HW1 Answer Key

Jan 14 2016

Problem 1

(a) $f(y;k) = k \frac{y^{k-1}}{y_{max}^k}$, taking log and we have $\log f(y;k) = (k-1)\log y + \log\left(\frac{k}{y_{max}^k}\right)$. Therefore, $a(y) = \log(y)$, b(k) = k-1, $c(k) = \log\left(\frac{k}{y_{max}^k}\right)$, and thus it belongs to exponential family.

(b)
$$E[Y] = \int_0^{y_{max}} y f(y) dy = \frac{k}{y_{max}^k} \int_0^{y_{max}} y^k dy = \frac{k}{y_{max}^k} \frac{y_{max}^{k+1}}{k+1} = \frac{k}{k+1} y_{max}$$

$$\operatorname{Var}\left[Y\right] = \operatorname{E}\left[Y^2\right] - (\operatorname{E}\left[Y\right])^2 = \int_0^{y_{max}} y^2 f(y) dy - (\operatorname{E}\left[Y\right])^2 = \frac{k}{k+2} y_{max}^2 - \frac{k^2}{(k+1)^2} y_{max}^2 = \frac{k}{(k+1)^2(k+2)} y_{max}^2 = \frac{k}{(k+1)^2(k+$$

(c)

```
ppareto<-function(x,ymax,k){(x/ymax)^k}
x<-seq(0,10,0.001)
plot(x,dpareto(x,10,0.5),type="l",col=2,ylab="pdf of x",main="Density of
    Pareto Distributions with Different Parameters")
lines(x,dpareto(x,10,1),col=3)</pre>
```

lines(x,dpareto(x,10,2),col=4)
legend("topright",pch=0,c("k=0.5", "k=1", "k=2"), col = 2:4)

dpareto<-function(x,ymax,k){k*(x^(k-1))/(ymax^k)}</pre>

plot(x,ppareto(x,10,0.5),type="l",col=2,ylab="cdf of x",main="Cumulative Density of Pareto Distributions with Different Parameters")

lines(x,ppareto(x,10,1),col=3)

lines(x,ppareto(x,10,2),col=4)

legend("bottomright",pch=0,c("k=0.5", "k=1", "k=2"), col = 2:4)

Notes: we first define the pdf and cdf function of Pareto distribution, then, generate a grid (x) and plot corresponding functions. Use lines to append another line on existing plot, and use legend to attach legend.

Problem 2

(a)

```
d<-read.dta("your directory/org_example.dta")
d<-subset(d,year==2008)</pre>
```

Density of Pareto Distributions with Different Parameters

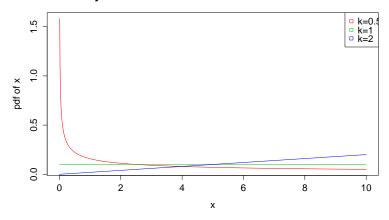


Figure 1: Plot

Cumulative Density of Pareto Distributions with Different Parameters

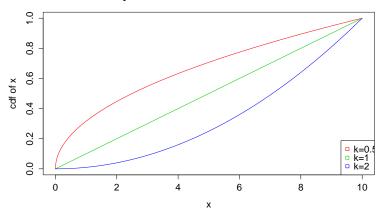


Figure 2: Plot

(b)
d<-subset(d,nilf==0)</pre>

(c) By typing

glm_probit<-glm(unem~educ,d,family=binomial(link="probit"))
summary(glm_probit)</pre>

you will get the following result

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glm(formula = unem ~ educ, family = binomial(link = "probit"),
 data = d)

Deviance Residuals:

```
Min 1Q Median 3Q Max
-0.5008 -0.3705 -0.3095 -0.2427 2.7574
```

Coefficients:

```
Estimate Std. Error z value Pr(>|z|)
(Intercept)
                -1.18579
                             0.01653 -71.75
                                               <2e-16 ***
educHS
                 -0.31792
                             0.01976 -16.09
                                               <2e-16 ***
educSome college -0.49126
                             0.02069 -23.75
                                               <2e-16 ***
educCollege
                -0.70958
                             0.02397 -29.60
                                               <2e-16 ***
educAdvanced
                -0.82203
                             0.03122 -26.33
                                               <2e-16 ***
Signif. codes: 0 *** 0.001 ** 0.01 * 0.05 . 0.1
```

(Dispersion parameter for binomial family taken to be 1)

Null deviance: 43241 on 104043 degrees of freedom Residual deviance: 41944 on 104039 degrees of freedom

AIC: 41954

Number of Fisher Scoring iterations: 6

A person without any degree has the expectation of unemployment rate of 11.79% (pnorm(-1.18579)). A high school degree decreases the unemployment rate by 5.15% (pnorm(-1.18579)-pnorm(-1.18579-0.31792)). Having some college decreases the unemployment rate by 7.11% (pnorm(-1.18579)-pnorm(-1.18579-0.49126)). A college degree decreases the unemployment rate by 8.88% (pnorm(-1.18579)-pnorm(-1.18579-0.70958)). Finally, if you have an advanced degree (Master or higher), your expected unemployment rate will decrease by 9.55% (pnorm(-1.18579)-pnorm(-1.18579-0.82203)). These effects are all statistically significant.

(d) The marginal effect of an advanced degree will reduce the unemployment rate by 0.67% (pnorm(-1.18579-0.70958)-pnorm(-1.18579-0.82203)).