

WMAN 633 Homework 7

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Question 1

Load data and place into an unmarkedFrameOccu object

```
#Set working directory
#setwd(choose.dir())
bobcat <- read.csv("Bobcat.csv") # detection/non-detection data
p_covs <- read.csv("p covariates.csv") #Detection covarites
psi_covs <- read.csv("psi covariates.csv") #Site level covairates

bobcat_mat <- as.matrix(bobcat)

det_covs <- list(
  People = data.frame(p_covs[,c(1:71)]))
head(det_covs$People)
```

```
##  People1 People2 People3 People4 People5 People6 People7 People8 People9
## 1    0.82    0.25    0.15    0.09    0.07    0.28    1.39    1.14    0.13
## 2    0.00    0.00    0.00    0.00    0.00    0.00    0.00    0.00    0.00
## 3    0.74    0.12    0.18    0.59    0.53    0.41    1.03    0.53    0.15
## 4    0.00    0.00    0.00    0.00    0.00    0.00    0.00    0.00    0.00
## 5    0.00    0.00    0.00    0.00    0.00    0.00    0.00    0.00    0.00
## 6    0.37    0.24    0.06    0.14    0.10    0.22    0.53    0.32    0.06
##  People10 People11 People12 People13 People14 People15 People16 People17
## 1    0.43    0.44    0.63    0.26    1.15    0.43    0.00    0.00
## 2    0.00    0.00    0.00    0.00    0.00    0.00    0.00    0.00
## 3    0.32    0.34    0.19    0.20    0.96    NA      NA      NA
## 4    0.00    0.00    0.00    0.00    0.01    0.01    0.00    0.00
## 5    0.00    0.00    0.00    0.00    0.02    0.00    0.00    0.00
## 6    0.02    0.01    0.06    0.09    0.37    0.55    0.05    0.04
##  People18 People19 People20 People21 People22 People23 People24 People25
## 1    0.00    0.00    0.00    0.0    0.91    NA      NA      NA
## 2    0.00    0.00    0.00    NA     NA     NA     NA     NA
## 3    NA     NA     NA     NA     NA     NA     NA     NA
## 4    0.00    NA     NA     NA     NA     NA     NA     NA
## 5    0.00    0.00    0.00    0.0    NA     NA     NA     NA
## 6    0.06    0.19    0.19    0.2    NA     NA     NA     NA
##  People26 People27 People28 People29 People30 People31 People32 People33
## 1    NA     NA     NA     NA     NA     NA     NA     NA
## 2    NA     NA     NA     NA     NA     NA     NA     NA
## 3    NA     NA     NA     NA     NA     NA     NA     NA
## 4    NA     NA     NA     NA     NA     NA     NA     NA
## 5    NA     NA     NA     NA     NA     NA     NA     NA
```

```
## 6      NA      NA      NA      NA      NA      NA      NA      NA
## People34 People35 People36 People37 People38 People39 People40 People41
## 1      NA      NA      NA      NA      NA      NA      NA      NA
## 2      NA      NA      NA      NA      NA      NA      NA      NA
## 3      NA      NA      NA      NA      NA      NA      NA      NA
## 4      NA      NA      NA      NA      NA      NA      NA      NA
## 5      NA      NA      NA      NA      NA      NA      NA      NA
## 6      NA      NA      NA      NA      NA      NA      NA      NA
## People42 People43 People44 People45 People46 People47 People48 People49
## 1      NA      NA      NA      NA      NA      NA      NA      NA
## 2      NA      NA      NA      NA      NA      NA      NA      NA
## 3      NA      NA      NA      NA      NA      NA      NA      NA
## 4      NA      NA      NA      NA      NA      NA      NA      NA
## 5      NA      NA      NA      NA      NA      NA      NA      NA
## 6      NA      NA      NA      NA      NA      NA      NA      NA
## People50 People51 People52 People53 People54 People55 People56 People57
## 1      NA      NA      NA      NA      NA      NA      NA      NA
## 2      NA      NA      NA      NA      NA      NA      NA      NA
## 3      NA      NA      NA      NA      NA      NA      NA      NA
## 4      NA      NA      NA      NA      NA      NA      NA      NA
## 5      NA      NA      NA      NA      NA      NA      NA      NA
## 6      NA      NA      NA      NA      NA      NA      NA      NA
## People58 People59 People60 People61 People62 People63 People64 People65
## 1      NA      NA      NA      NA      NA      NA      NA      NA
## 2      NA      NA      NA      NA      NA      NA      NA      NA
## 3      NA      NA      NA      NA      NA      NA      NA      NA
## 4      NA      NA      NA      NA      NA      NA      NA      NA
## 5      NA      NA      NA      NA      NA      NA      NA      NA
## 6      NA      NA      NA      NA      NA      NA      NA      NA
## People66 People67 People68 People69 People70 People71
## 1      NA      NA      NA      NA      NA      NA
## 2      NA      NA      NA      NA      NA      NA
## 3      NA      NA      NA      NA      NA      NA
## 4      NA      NA      NA      NA      NA      NA
## 5      NA      NA      NA      NA      NA      NA
## 6      NA      NA      NA      NA      NA      NA
```

```
library(unmarked)
```

```
## Warning: package 'unmarked' was built under R version 4.0.4
```

```
## Loading required package: lattice
```

```
occu_data <- unmarkedFrameOccu(y = bobcat_mat, siteCovs = psi_covs, obsCovs = det_covs)
head(occu_data)
```

```
## Data frame representation of unmarkedFrame object.
```

```
##      y.1 y.2 y.3 y.4 y.5 y.6 y.7 y.8 y.9 y.10 y.11 y.12 y.13 y.14 y.15 y.16 y.17
## 1      0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0
## 2      0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0
## 3      0  0  0  0  0  0  0  0  0  0  0  0  0  0  NA  NA  NA
## 4      0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0
```

## 5	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
## 7	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
## 8	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
## 9	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
## 10	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
##	y.18	y.19	y.20	y.21	y.22	y.23	y.24	y.25	y.26	y.27	y.28	y.29	y.30	y.31	y.32		
## 1	0	0	0	0	0	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 2	0	0	0	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 3	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 4	0	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 5	0	0	0	0	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 6	0	0	0	0	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 7	0	0	0	0	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 8	0	0	0	0	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 9	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	NA
## 10	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	NA
##	y.33	y.34	y.35	y.36	y.37	y.38	y.39	y.40	y.41	y.42	y.43	y.44	y.45	y.46	y.47		
## 1	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 2	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 3	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 4	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 5	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 6	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 7	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 8	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 9	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 10	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
##	y.48	y.49	y.50	y.51	y.52	y.53	y.54	y.55	y.56	y.57	y.58	y.59	y.60	y.61	y.62		
## 1	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 2	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 3	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 4	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 5	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 6	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 7	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 8	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 9	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 10	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
##	y.63	y.64	y.65	y.66	y.67	y.68	y.69	y.70	y.71	Dist_5km	People.1	People.2					
## 1	NA	NA	NA	NA	NA	NA	NA	NA	NA	0.04	0.82	0.25					
## 2	NA	NA	NA	NA	NA	NA	NA	NA	NA	0.04	0.00	0.00					
## 3	NA	NA	NA	NA	NA	NA	NA	NA	NA	0.03	0.74	0.12					
## 4	NA	NA	NA	NA	NA	NA	NA	NA	NA	0.03	0.00	0.00					
## 5	NA	NA	NA	NA	NA	NA	NA	NA	NA	0.03	0.00	0.00					
## 6	NA	NA	NA	NA	NA	NA	NA	NA	NA	0.03	0.37	0.24					
## 7	NA	NA	NA	NA	NA	NA	NA	NA	NA	0.03	0.00	0.00					
## 8	NA	NA	NA	NA	NA	NA	NA	NA	NA	0.03	0.00	0.00					
## 9	NA	NA	NA	NA	NA	NA	NA	NA	NA	0.03	0.24	0.14					
## 10	NA	NA	NA	NA	NA	NA	NA	NA	NA	0.03	0.00	0.00					
##	People.3	People.4	People.5	People.6	People.7	People.8	People.9	People.10									
## 1	0.15	0.09	0.07	0.28	1.39	1.14	0.13	0.43									
## 2	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00									
## 3	0.18	0.59	0.53	0.41	1.03	0.53	0.15	0.32									

## 4	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
## 5	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
## 6	0.06	0.14	0.10	0.22	0.53	0.32	0.06	0.02
## 7	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
## 8	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
## 9	0.07	0.05	0.10	0.12	0.30	0.27	0.27	0.06
## 10	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
##	People.11	People.12	People.13	People.14	People.15	People.16	People.17	
## 1	0.44	0.63	0.26	1.15	0.43	0.00	0.00	
## 2	0.00	0.00	0.00	0.00	0.00	0.00	0.00	
## 3	0.34	0.19	0.20	0.96	NA	NA	NA	
## 4	0.00	0.00	0.00	0.01	0.01	0.00	0.00	
## 5	0.00	0.00	0.00	0.02	0.00	0.00	0.00	
## 6	0.01	0.06	0.09	0.37	0.55	0.05	0.04	
## 7	0.00	0.00	0.00	0.00	0.00	0.00	0.00	
## 8	0.00	0.00	0.00	0.00	0.00	0.00	0.00	
## 9	0.08	0.15	0.04	0.18	0.46	0.09	0.04	
## 10	0.00	0.00	0.00	0.00	0.00	0.00	0.00	
##	People.18	People.19	People.20	People.21	People.22	People.23	People.24	
## 1	0.00	0.00	0.00	0.00	0.91	NA	NA	
## 2	0.00	0.00	0.00	NA	NA	NA	NA	
## 3	NA	NA	NA	NA	NA	NA	NA	
## 4	0.00	NA	NA	NA	NA	NA	NA	
## 5	0.00	0.00	0.00	0.00	NA	NA	NA	
## 6	0.06	0.19	0.19	0.20	NA	NA	NA	
## 7	0.00	0.03	0.00	0.00	NA	NA	NA	
## 8	0.00	0.00	0.00	0.00	NA	NA	NA	
## 9	0.09	0.03	0.03	0.04	0.03	0.18	0.2	
## 10	0.00	0.00	0.00	0.00	0.00	0.00	0.0	
##	People.25	People.26	People.27	People.28	People.29	People.30	People.31	
## 1	NA	NA	NA	NA	NA	NA	NA	
## 2	NA	NA	NA	NA	NA	NA	NA	
## 3	NA	NA	NA	NA	NA	NA	NA	
## 4	NA	NA	NA	NA	NA	NA	NA	
## 5	NA	NA	NA	NA	NA	NA	NA	
## 6	NA	NA	NA	NA	NA	NA	NA	
## 7	NA	NA	NA	NA	NA	NA	NA	
## 8	NA	NA	NA	NA	NA	NA	NA	
## 9	0.14	0.11	0.06	0.18	0.38	0.21	0.09	
## 10	0.00	0.00	0.00	0.00	0.00	0.00	0.00	
##	People.32	People.33	People.34	People.35	People.36	People.37	People.38	
## 1	NA	NA	NA	NA	NA	NA	NA	
## 2	NA	NA	NA	NA	NA	NA	NA	
## 3	NA	NA	NA	NA	NA	NA	NA	
## 4	NA	NA	NA	NA	NA	NA	NA	
## 5	NA	NA	NA	NA	NA	NA	NA	
## 6	NA	NA	NA	NA	NA	NA	NA	
## 7	NA	NA	NA	NA	NA	NA	NA	
## 8	NA	NA	NA	NA	NA	NA	NA	
## 9	NA	NA	NA	NA	NA	NA	NA	
## 10	NA	NA	NA	NA	NA	NA	NA	
##	People.39	People.40	People.41	People.42	People.43	People.44	People.45	
## 1	NA	NA	NA	NA	NA	NA	NA	
## 2	NA	NA	NA	NA	NA	NA	NA	

## 3	NA	NA	NA	NA	NA	NA	NA
## 4	NA	NA	NA	NA	NA	NA	NA
## 5	NA	NA	NA	NA	NA	NA	NA
## 6	NA	NA	NA	NA	NA	NA	NA
## 7	NA	NA	NA	NA	NA	NA	NA
## 8	NA	NA	NA	NA	NA	NA	NA
## 9	NA	NA	NA	NA	NA	NA	NA
## 10	NA	NA	NA	NA	NA	NA	NA
##	People.46	People.47	People.48	People.49	People.50	People.51	People.52
## 1	NA	NA	NA	NA	NA	NA	NA
## 2	NA	NA	NA	NA	NA	NA	NA
## 3	NA	NA	NA	NA	NA	NA	NA
## 4	NA	NA	NA	NA	NA	NA	NA
## 5	NA	NA	NA	NA	NA	NA	NA
## 6	NA	NA	NA	NA	NA	NA	NA
## 7	NA	NA	NA	NA	NA	NA	NA
## 8	NA	NA	NA	NA	NA	NA	NA
## 9	NA	NA	NA	NA	NA	NA	NA
## 10	NA	NA	NA	NA	NA	NA	NA
##	People.53	People.54	People.55	People.56	People.57	People.58	People.59
## 1	NA	NA	NA	NA	NA	NA	NA
## 2	NA	NA	NA	NA	NA	NA	NA
## 3	NA	NA	NA	NA	NA	NA	NA
## 4	NA	NA	NA	NA	NA	NA	NA
## 5	NA	NA	NA	NA	NA	NA	NA
## 6	NA	NA	NA	NA	NA	NA	NA
## 7	NA	NA	NA	NA	NA	NA	NA
## 8	NA	NA	NA	NA	NA	NA	NA
## 9	NA	NA	NA	NA	NA	NA	NA
## 10	NA	NA	NA	NA	NA	NA	NA
##	People.60	People.61	People.62	People.63	People.64	People.65	People.66
## 1	NA	NA	NA	NA	NA	NA	NA
## 2	NA	NA	NA	NA	NA	NA	NA
## 3	NA	NA	NA	NA	NA	NA	NA
## 4	NA	NA	NA	NA	NA	NA	NA
## 5	NA	NA	NA	NA	NA	NA	NA
## 6	NA	NA	NA	NA	NA	NA	NA
## 7	NA	NA	NA	NA	NA	NA	NA
## 8	NA	NA	NA	NA	NA	NA	NA
## 9	NA	NA	NA	NA	NA	NA	NA
## 10	NA	NA	NA	NA	NA	NA	NA
##	People.67	People.68	People.69	People.70	People.71		
## 1	NA	NA	NA	NA	NA		
## 2	NA	NA	NA	NA	NA		
## 3	NA	NA	NA	NA	NA		
## 4	NA	NA	NA	NA	NA		
## 5	NA	NA	NA	NA	NA		
## 6	NA	NA	NA	NA	NA		
## 7	NA	NA	NA	NA	NA		
## 8	NA	NA	NA	NA	NA		
## 9	NA	NA	NA	NA	NA		
## 10	NA	NA	NA	NA	NA		

Question 2

Fit the following candidate set of models:

Detection model	Occupancy model
intercept-only	intercept-only
people	intercept-only
intercept-only	disturbance
people	disturbance

```
fit1 <- occu(~ 1 ~ 1, data = occu_data)
fit2 <- occu(~ People ~ 1, data = occu_data)
fit3 <- occu(~ 1 ~ Dist_5km, data = occu_data)
fit4 <- occu(~ People ~ Dist_5km, data = occu_data)
summary(fit1)
```

```
##
## Call:
## occu(formula = ~1 ~ 1, data = occu_data)
##
## Occupancy (logit-scale):
## Estimate      SE      z P(>|z|)
##      -1.32 0.0879 -15 7.6e-51
##
## Detection (logit-scale):
## Estimate      SE      z P(>|z|)
##      -3.02 0.0734 -41.1      0
##
## AIC: 4490.771
## Number of sites: 1951
## optim convergence code: 0
## optim iterations: 29
## Bootstrap iterations: 0
```

```
summary(fit2)
```

```
##
## Call:
## occu(formula = ~People ~ 1, data = occu_data)
##
## Occupancy (logit-scale):
## Estimate      SE      z P(>|z|)
##      -1.32 0.0879 -15 9.21e-51
##
## Detection (logit-scale):
##           Estimate      SE      z P(>|z|)
## (Intercept) -3.0141 0.0743 -40.550  0.000
## People      -0.0765 0.3759  -0.203  0.839
##
## AIC: 4492.729
## Number of sites: 1951
```

```
## optim convergence code: 0
## optim iterations: 35
## Bootstrap iterations: 0
```

```
summary(fit3)
```

```
##
## Call:
## occu(formula = ~1 ~ Dist_5km, data = occu_data)
##
## Occupancy (logit-scale):
##      Estimate      SE      z P(>|z|)
## (Intercept)  -1.05 0.101 -10.40 2.42e-25
## Dist_5km     -23.65 4.773  -4.96 7.23e-07
##
## Detection (logit-scale):
##      Estimate      SE      z P(>|z|)
##      -3.02 0.0732 -41.2      0
##
## AIC: 4461.006
## Number of sites: 1951
## optim convergence code: 0
## optim iterations: 52
## Bootstrap iterations: 0
```

```
summary(fit4)
```

```
##
## Call:
## occu(formula = ~People ~ Dist_5km, data = occu_data)
##
## Occupancy (logit-scale):
##      Estimate      SE      z P(>|z|)
## (Intercept)  -1.05 0.101 -10.39 2.63e-25
## Dist_5km     -23.64 4.773  -4.95 7.28e-07
##
## Detection (logit-scale):
##      Estimate      SE      z P(>|z|)
## (Intercept) -3.0133 0.0741 -40.646 0.000
## People      -0.0609 0.3779  -0.161 0.872
##
## AIC: 4462.98
## Number of sites: 1951
## optim convergence code: 0
## optim iterations: 39
## Bootstrap iterations: 0
```

Question 3

Perform model selection with AIC. What is your top model? How do you know? Is there model selection uncertainty?

```
library(AICcmodavg)
```

```
## Warning: package 'AICcmodavg' was built under R version 4.0.5
```

```
occu_cand.set <- list(  
  F1 = fit1, F2 = fit2, F3 = fit3, F4 = fit4)  
  
occu_mods <- aictab(cand.set = occu_cand.set, second.ord = F)  
occu_mods
```

```
##  
## Model selection based on AIC:  
##  
##      K      AIC Delta_AIC AICWt Cum.Wt      LL  
## F3 3 4461.01      0.00  0.73  0.73 -2227.50  
## F4 4 4462.98      1.97  0.27  1.00 -2227.49  
## F1 2 4490.77     29.77  0.00  1.00 -2243.39  
## F2 3 4492.73     31.72  0.00  1.00 -2243.36
```

My top model is model 3, fit3, which uses the intercept-only for the detection model and the disturbance within 5km for the abundance model. I know this is my top model because the ΔAIC for Model 1 and Model 2 is greater than 2. There is some uncertainty between model 3 and model because the ΔAIC between Model 3 and Model 4 is less than 2.

Question 4

Average both the effect of people on detection, and disturbance on occupancy, over all models. Report model-averaged slope coefficients and 95% confidence intervals.

```
#People's effect on detection across all models  
people_avg <- modavgShrink(cand.set = occu_cand.set, parm = 'People', second.ord = F, parm.type = 'detection')  
  
People_slopecoef <- people_avg$Mod.avg.beta  
People_slopecoef
```

```
## [1] -0.01653469
```

```
people_95 <- cbind(people_avg$Lower.CL, people_avg$Upper.CL)  
people_95
```

```
##           [,1]      [,2]  
## [1,] -0.4061781 0.3731087
```

```
#Effect of disturbance on occupancy across all models  
dist_avg <- modavgShrink(cand.set = occu_cand.set, parm = 'Dist_5km', second.ord = F, parm.type = 'psi')  
  
Dist_5km_slopecoef <- dist_avg$Mod.avg.beta  
Dist_5km_slopecoef
```

```
## [1] -23.65047
```



```
dist_95 <- cbind(dist_avg$Lower.CL, dist_avg$Upper.CL)
dist_95
```

```
##           [,1]      [,2]
## [1,] -33.006 -14.29494
```

Question 5

Obtain and plot model-averaged predictions of occupancy probability and detection probability. Average over all models, and make predictions over the observed range of each variable.

```
#Detection predictions
p_covs[is.na(p_covs)] = 0 #Necessary because you can't generate new data with NAs
new_p <- data.frame(
  People = seq(min(p_covs[1:71]), to = max(p_covs[1:71]), length.out = 100))

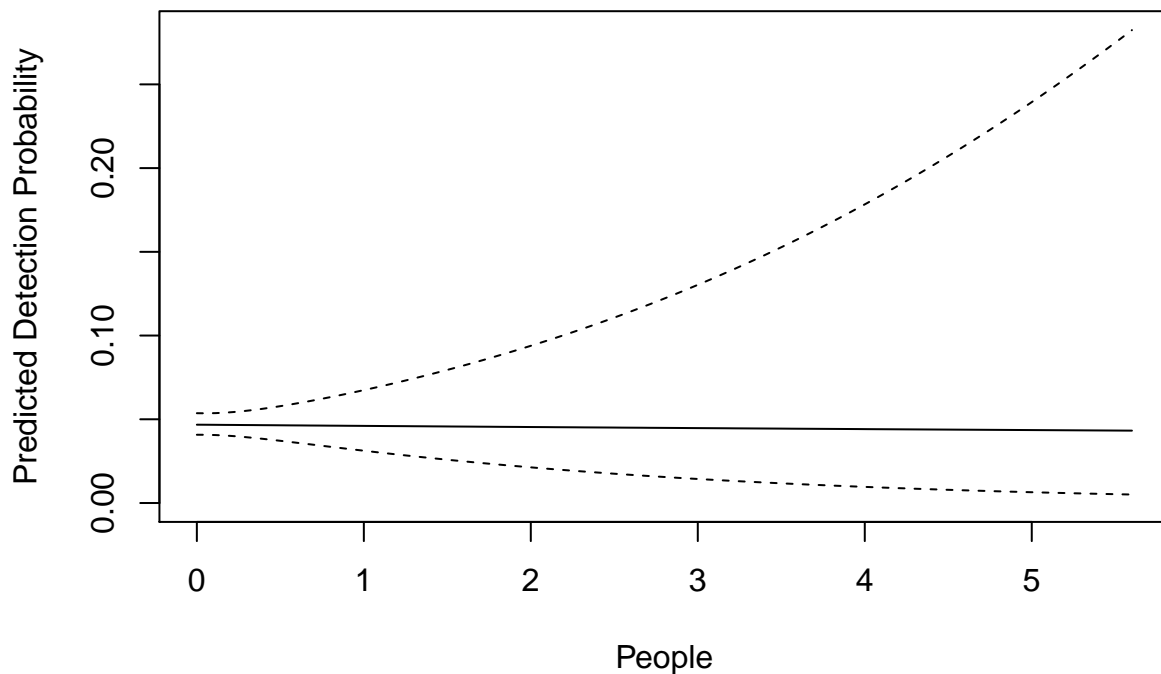
avg_prd_det <- modavgPred(cand.set = occu_cand.set, newdata = new_p, second.ord = F, parm.type = 'detection')
avg_prd_det
```

```
##
## Model-averaged predictions on the response scale
## based on entire model set and 95% confidence interval:
##
##      mod.avg.pred  uncond.se  lower.CL  upper.CL
## 1      0.047      0.003      0.041      0.054
## 2      0.047      0.003      0.041      0.054
## 3      0.047      0.003      0.041      0.054
## 4      0.047      0.003      0.040      0.054
## 5      0.047      0.004      0.040      0.054
## 6      0.047      0.004      0.039      0.055
## 7      0.047      0.004      0.039      0.056
## 8      0.046      0.005      0.038      0.056
## 9      0.046      0.005      0.038      0.057
## 10     0.046      0.005      0.037      0.058
## 11     0.046      0.006      0.036      0.059
## 12     0.046      0.006      0.036      0.060
## 13     0.046      0.006      0.035      0.061
## 14     0.046      0.007      0.034      0.062
## 15     0.046      0.007      0.034      0.063
## 16     0.046      0.008      0.033      0.064
## 17     0.046      0.008      0.032      0.065
## 18     0.046      0.008      0.032      0.067
## 19     0.046      0.009      0.031      0.068
## 20     0.046      0.009      0.030      0.069
## 21     0.046      0.010      0.030      0.070
## 22     0.046      0.010      0.029      0.072
## 23     0.046      0.011      0.028      0.073
## 24     0.046      0.011      0.028      0.074
## 25     0.046      0.011      0.027      0.076
## 26     0.046      0.012      0.027      0.077
## 27     0.046      0.012      0.026      0.079
## 28     0.046      0.013      0.026      0.080
```

## 29	0.046	0.013	0.025	0.082
## 30	0.046	0.013	0.024	0.083
## 31	0.046	0.014	0.024	0.085
## 32	0.046	0.014	0.023	0.086
## 33	0.046	0.015	0.023	0.088
## 34	0.045	0.015	0.022	0.090
## 35	0.045	0.015	0.022	0.092
## 36	0.045	0.016	0.021	0.093
## 37	0.045	0.016	0.021	0.095
## 38	0.045	0.017	0.021	0.097
## 39	0.045	0.017	0.020	0.099
## 40	0.045	0.017	0.020	0.101
## 41	0.045	0.018	0.019	0.102
## 42	0.045	0.018	0.019	0.104
## 43	0.045	0.019	0.018	0.106
## 44	0.045	0.019	0.018	0.108
## 45	0.045	0.019	0.018	0.110
## 46	0.045	0.020	0.017	0.112
## 47	0.045	0.020	0.017	0.114
## 48	0.045	0.020	0.016	0.117
## 49	0.045	0.021	0.016	0.119
## 50	0.045	0.021	0.016	0.121
## 51	0.045	0.021	0.015	0.123
## 52	0.045	0.022	0.015	0.126
## 53	0.045	0.022	0.015	0.128
## 54	0.045	0.023	0.014	0.130
## 55	0.045	0.023	0.014	0.133
## 56	0.045	0.023	0.014	0.135
## 57	0.045	0.024	0.013	0.137
## 58	0.045	0.024	0.013	0.140
## 59	0.045	0.024	0.013	0.143
## 60	0.045	0.025	0.013	0.145
## 61	0.044	0.025	0.012	0.148
## 62	0.044	0.025	0.012	0.150
## 63	0.044	0.026	0.012	0.153
## 64	0.044	0.026	0.011	0.156
## 65	0.044	0.026	0.011	0.159
## 66	0.044	0.027	0.011	0.161
## 67	0.044	0.027	0.011	0.164
## 68	0.044	0.027	0.010	0.167
## 69	0.044	0.028	0.010	0.170
## 70	0.044	0.028	0.010	0.173
## 71	0.044	0.028	0.010	0.176
## 72	0.044	0.028	0.010	0.179
## 73	0.044	0.029	0.009	0.182
## 74	0.044	0.029	0.009	0.185
## 75	0.044	0.029	0.009	0.189
## 76	0.044	0.030	0.009	0.192
## 77	0.044	0.030	0.009	0.195
## 78	0.044	0.030	0.008	0.199
## 79	0.044	0.031	0.008	0.202
## 80	0.044	0.031	0.008	0.205
## 81	0.044	0.031	0.008	0.209
## 82	0.044	0.031	0.008	0.212

```
## 83      0.044      0.032      0.007      0.216
## 84      0.044      0.032      0.007      0.219
## 85      0.044      0.032      0.007      0.223
## 86      0.044      0.033      0.007      0.227
## 87      0.044      0.033      0.007      0.230
## 88      0.044      0.033      0.007      0.234
## 89      0.044      0.033      0.006      0.238
## 90      0.044      0.034      0.006      0.242
## 91      0.043      0.034      0.006      0.246
## 92      0.043      0.034      0.006      0.250
## 93      0.043      0.035      0.006      0.254
## 94      0.043      0.035      0.006      0.258
## 95      0.043      0.035      0.006      0.262
## 96      0.043      0.035      0.006      0.266
## 97      0.043      0.036      0.005      0.270
## 98      0.043      0.036      0.005      0.274
## 99      0.043      0.036      0.005      0.278
## 100     0.043      0.036      0.005      0.282
```

```
plot(x = new_p$People, y = avg_prd_det$mod.avg.pred, type = 'l', ylab = 'Predicted Detection Probability',
lines(x = new_p$People, y = avg_prd_det$upper.CL, lwd = 1, lty = 2)
lines(x = new_p$People, y = avg_prd_det$lower.CL , lty = 2)
```



```
#Occupancy predictions
new_psi <- data.frame(Dist_5km = seq(from = min(psi_covs), to = max(psi_covs), length.out = 100))
new_psi
```

##	Dist_5km
## 1	0.000000000
## 2	0.001313131
## 3	0.002626263
## 4	0.003939394
## 5	0.005252525
## 6	0.006565657
## 7	0.007878788
## 8	0.009191919
## 9	0.010505051
## 10	0.011818182
## 11	0.013131313
## 12	0.014444444
## 13	0.015757576
## 14	0.017070707
## 15	0.018383838
## 16	0.019696970
## 17	0.021010101
## 18	0.022323232
## 19	0.023636364
## 20	0.024949495
## 21	0.026262626
## 22	0.027575758
## 23	0.028888889
## 24	0.030202020
## 25	0.031515152
## 26	0.032828283
## 27	0.034141414
## 28	0.035454545
## 29	0.036767677
## 30	0.038080808
## 31	0.039393939
## 32	0.040707071
## 33	0.042020202
## 34	0.043333333
## 35	0.044646465
## 36	0.045959596
## 37	0.047272727
## 38	0.048585859
## 39	0.049898990
## 40	0.051212121
## 41	0.052525253
## 42	0.053838384
## 43	0.055151515
## 44	0.056464646
## 45	0.057777778
## 46	0.059090909
## 47	0.060404040
## 48	0.061717172
## 49	0.063030303
## 50	0.064343434
## 51	0.065656566
## 52	0.066969697
## 53	0.068282828

```
## 54 0.069595960
## 55 0.070909091
## 56 0.072222222
## 57 0.073535354
## 58 0.074848485
## 59 0.076161616
## 60 0.077474747
## 61 0.078787879
## 62 0.080101010
## 63 0.081414141
## 64 0.082727273
## 65 0.084040404
## 66 0.085353535
## 67 0.086666667
## 68 0.087979798
## 69 0.089292929
## 70 0.090606061
## 71 0.091919192
## 72 0.093232323
## 73 0.094545455
## 74 0.095858586
## 75 0.097171717
## 76 0.098484848
## 77 0.099797980
## 78 0.101111111
## 79 0.102424242
## 80 0.103737374
## 81 0.105050505
## 82 0.106363636
## 83 0.107676768
## 84 0.108989899
## 85 0.110303030
## 86 0.111616162
## 87 0.112929293
## 88 0.114242424
## 89 0.115555556
## 90 0.116868687
## 91 0.118181818
## 92 0.119494949
## 93 0.120808081
## 94 0.122121212
## 95 0.123434343
## 96 0.124747475
## 97 0.126060606
## 98 0.127373737
## 99 0.128686869
## 100 0.130000000
```

```
avg_prd_occupancy <- modavgPred(cand.set = occu_cand.set, newdata = new_psi, second.ord = F, parm.type = "avg")
avg_prd_occupancy
```

```
##
## Model-averaged predictions on the response scale
## based on entire model set and 95% confidence interval:
```

##	mod.avg.pred	uncond.se	lower.CL	upper.CL
## 1	0.260	0.019	0.224	0.300
## 2	0.254	0.019	0.219	0.292
## 3	0.248	0.018	0.215	0.285
## 4	0.242	0.017	0.210	0.278
## 5	0.237	0.017	0.206	0.271
## 6	0.231	0.016	0.201	0.264
## 7	0.226	0.016	0.196	0.258
## 8	0.220	0.015	0.192	0.252
## 9	0.215	0.015	0.187	0.246
## 10	0.210	0.015	0.182	0.241
## 11	0.205	0.015	0.177	0.235
## 12	0.200	0.015	0.172	0.230
## 13	0.195	0.015	0.168	0.225
## 14	0.190	0.015	0.163	0.221
## 15	0.185	0.015	0.158	0.216
## 16	0.181	0.015	0.153	0.212
## 17	0.176	0.015	0.148	0.208
## 18	0.172	0.015	0.144	0.204
## 19	0.167	0.016	0.139	0.200
## 20	0.163	0.016	0.134	0.196
## 21	0.159	0.016	0.130	0.193
## 22	0.155	0.016	0.126	0.189
## 23	0.151	0.016	0.121	0.186
## 24	0.147	0.017	0.117	0.182
## 25	0.143	0.017	0.113	0.179
## 26	0.139	0.017	0.109	0.176
## 27	0.135	0.017	0.105	0.173
## 28	0.132	0.017	0.101	0.170
## 29	0.128	0.018	0.098	0.167
## 30	0.125	0.018	0.094	0.164
## 31	0.122	0.018	0.091	0.161
## 32	0.118	0.018	0.087	0.158
## 33	0.115	0.018	0.084	0.156
## 34	0.112	0.018	0.081	0.153
## 35	0.109	0.018	0.078	0.150
## 36	0.106	0.018	0.075	0.148
## 37	0.103	0.019	0.072	0.145
## 38	0.100	0.019	0.069	0.143
## 39	0.097	0.019	0.067	0.140
## 40	0.095	0.019	0.064	0.138
## 41	0.092	0.019	0.061	0.136
## 42	0.090	0.019	0.059	0.133
## 43	0.087	0.019	0.057	0.131
## 44	0.085	0.019	0.055	0.129
## 45	0.082	0.019	0.052	0.127
## 46	0.080	0.019	0.050	0.125
## 47	0.078	0.018	0.048	0.122
## 48	0.075	0.018	0.046	0.120
## 49	0.073	0.018	0.045	0.118
## 50	0.071	0.018	0.043	0.116
## 51	0.069	0.018	0.041	0.114
## 52	0.067	0.018	0.039	0.112

## 53	0.065	0.018	0.038	0.110
## 54	0.063	0.018	0.036	0.109
## 55	0.062	0.018	0.035	0.107
## 56	0.060	0.018	0.033	0.105
## 57	0.058	0.017	0.032	0.103
## 58	0.056	0.017	0.031	0.101
## 59	0.055	0.017	0.030	0.100
## 60	0.053	0.017	0.028	0.098
## 61	0.052	0.017	0.027	0.096
## 62	0.050	0.017	0.026	0.095
## 63	0.049	0.016	0.025	0.093
## 64	0.047	0.016	0.024	0.091
## 65	0.046	0.016	0.023	0.090
## 66	0.045	0.016	0.022	0.088
## 67	0.043	0.016	0.021	0.087
## 68	0.042	0.015	0.020	0.085
## 69	0.041	0.015	0.019	0.084
## 70	0.040	0.015	0.019	0.082
## 71	0.038	0.015	0.018	0.081
## 72	0.037	0.015	0.017	0.079
## 73	0.036	0.014	0.016	0.078
## 74	0.035	0.014	0.016	0.077
## 75	0.034	0.014	0.015	0.075
## 76	0.033	0.014	0.014	0.074
## 77	0.032	0.014	0.014	0.073
## 78	0.031	0.013	0.013	0.071
## 79	0.030	0.013	0.013	0.070
## 80	0.029	0.013	0.012	0.069
## 81	0.028	0.013	0.012	0.068
## 82	0.028	0.013	0.011	0.067
## 83	0.027	0.012	0.011	0.065
## 84	0.026	0.012	0.010	0.064
## 85	0.025	0.012	0.010	0.063
## 86	0.024	0.012	0.009	0.062
## 87	0.024	0.012	0.009	0.061
## 88	0.023	0.011	0.009	0.060
## 89	0.022	0.011	0.008	0.059
## 90	0.022	0.011	0.008	0.058
## 91	0.021	0.011	0.008	0.057
## 92	0.020	0.011	0.007	0.056
## 93	0.020	0.010	0.007	0.055
## 94	0.019	0.010	0.007	0.054
## 95	0.019	0.010	0.006	0.053
## 96	0.018	0.010	0.006	0.052
## 97	0.018	0.010	0.006	0.051
## 98	0.017	0.009	0.006	0.050
## 99	0.016	0.009	0.005	0.049
## 100	0.016	0.009	0.005	0.048

```

plot(x = new_psi$Dist_5km, y = avg_prd_occupancy$mod.avg.pred, type = 'l', ylab = 'Predicted Occupancy')
lines(x = new_psi$Dist_5km, y = avg_prd_occupancy$upper.CL, lwd = 1, lty = 2)
lines(x = new_psi$Dist_5km, y = avg_prd_occupancy$lower.CL, lty = 2)

```

