MATH 668 Homework 1 Solutions

1. The following R commands can be used to obtain the answers.

```
require(HistData)
## Loading required package: HistData
attach(GaltonFamilies)
mean(childHeight)
## [1] 66.74593
 (a) The average height of all children is 66.74 inches.
mean(gender=="female")
## [1] 0.4850107
 (b) The percentage of female children is 48.50%.
max(childHeight)
## [1] 79
 (c) The height of the tallest child is 79 inches.
sum(children==4)/4
## [1] 31
 (d) The number of families with exactly 4 children is 31.
tallest.mother.height=max(mother)
GaltonFamilies[mother==tallest.mother.height,]
##
       family father mother midparentHeight children childNum gender
## 571
           128
                 68.7
                         70.5
                                          72.42
                                                        2
                                                                       male
                 68.7
                         70.5
                                          72.42
                                                        2
                                                                   2 female
## 572
           128
##
       childHeight
## 571
               71.0
## 572
               61.7
 (e) The heights of the children in the family with the tallest mother are 71 and 61.7 inches.
mean((childHeight>father)&(childHeight>mother))
## [1] 0.238758
  (f) The percentage of children that are taller than both of their parents is 23.88%.
mean(childHeight[(gender=="female")&(mother>=68)])
## [1] 66.46667
 (g) The average height of females whose mother was at least 68 inches tall is 66.47 inches.
detach(GaltonFamilies)
```

2.(a) Since

$$\mathbf{H} \left(\alpha_1 \mathbf{H}_1^- + \ldots + \alpha_m \mathbf{H}_m^- \right) \mathbf{H} = \mathbf{H} (\alpha_1 \mathbf{H}_1^-) \mathbf{H} + \ldots + \mathbf{H} (\alpha_m \mathbf{H}_m^-) \mathbf{H}$$

$$= \alpha_1 \mathbf{H} \mathbf{H}_1^- \mathbf{H} + \ldots + \alpha_m \mathbf{H} \mathbf{H}_m^- \mathbf{H}$$

$$= \alpha_1 \mathbf{H} + \ldots + \alpha_m \mathbf{H}$$

$$= (\alpha_1 + \ldots + \alpha_m) \mathbf{H},$$

 $\alpha_1 \mathbf{H}_1^- + \ldots + \alpha_m \mathbf{H}_m^-$ is a generalized inverse of **H** if and only if $\alpha_1 + \ldots + \alpha_m = 1$.

(b) Since
$$\mathbf{H}_{1}^{-} = \begin{pmatrix} 1/4 & 1/4 & 0 \\ -1/2 & 1/2 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$
, $\mathbf{H}_{2}^{-} = \begin{pmatrix} 2 & 0 & 3 \\ 0 & 0 & 0 \\ -1 & 0 & -2 \end{pmatrix}$, and $\mathbf{H}_{3}^{-} = \begin{pmatrix} 0 & 0 & 0 \\ 0 & 2/3 & 1/3 \\ 0 & 1/3 & -1/3 \end{pmatrix}$ are generalized inverses of \mathbf{H} ,

$$\alpha_1 \mathbf{H}_1^- + \alpha_2 \mathbf{H}_2^- + \alpha_3 \mathbf{H}_3^- = \begin{pmatrix} \frac{1}{4}\alpha_1 + 2\alpha_2 & \frac{1}{4}\alpha_1 & 3\alpha_2 \\ -\frac{1}{2}\alpha_1 & \frac{1}{2}\alpha_1 + \frac{2}{3}\alpha_3 & +\frac{1}{3}\alpha_3 \\ -\alpha_2 & \frac{1}{3}\alpha_3 & -2\alpha_2 - \frac{1}{3}\alpha_3 \end{pmatrix}$$

is a generalized inverse of **H** if $\alpha_1 + \alpha_2 + \alpha_3 = 1$. So, one example with no zero elements is

$$4\mathbf{H}_1^- + 3\mathbf{H}_2^- - 6\mathbf{H}_3^- = \begin{pmatrix} 7 & 1 & 9 \\ -2 & -2 & -2 \\ -3 & -2 & -4 \end{pmatrix}.$$

3. First, we read the data into R.

```
0=rbind(
c(255, 54, 71, 87, 88, 93,255),
c(54,255,255,255,255,255,97),
c(71,255,255,255,255,255,101),
c(87,255,255,255,255,255,103),
c(88,255,255,255,255,255,108),
c(93,255,255,255,255,255,115),
c(255, 97,101,103,108,115,255))
```

The eigenvalues and eigenvectors can be obtained with the following commands.

```
E=eigen(0)
l=E$val;1
```

```
## [1] 1.373685e+03 4.151384e+02 1.289019e+01 1.256982e-13 2.620790e-14 ## [6] -5.587959e-01 -1.615523e+01
```

v=E\$vec;v

```
##
              [,1]
                         [,2]
                                     [,3]
                                                   [,4]
                                                                 [,5]
## [1,] -0.2049720  0.6922375 -0.45765380
                                          0.000000e+00 0.000000e+00
## [2,] -0.4186751 -0.1949149 0.53020254 3.521904e-01 -2.284790e-01
## [3,] -0.4219260 -0.1603606 0.09424771 -7.566207e-01 3.906608e-01
## [4,] -0.4246705 -0.1305774 -0.39000992 4.970539e-02 -4.997859e-01
## [5,] -0.4257125 -0.1211512 -0.21599687 5.223998e-01 6.629717e-01
## [6,] -0.4277085 -0.1019515 -0.10019312 -1.676749e-01 -3.253676e-01
## [7,] -0.2452864 0.6441876 0.54014312 1.249001e-14 6.938894e-16
##
               [,6]
                          [,7]
## [1,] -0.02098782 0.5185617
## [2,]
        0.09260833 0.5663812
## [3,]
        0.20737922 0.1388647
## [4,] 0.54645723 -0.3156338
## [5,] -0.07064908 -0.2000308
```

```
## [6,] -0.79938550 -0.1537420
## [7,] 0.07316486 -0.4772307
```

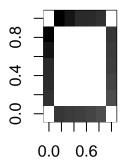
So, the two largest eigenvalues are 1373.685 and 415.1384. Then, we compute **A** as follows.

```
A=1[1]*v[,1]%*%t(v[,1])+1[2]*v[,2]%*%t(v[,2])
A
```

```
##
             [,1]
                                                               [,6]
                                                                         [,7]
                       [,2]
                                 [,3]
                                           [,4]
                                                     [,5]
## [1,] 256.64467 61.87156 72.71689 82.04861 85.05086 91.13034 254.18759
## [2,]
        61.87156 256.56357 255.63722 254.80571 254.64227 254.23665 88.94556
        72.71689 255.63722 255.22106 254.82905 254.80547 254.68418 99.28166
## [3,]
## [4,]
        82.04861 254.80571 254.82905 254.81563 254.91253 255.03615 108.17127
## [5,]
        85.05086 254.64227 254.80547 254.91253 255.04781 255.24942 111.04319
## [6,] 91.13034 254.23665 254.68418 255.03615 255.24942 255.60953 116.85023
## [7,] 254.18759 88.94556 99.28166 108.17127 111.04319 116.85023 254.92156
```

The following figure compares the two images.

```
image(t(0[7:1,]),col=gray((0:255)/255))
```



image(t(A[7:1,]),col=gray((0:255)/255))

