sandwich)
library(survival)
library(MASS)
library(pscl)
```

Classes and Methods for R originally developed in the Political Science Computational Laboratory
Department of Political Science
Stanford University (2002-2015),
by and under the direction of Simon Jackman.
hurdle and zeroinfl functions by Achim Zeileis.

library(quantreg) Loading required package: SparseM Attaching package: 'SparseM' The following object is masked from 'package:base': backsolve Attaching package: 'quantreg' The following object is masked from 'package:survival': untangle.specials library(VGAM) Loading required package: stats4 Loading required package: splines Attaching package: 'VGAM' The following object is masked from 'package:AER': tobit The following object is masked from 'package:lmtest': lrtest The following object is masked from 'package:car':

logit

library(truncreg)

```
Loading required package: maxLik
```

Loading required package: miscTools

```
Please cite the 'maxLik' package as:
```

Henningsen, Arne and Toomet, Ott (2011). maxLik: A package for maximum likelihood estimation

If you have questions, suggestions, or comments regarding the 'maxLik' package, please use a https://r-forge.r-project.org/projects/maxlik/

library(censReg)

```
Please cite the 'censReg' package as:
```

Henningsen, Arne (2017). censReg: Censored Regression (Tobit) Models. R package version 0.5.

If you have questions, suggestions, or comments regarding the 'censReg' package, please use https://r-forge.r-project.org/projects/sampleselection/

Linear Regression:

```
data(mtcars)
linear_model <- lm(mpg ~ ., data = mtcars)
summary(linear_model)</pre>
```

Call:

```
lm(formula = mpg ~ ., data = mtcars)
```

Residuals:

```
Min 1Q Median 3Q Max -3.4506 -1.6044 -0.1196 1.2193 4.6271
```

Coefficients:

```
Estimate Std. Error t value Pr(>|t|)
(Intercept) 12.30337 18.71788 0.657 0.5181
```

```
-0.11144
                   1.04502 -0.107
                                     0.9161
cyl
          0.01334 0.01786 0.747
disp
                                     0.4635
          -0.02148 0.02177 -0.987
                                     0.3350
hp
          0.78711 1.63537 0.481
                                     0.6353
drat
          -3.71530 1.89441 -1.961
wt
                                     0.0633 .
          0.82104 0.73084 1.123
                                     0.2739
qsec
           0.31776 2.10451 0.151
                                     0.8814
٧S
           2.52023 2.05665 1.225
am
                                     0.2340
          0.65541
                   1.49326 0.439
                                     0.6652
gear
carb
          -0.19942 0.82875 -0.241
                                    0.8122
Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
Residual standard error: 2.65 on 21 degrees of freedom
Multiple R-squared: 0.869, Adjusted R-squared: 0.8066
F-statistic: 13.93 on 10 and 21 DF, p-value: 3.793e-07
```

#Properties: Continuous response variable (mpg), multiple predictors

ANOVA:

ANCOVA:

```
data(ToothGrowth)
ancova_model <- lm(len ~ supp + dose + supp:dose, data = ToothGrowth)
summary(ancova_model)</pre>
```

```
Call:
lm(formula = len ~ supp + dose + supp:dose, data = ToothGrowth)
Residuals:
   Min
            1Q Median
                          3Q
                                 Max
-8.2264 -2.8462 0.0504 2.2893 7.9386
Coefficients:
          Estimate Std. Error t value Pr(>|t|)
            (Intercept)
suppVC
            -8.255
                      2.236 -3.691 0.000507 ***
                       1.195 6.534 2.03e-08 ***
dose
             7.811
             3.904
                       1.691 2.309 0.024631 *
suppVC:dose
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 4.083 on 56 degrees of freedom
Multiple R-squared: 0.7296, Adjusted R-squared: 0.7151
F-statistic: 50.36 on 3 and 56 DF, p-value: 6.521e-16
# Properties: Continuous response variable, continuous and
# categorical predictors
Generalized Linear Model (Poisson regression):
data(warpbreaks)
glm_model <- glm(breaks ~ wool + tension, family = poisson, data = warpbreaks)</pre>
summary(glm_model)
Call:
glm(formula = breaks ~ wool + tension, family = poisson, data = warpbreaks)
Coefficients:
           Estimate Std. Error z value Pr(>|z|)
(Intercept) 3.69196 0.04541 81.302 < 2e-16 ***
                      0.05157 -3.994 6.49e-05 ***
woolB
          -0.20599
tensionM -0.32132 0.06027 -5.332 9.73e-08 ***
          tensionH
```

```
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
(Dispersion parameter for poisson family taken to be 1)
    Null deviance: 297.37 on 53 degrees of freedom
Residual deviance: 210.39 on 50 degrees of freedom
AIC: 493.06
Number of Fisher Scoring iterations: 4
# Properties: The warpbreaks dataset is used because it has a
#count response variable (breaks), suitable for Poisson regression.
Linear Probability Model:
lpm_model <- lm(am ~ wt + hp + qsec, data = mtcars)</pre>
summary(lpm_model)
Call:
lm(formula = am ~ wt + hp + qsec, data = mtcars)
Residuals:
               1Q Median
    Min
                                 3Q
                                        Max
-0.47595 -0.24600 -0.01487 0.28334 0.50096
Coefficients:
              Estimate Std. Error t value Pr(>|t|)
(Intercept) 3.476e+00 1.076e+00 3.232 0.003143 **
           -3.831e-01 9.617e-02 -3.984 0.000439 ***
wt
           -6.007e-05 1.914e-03 -0.031 0.975187
hp
            -1.025e-01 5.612e-02 -1.826 0.078532 .
qsec
Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
Residual standard error: 0.3293 on 28 degrees of freedom
Multiple R-squared: 0.6065,
                              Adjusted R-squared: 0.5644
F-statistic: 14.39 on 3 and 28 DF, p-value: 7.347e-06
```

[#] Properties: The mtcars dataset has a binary response variable (am) and # several predictors, suitable for Linear Probability Model.

```
Probit Model:
```

```
probit_model <- glm(am ~ wt + hp + qsec, family = binomial(link = "probit"), data = mtcars)</pre>
Warning: glm.fit: algorithm did not converge
Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
summary(probit_model)
Call:
glm(formula = am ~ wt + hp + qsec, family = binomial(link = "probit"),
    data = mtcars)
Coefficients:
             Estimate Std. Error z value Pr(>|z|)
(Intercept) 6.808e+02 1.890e+05 0.004 0.997
           -6.612e+01 1.737e+04 -0.004
wt
                                            0.997
hp
           -2.137e-01 9.226e+01 -0.002 0.998
           -2.515e+01 7.283e+03 -0.003 0.997
qsec
(Dispersion parameter for binomial family taken to be 1)
    Null deviance: 4.3230e+01 on 31 degrees of freedom
Residual deviance: 8.3545e-09 on 28 degrees of freedom
AIC: 8
Number of Fisher Scoring iterations: 25
# Properties: The mtcars dataset has a binary response variable (am)
# and several predictors, suitable for Probit regression.
```

Truncated Regression:,

```
data("Affairs", package="AER")
fit <- truncreg(affairs ~ age + yearsmarried + religiousness + rating, data=Affairs, point=0
summary(fit)
```

```
Call:
truncreg(formula = affairs ~ age + yearsmarried + religiousness +
    rating, data = Affairs, point = 0, direction = "left")
BFGS maximization method
75 iterations, 0h:0m:0s
g'(-H)^-1g = 0.301
Coefficients:
             Estimate Std. Error t-value Pr(>|t|)
              30.6816 41.4946 0.7394 0.45966
(Intercept)
                         1.5007 -1.5926 0.11126
              -2.3899
age
yearsmarried 17.0571
                         8.1109 2.1030 0.03547 *
religiousness -57.6926
                         26.7378 -2.1577 0.03095 *
             -55.4266
                         25.2016 -2.1993 0.02785 *
rating
              17.7881
                         4.4007 4.0421 5.297e-05 ***
sigma
Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
Log-Likelihood: -681.72 on 6 Df
# Properties: The Affairs dataset contains a continuous response variable
# (affairs) that is truncated, suitable for truncated regression.
Censored Regression (Tobit Model):
fit <- censReg(affairs ~ age + yearsmarried + religiousness + rating, data=Affairs, left=0)
summary(fit)
Call:
censReg(formula = affairs ~ age + yearsmarried + religiousness +
    rating, left = 0, data = Affairs)
Observations:
        Total Left-censored
                                 Uncensored Right-censored
           601
                         451
                                        150
```

```
Coefficients:
```

```
Estimate Std. error t value Pr(> t)

(Intercept) 9.08289 2.65881 3.416 0.000635 ***

age -0.16034 0.07772 -2.063 0.039095 *

yearsmarried 0.53890 0.13417 4.016 5.91e-05 ***

religiousness -1.72337 0.40471 -4.258 2.06e-05 ***

rating -2.26735 0.40813 -5.556 2.77e-08 ***

logSigma 2.11310 0.06712 31.482 < 2e-16 ***

---

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

Newton-Raphson maximisation, 7 iterations
```

Return code 1: gradient close to zero (gradtol)
Log-likelihood: -706.4048 on 6 Df

Properties: The Affairs dataset has a continuous response variable (affairs) #that is censored, suitable for a Tobit model.

Poisson Model:

```
model <- glm(breaks ~ wool + tension, data=warpbreaks, family=poisson)
summary(model)</pre>
```

Call:

```
glm(formula = breaks ~ wool + tension, family = poisson, data = warpbreaks)
```

Coefficients:

```
Estimate Std. Error z value Pr(>|z|)

(Intercept) 3.69196    0.04541   81.302   < 2e-16 ***

woolB     -0.20599    0.05157   -3.994 6.49e-05 ***

tensionM     -0.32132    0.06027   -5.332 9.73e-08 ***

tensionH     -0.51849    0.06396   -8.107 5.21e-16 ***

---

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

(Dispersion parameter for poisson family taken to be 1)
```

Null deviance: 297.37 on 53 degrees of freedom Residual deviance: 210.39 on 50 degrees of freedom

AIC: 493.06

Number of Fisher Scoring iterations: 4

```
# Properties: The warpbreaks dataset has a count response variable (breaks), #suitable for a Poisson model.
```

Negative-Binomial Model:

```
data("quine", package="MASS")
model2 <- glm.nb(Days ~ Sex + Age + Lrn + Eth, data=quine)
summary(model2)</pre>
```

Call:

```
glm.nb(formula = Days ~ Sex + Age + Lrn + Eth, data = quine,
    init.theta = 1.274892646, link = log)
```

Coefficients:

```
Estimate Std. Error z value Pr(>|z|)

(Intercept) 2.89458 0.22842 12.672 < 2e-16 ***

SexM 0.08232 0.15992 0.515 0.606710

AgeF1 -0.44843 0.23975 -1.870 0.061425 .

AgeF2 0.08808 0.23619 0.373 0.709211

AgeF3 0.35690 0.24832 1.437 0.150651

LrnSL 0.29211 0.18647 1.566 0.117236

EthN -0.56937 0.15333 -3.713 0.000205 ***
```

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

(Dispersion parameter for Negative Binomial(1.2749) family taken to be 1)

Null deviance: 195.29 on 145 degrees of freedom Residual deviance: 167.95 on 139 degrees of freedom

AIC: 1109.2

Number of Fisher Scoring iterations: 1

Theta: 1.275 Std. Err.: 0.161

2 x log-likelihood: -1093.151

```
# Properties: The quine dataset has a count response variable (Days), suitable
# for a negative-binomial model due to overdispersion.
```

Zero-inflated Model:

```
data("bioChemists", package="pscl")
model3 <- zeroinfl(art ~ fem + mar + kid5 + phd + ment, data=bioChemists, dist="poisson")</pre>
summary(model3)
Call:
zeroinfl(formula = art ~ fem + mar + kid5 + phd + ment, data = bioChemists,
   dist = "poisson")
Pearson residuals:
          1Q Median
                       3Q
                             Max
-2.3253 -0.8652 -0.2826 0.5404 7.2976
Count model coefficients (poisson with log link):
          Estimate Std. Error z value Pr(>|z|)
(Intercept) 0.640838 0.121307 5.283 1.27e-07 ***
femWomen
         marMarried 0.103751 0.071111 1.459 0.144565
         kid5
phd
         -0.006166 0.031008 -0.199 0.842378
          ment
Zero-inflation model coefficients (binomial with logit link):
          Estimate Std. Error z value Pr(>|z|)
(Intercept) -0.577059 0.509386 -1.133 0.25728
femWomen
          marMarried -0.354014 0.317611 -1.115 0.26502
kid5
          0.217097 0.196482 1.105 0.26919
          0.001274 0.145263 0.009 0.99300
phd
         ment
Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
Number of iterations in BFGS optimization: 21
Log-likelihood: -1605 on 12 Df
```

Properties: The bioChemists dataset can be used with zero-inflated models # due to the nature of count data with potential excess zeros.

Quantile Regression:

```
fit <- rq(medv ~ ., data=Boston, tau=0.5)
summary(fit)</pre>
```

Call: rq(formula = medv ~ ., tau = 0.5, data = Boston)

tau: [1] 0.5

Coefficients:

	coefficients	lower bd	upper bd
(Intercept)	14.85002	6.34690	25.81307
crim	-0.14446	-0.15448	-0.02408
zn	0.03703	0.01835	0.06029
indus	0.02166	-0.05594	0.06244
chas	1.30227	0.60041	2.52568
nox	-9.18412	-14.87307	-2.76601
rm	5.32517	4.09059	6.44724
age	-0.03135	-0.04396	-0.00409
dis	-1.04478	-1.24708	-0.75657
rad	0.18003	0.07576	0.28132
tax	-0.00994	-0.01472	-0.00559
ptratio	-0.73731	-0.91502	-0.56798
black	0.01125	0.00820	0.01662
lstat	-0.29766	-0.40363	-0.21898

[#] Proporties: The Boston housing dataset is used because it has a continuous # response variable and several predictors, suitable for quantile regression.