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#### STAT 641 Assignment 4 ####
# P1: (50 points) A researcher is studying the relative brain weights (brain
weight divided by body weight)
# for 51 species of mammals whose litter size is 1 and for 44 species of
mammals whose average litter size is
# greater thanor equal to 2. The researcher was interested in determining
what evidence that brain sizes
# tend to be differentfor the two groups. (Data fromThe Statistical Sleuthby
Fred Ramsey and Daniel Schafer).
BrainSize = read.csv("C:/Users/jackr/OneDrive/Desktop/Graduate School Courses/
STAT 641 - Methods of STAT I/RawData/Assign3 BrainSize.csv")
BrainSize = data.frame(SmallLitter = BrainSize$:..Small.Litter.Size,
LargeLitter = BrainSize$Large.Litter.Size)
LLitter = na.omit(BrainSize$LargeLitter)
SLitter = BrainSize$SmallLitter
#1.) For the Small Litter Size mammals, answer the following questions: The
data is given in the file:
     BrainWeight Data.txt in Canvas
#(a) Compute a 10% trimmed mean, and compare it to the untrimmed sample mean.
Does this comparison suggest
     any extreme values in the data?
mean(BrainSize$SmallLitter)
mean(BrainSize$SmallLitter, trim = .10)
# the trimmed mean and the untrimmed sample mean are very similar. This would
suggest there are few (if any)
# extreme values in the data
#(b) The researcher suggested a Weibull distribution to model the data for
the Small Litter Size mammals.
     Assuming that the Weibull distribution is an appropriate model for the
Small Litter Size data,
     obtainthe MLE estimates of the Weibull parameters for the Small Litter
Size data.
library (MASS)
mle weibull=fitdistr(BrainSize$SmallLitter, "weibull", lower=c(0,0))
# shape:
mle weibull$estimate[1]
# scale:
mle weibull$estimate[2]
#(c) Estimate the probability that a randomly selected mammal with a litter
size of 1 will have a relative brain
     weight greater than 15, first using the Weibull model and
secondly using a distribution-free
    estimate.
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# Weibull:
1 - pweibull(15, shape = mle weibull$estimate[1], scale =
mle weibull$estimate[2])
exp(-(15/mle weibull$estimate[2])^mle weibull$estimate[1])
# Distribution Free Estimate:
1-ecdf(BrainSize$SmallLitter)(15)
\#(d) Compare the MLE estimates of \mu and \sigma based on the Weibull model to the
distribution-free
     estimates of \boldsymbol{\mu} and \boldsymbol{\sigma} for the Small Litter Size data.
lambda = mle weibull$estimate[2]
k = mle weibull$estimate[1]
lambda*gamma(1+1/k)
mean(BrainSize$SmallLitter)
sqrt((lambda^2)*(gamma(1+2/k) - (gamma(1+1/k)^2)))
sd(BrainSize$SmallLitter)
# the MLE estimates and the distribution free estimates are very similar
# (e) Compare the MLE estimates of median and IQR based on the Weibull model
to the distribution-free estimates
      of median and IQR for the Small Litter Size data.
# Distribution Free
median(BrainSize$SmallLitter)
IQR(BrainSize$SmallLitter)
quantile(BrainSize$SmallLitter)
# MLE Estimates
qweibull(.5, shape = mle weibull$estimate[1], scale = mle weibull$estimate[2])
qweibull(c(.25,.75), shape = mle weibull$estimate[1], scale =
mle weibull$estimate[2])
#2.) Without any assumed model, estimate the mean and standard deviation of
the relative brain weights forboth Large
    and Small litter sizes.
mean(LLitter)
sd(LLitter)
mean(BrainSize$SmallLitter)
sd(BrainSize$SmallLitter)
# 3.) Estimate the median and MAD of the relative brain weights for both Large
and Small litter sizes.
absDevS = NULL
for(i in seg along(SLitter)){
  absDevS[i] = abs(SLitter[i]-quantile(BrainSize$SmallLitter,.5))
quantile(absDevS, 0.5)/.6745
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absDevL = NULL
for(i in seq along(LLitter)){
  absDevL[i] = abs(LLitter[i]-quantile(LLitter,.5))
quantile(absDevL, 0.5)/.6745
# 4.) Based on your plots from Assignment #3, which pair of estimates of the
center and spread in the two
     datasets best represents the center and spread in the two populations of
relative brain weights?
# The sample mean seems to be the best estimate of the center and the MAD
seems to be the best estimate of
# the spread because we have highly skewed data
# 5.) Using your answers from the previous three questions, suggest a
relationship (if any) between litter
     size andrelative brain weights.
# Based on the previous three questions, I would say that there is a positive
relationship between average
# litter size and relative brain weights.
# P2. ( 30 points) Twenty-five patients diagnosed with rare skin disease are
randomly assigned to two drug
     treat-ments. The following times are either the time in days
from the point of
     randomization to either a completerecovery or censoring (as
indicated by the status variable:
     O means censored, i.e., time at which patient left studyprior to a
complete recovery, 1 means patient's
   time to recovery).
library(survival)
Time = c(180, 632, 2240, 195, 76, 70, 13, 1990, 18, 700, 210, 1296, 23)
Status = c(1,1,1,1,1,1,0,0,1,1,1,1,1)
Time2 = c(8, 852, 52, 220, 63, 8, 1976, 1296, 1460, 63, 1328, 365)
Status2 = c(0,1,1,1,1,1,0,0,1,1,1,1)
# 1.) Estimate the survival function for the two treatments.
Surv(Time, Status)
cords.surv1 <- survfit(Surv(Time, Status) ~ 1,conf.type="log-log")</pre>
summary(cords.surv1)
print(cords.surv1,print.rmean=TRUE)
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Surv(Time2, Status2)
cords.surv2 <- survfit(Surv(Time2, Status2) ~ 1,conf.type="log-log")</pre>
summary(cords.surv2)
print(cords.surv2,print.rmean=TRUE)
# 2.) Compute the mean and median time to recovery for the two treatments
using the estimated survival function.
# Treatment 1: mean - 657, median - 202
# Treatment 2: mean - 731, median - 365
# 3.) Which treatment appears to be most effective in the treatment of the
skin disease?
# Treatment 1 seems to be most effective in the treatment of the skin disease
as the mean recovery
# time is lower than that of Treatment 2
# 4.) Estimate the mean and median time to recovery ignoring the censoring and
compare these values to values
    obtained in part 2.
mean(cords.surv1$time)
median(cords.surv1$time)
mean(cords.surv2$time)
median(cords.surv2$time)
\# the mean and median for treatment 1 is lower and the mean and median for
treatment 2 is higher
# P3. (20 points) Selectthe letter of the BESTanswer.
# 1.) A
# 2.) D
# 3.) B
# 4.) C
# 5.) B
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