## R Code for GLMs and model comparison in Dalziel et al. 2018

Read data: flu <- read.csv("influenza cities.csv")</pre> names(flu) ## [1] "zip" "population\_size" "kappa" ## [4] "nu" "avgsh" "sdsh" ## [7] "lon" "lat" "h crowding" ## [10] "w\_crowding" keep <- complete.cases(flu) flu <- flu[keep,] rm(keep) Fit crowding as a function of population size: #residential crowding: hcrowd\_pop <- glm(log(h\_crowding) ~ log(population\_size), data = flu) summary(hcrowd\_pop) ## ## Call: ## glm(formula = log(h\_crowding) ~ log(population\_size), data = flu) ## Deviance Residuals: ## Min 10 Median 30 Max ## -1.22610 -0.28681 -0.03627 0.30772 2.29737 ## ## Coefficients: ## Estimate Std. Error t value Pr(>|t|) ## (Intercept) -0.63869 0.28781 -2.219 0.0269 \* ## log(population\_size) 0.41169 0.02264 18.187 <2e-16 \*\*\* ## Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.05 '.' 0.1 ' ' 1 ## (Dispersion parameter for gaussian family taken to be 0.2014433) ## ## Null deviance: 186.90 on 598 degrees of freedom ## Residual deviance: 120.26 on 597 degrees of freedom ## AIC: 744.14 ## Number of Fisher Scoring iterations: 2 #daytime crowding: wcrowd\_pop <- glm(log(w\_crowding) ~ log(population\_size), data = flu)</pre> summary(wcrowd\_pop) ## ## Call: ## glm(formula = log(w\_crowding) ~ log(population\_size), data = flu)

```
## Deviance Residuals:
##
      Min
           10 Median
                                   30
                                           Max
## -1.1860 -0.5105 -0.1579 0.3313
                                        3.7485
##
## Coefficients:
                        Estimate Std. Error t value Pr(>|t|)
##
                        -0.17904
                                    0.46372 -0.386
## (Intercept)
## log(population_size) 0.53780
                                    0.03647 14.746
                                                      <2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## (Dispersion parameter for gaussian family taken to be 0.5229336)
##
       Null deviance: 425.90 on 598 degrees of freedom
##
## Residual deviance: 312.19 on 597 degrees of freedom
## AIC: 1315.6
##
## Number of Fisher Scoring iterations: 2
Fit base transmission potential (kappa) as a function of population size and crowding
kappa_pop <- glm(log(kappa) ~ log(population_size), data = flu)</pre>
summary(kappa_pop)
##
## Call:
## glm(formula = log(kappa) ~ log(population_size), data = flu)
##
## Deviance Residuals:
##
       Min
                  1Q
                        Median
                                       3Q
                                                Max
## -0.43489 -0.07958
                       0.00826
                                 0.07874
                                            0.37473
##
## Coefficients:
                        Estimate Std. Error t value Pr(>|t|)
##
## (Intercept)
                        -1.050138
                                    0.075369 -13.93
                                                       <2e-16 ***
## log(population_size) 0.231399
                                    0.005928
                                               39.04
                                                       <2e-16 ***
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for gaussian family taken to be 0.01381437)
##
##
       Null deviance: 29.2983 on 598 degrees of freedom
## Residual deviance: 8.2472 on 597 degrees of freedom
## AIC: -861.06
##
## Number of Fisher Scoring iterations: 2
kappa_popXcrowd <- glm(log(kappa) ~ log(population_size)*log(h_crowding)*log(w_crowding), data = flu)
summary(kappa_popXcrowd)
##
## Call:
## glm(formula = log(kappa) ~ log(population_size) * log(h_crowding) *
       log(w_crowding), data = flu)
##
## Deviance Residuals:
```

```
Median
##
                   1Q
                                                 Max
                                  0.07629
## -0.45469 -0.07227
                        0.00981
                                             0.34636
##
## Coefficients:
##
                                                          Estimate Std. Error
## (Intercept)
                                                          6.463602
                                                                     3.644330
## log(population_size)
                                                         -0.371283
                                                                     0.300631
## log(h_crowding)
                                                         -1.593476
                                                                     0.739319
## log(w_crowding)
                                                         -0.838555
                                                                     0.578620
## log(population_size):log(h_crowding)
                                                          0.127801
                                                                     0.060458
## log(population_size):log(w_crowding)
                                                          0.063371
                                                                     0.046952
## log(h_crowding):log(w_crowding)
                                                          0.184695
                                                                     0.114347
## log(population_size):log(h_crowding):log(w_crowding) -0.013984
                                                                     0.009224
                                                         t value Pr(>|t|)
##
                                                           1.774
## (Intercept)
                                                                   0.0766 .
## log(population_size)
                                                          -1.235
                                                                   0.2173
## log(h_crowding)
                                                          -2.155
                                                                   0.0315 *
## log(w_crowding)
                                                          -1.449
                                                                   0.1478
## log(population_size):log(h_crowding)
                                                           2.114
                                                                   0.0349 *
## log(population_size):log(w_crowding)
                                                           1.350
                                                                   0.1776
## log(h_crowding):log(w_crowding)
                                                           1.615
                                                                   0.1068
## log(population_size):log(h_crowding):log(w_crowding) -1.516
                                                                   0.1300
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
##
   (Dispersion parameter for gaussian family taken to be 0.01285076)
##
##
       Null deviance: 29.2983
                               on 598
                                       degrees of freedom
## Residual deviance: 7.5948
                               on 591
                                       degrees of freedom
## AIC: -898.42
##
## Number of Fisher Scoring iterations: 2
```

Define predicted kappas from census data using both population size and crowding to use when we want a measure of kappa that derives from census data (not incidence data):

```
flu$kpred <- kappa_popXcrowd$fitted.values</pre>
```

Define excess kappa as the residual variation in base transmission not explained by population size and excess crowding as the residual variation in crowding not explained by population size:

```
flu$exkappa <- kappa_pop$residuals
flu$exhcrowd <- hcrowd_pop$residuals
flu$exwcrowd <- wcrowd_pop$residuals</pre>
```

Fit excess kappa as a function of excess crowding:

```
exkappa_exhcrowd <- glm(exkappa ~ exhcrowd, data = flu)
summary(exkappa_exhcrowd)</pre>
```

```
##
## Coefficients:
##
                Estimate Std. Error t value Pr(>|t|)
## (Intercept) 1.185e-15 4.694e-03
                                       0.000
## exhcrowd
               5.540e-02 1.048e-02
                                       5.289 1.73e-07 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for gaussian family taken to be 0.01319609)
##
##
       Null deviance: 8.2472 on 598 degrees of freedom
## Residual deviance: 7.8781 on 597 degrees of freedom
## AIC: -888.49
##
## Number of Fisher Scoring iterations: 2
exkappa_exwcrowd <- glm(exkappa ~ exwcrowd, data = flu)
summary(exkappa_exwcrowd)
##
## Call:
## glm(formula = exkappa ~ exwcrowd, data = flu)
## Deviance Residuals:
        Min
                         Median
                                        30
                   10
                                                  Max
                         0.00679
## -0.44068 -0.07568
                                   0.07955
                                              0.36937
##
## Coefficients:
                Estimate Std. Error t value Pr(>|t|)
##
## (Intercept) 1.284e-15 4.783e-03
                                        0.00
                                                1.0000
## exwcrowd
               1.444e-02 6.626e-03
                                               0.0297 *
                                        2.18
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## (Dispersion parameter for gaussian family taken to be 0.01370529)
##
       Null deviance: 8.2472 on 598 degrees of freedom
##
## Residual deviance: 8.1821 on 597 degrees of freedom
## AIC: -865.81
## Number of Fisher Scoring iterations: 2
Fit epidemic intensity to population size, transmission potential (estimated independently from census data),
and specific humidity:
nu_pop <- glm(log(nu) ~ log(population_size), data = flu)</pre>
nu_sh \leftarrow glm(log(nu) \sim avgsh * sdsh + I(avgsh^2) + I(sdsh^2), data = flu)
nu_kappa <- glm(log(nu) ~ log(kpred), data = flu)</pre>
nu_popXsh <- glm(log(nu) ~ log(population_size) * avgsh * sdsh + I(avgsh^2) + I(sdsh^2), data = flu)
nu_kappaXsh <- glm(log(nu) ~ log(kpred) * avgsh * sdsh + I(avgsh^2) + I(sdsh^2), data = flu)</pre>
Do model comparison for the intensity GLMs
modcomp <- AIC(nu_pop,nu_sh,nu_kappa,nu_popXsh,nu_kappaXsh)</pre>
modcomp$Model <- c("Population size (P)", "Specific humidity (SH)", "Base transmission (K)", "P x SH",
modcomp \leftarrow modcomp[,c(3,1,2)]
```

```
rownames(modcomp) <- NULL</pre>
modcomp <- modcomp[order(modcomp$AIC),]</pre>
modcomp$deltaAIC <- modcomp$AIC - modcomp$AIC[1]</pre>
modcomp
##
                       Model df
                                       AIC deltaAIC
## 5
                      K x SH 11 931.4079
                                             0.0000
## 4
                      P x SH 11 969.3593 37.9514
## 2 Specific humidity (SH)
                             7 1121.0814 189.6735
## 3 Base transmission (K)
                              3 1142.9572 211.5493
        Population size (P)
## 1
                              3 1183.9871 252.5793
Bar plot of deltaAICs for intensity models
pal <- 1 #brewer.pal(5, "Set1")</pre>
x <- barplot(modcomp$deltaAIC,xlab = expression(paste(Delta,"AIC")),ylab="Model",srt=45,horiz=T,col=pal
text(rep(30,5),x,modcomp$Model,pos=4,col='white', cex=0.95)
text(35,x[2],modcomp$Model[2],pos=4,font=1,cex=0.95)
text(35,x[1],modcomp$Model[1],pos=4,font=2,cex=0.95)
                      Population size (P)
                      Base transmission (K)
                      Specific humidity (SH)
                       P x SH
                       K x SH
           0
                         50
                                       100
                                                      150
                                                                     200
                                                                                    250
```

ΔΑΙΟ