

## **Electronic Supplementary Information (ESM)**

Estimating the abundance of rare and elusive carnivores from photographic-sampling data when the population size is very small

Brian D. Gerber<sup>1</sup>, Jacob S. Ivan<sup>2</sup>, and Kenneth P. Burnham<sup>3</sup>

<sup>1</sup>Colorado Cooperative Fish and Wildlife Research Unit, Department of Fish, Wildlife and Conservation Biology, Colorado State University, Fort Collins, CO 80523, USA.

<sup>2</sup>Colorado Division of Parks and Wildlife, Wildlife Research Center, 317 West Prospect Road, Fort Collins, Colorado 80526-2097, USA

<sup>3</sup>Professor Emeritus, Department of Fish, Wildlife, and Conservation Biology, Colorado State University, Fort Collins, CO 80523, USA.

Corresponding Author:

Brian D. Gerber, Colorado Cooperative Fish and Wildlife Research Unit Department of Fish, Wildlife and Conservation Biology, Colorado State University, Fort Collins, CO 80523, USA.

bgerber@colostate.edu

## R code for simulations

```
#####  
#Simulate heterogeneous capture-recapture  
#data to examine how many individuals will  
#likely be detected and secondly, to  
#estimate abundance using the M0-unconditional  
#likelihood C-R estimator and quantify uncertainty  
#with profile-likelihood confidence intervals  
#####  
#Author: Brian D. Gerber  
#Contact: bgerber@colostate.edu  
#Last Modified: 8/22/2012  
#####  
  
#####  
#Load Functions and Libraries  
#####  
library(RMark)  
library(gregmisc)  
  
  expit = function (logit){  
    exp(logit)/(1+exp(logit))  
  }  
  
  logit = function (expit){  
    log(expit/(1-expit))  
  }  
  
#####  
#Setup Variables  
#####  
J=30 #J will be the number of sampling occasions, identified as t in manuscript  
N=10 #True abundance- animals that have some chance of capture  
sets=100 #Number of datasets to simulate  
l=0  
num_capt=rep(0,sets) #number of animals detected- M(t+1)  
M0_N=M0_se=M0_lcl=M0_ucl=rep(0,sets) # Store abundance estimates  
#Set the heterogeneity for detection probability- logit(p) as  
mu=0.1  
sigma=0.1  
  
#####  
#Plot heterogeneity of p
```

```
#####

plot(1, type="n", axes=T, xlab="", ylab="", xlim=c(0,1), ylim=c(0,4))
curve(dnorm(x,mu,sigma),col=2,lwd=2,add=TRUE)

#####
#Create heterogeneity
#C-R Data and start simulation
#####

while (l<sets){
  if(1%%10==0)cat("Simulation # ",l,"\n"); flush.console()
  mat=data.frame(matrix(0,N,J))
  for (q in 1:N){
    detection=expit(rnorm(1,logit(mu),sigma))
    for (r in 1:J){
      sample_unif=runif(1,0,1)
      if(sample_unif<=detection){
        mat[q,r]=1 }
      else{mat[q,r]=0}
    }
  }
}

#Drop the capture histories with no detections
Rcapture_input=as.matrix(mat[rowSums(mat) != 0, , drop=FALSE])

#Create input for RMARK
mark_input=mat[rowSums(mat) != 0, , drop=FALSE]
mark_input=as.data.frame(apply(mark_input,1,paste,collapse=""))
mark_input <- data.frame(lapply(mark_input, as.character), stringsAsFactors=FALSE)
colnames(mark_input)= c("ch")
num_capt_temp=dim(Rcapture_input)[1]

#With low p and N, sum C-H will have zero animals
#or some C-H will be too small, such as 2 animals, which
#can't be used to estimate N

if(num_capt_temp<=2){
  next}
else{
  l=l+1
  num_capt[l]=dim(Rcapture_input)[1] #this is the number of animals caught
}
```

```

#Define Null model
pdotshared=list(formula=~1,share=TRUE)

#Run model through RMARK- M0-unconditional with profile likelihood CI.
m0=try(mark(mark_input,model="Closed",profile.int =
TRUE,adjust=FALSE,model.parameters=list(p=pdotshared), brief=FALSE,output=FALSE,
delete=TRUE, invisible=TRUE), silent=TRUE)

#Save estimates in simulation
M0_N[I]=as.numeric(m0$results$real[2,1]) #Abundance only
M0_se[I]=as.numeric(m0$results$real[2,2]) #Abundance SE
M0_lcl[I]=as.numeric(m0$results$real[2,3]) #LCL
M0_ucl[I]=as.numeric(m0$results$real[2,4]) #UCL
#   if(!is.na(M0_se[I])){if(M0_se[I]>N*inflation){M0_lcl[I]=M0_ucl[I]=NA
#
#               M0_se[I]= NA}}

}
}
#####
M0_N=as.numeric(M0_N)
M0_se=as.numeric(M0_se)
M0_lcl=as.numeric(M0_lcl)
M0_ucl=as.numeric(M0_ucl)
M0_se[M0_se=="NaN"]=NA
M0=cbind(M0_N,M0_se,M0_lcl,M0_ucl)

#Display frequency of captures by individual
rowSums(Rcapture_input)
mean(rowSums(Rcapture_input)) #Mean frequency of detection

# Plot the histogram of individuals detected
hist(num_capt, xlim=c(0,(N+10)), freq=FALSE)
abline(v=N, lwd=2, col=2) #Plots truth

# Plot proportion of population detected
hist((num_capt/N), xlim=c(0,1), freq=FALSE)

#Plot histogram of abundance estimates
hist(M0_N, breaks=10, main="Abundance Estimate", xlim=c(0,(N+10)),freq=FALSE)
abline(v=mean(M0_N), col=1, lwd=2) #Expected value of N-hat
abline(v=N, col=2, lwd=2) #N-truth

#Expected_bias
mean(M0_N)-N

```

```
#Plot all simulations- abundance estimates and profile condidence intervals
win.graph()
plotCI(seq(1,sets,1), M0_N, pch=16,ui=M0_ucl, li=M0_lcl,xlim=c(1,sets), ylim=c(0,50), gap=0,
xlab="Simulation #",ylab=expression(hat(N)))
abline(h=N, lwd=2, col=2)
```

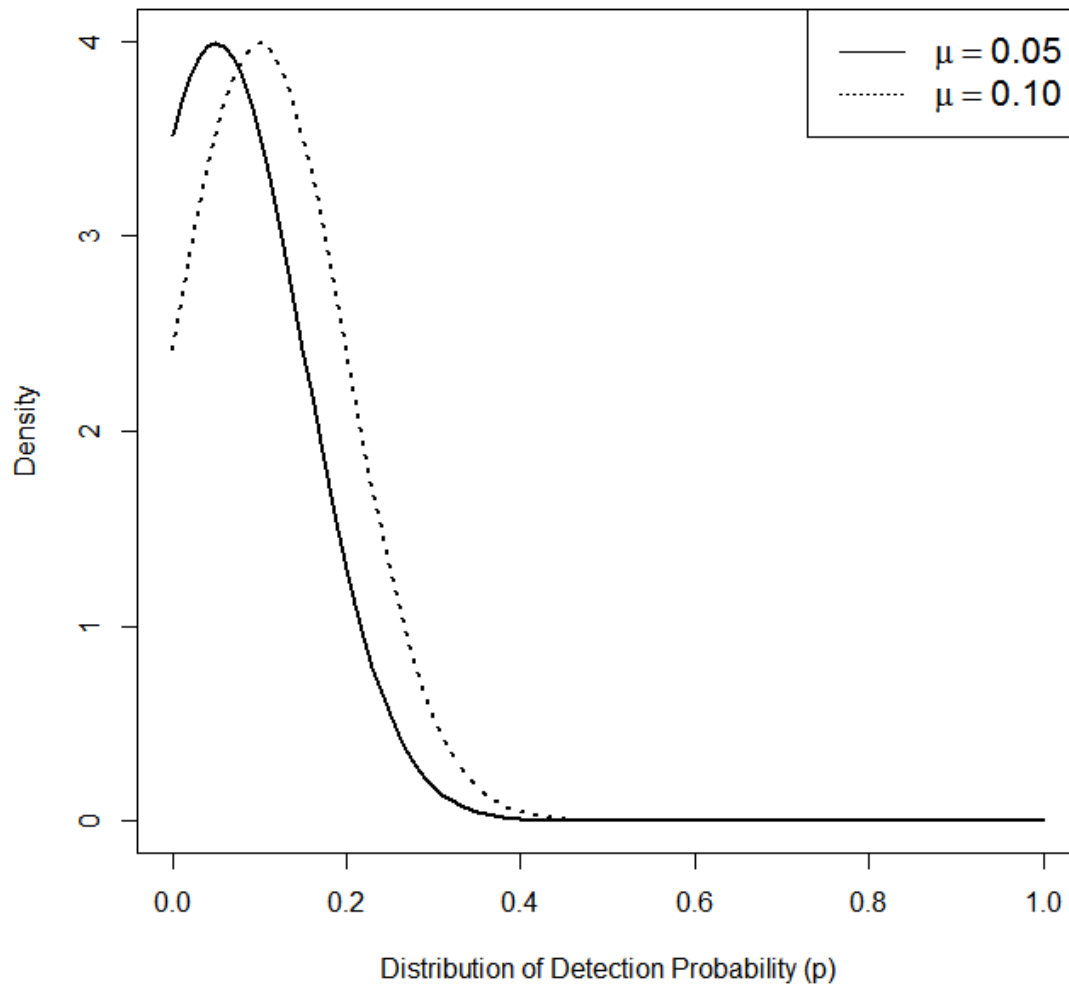
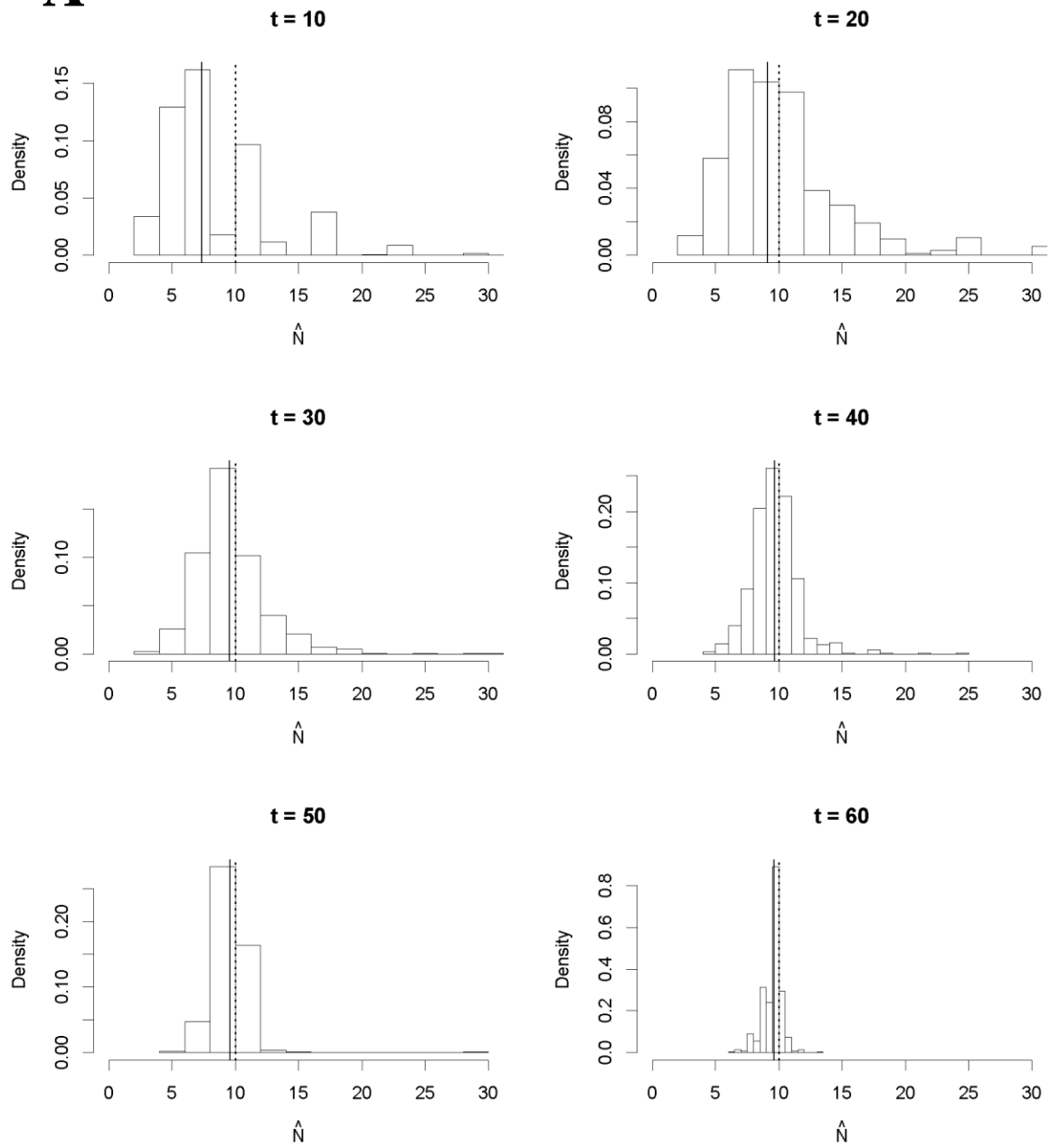


Fig. S1 Distributions of detection probability used to model individual heterogeneity, where  $\text{logit}(p_i) = N(\text{expit}(\mu), \sigma)$  and  $\sigma = 0.1$ .

**A**

# B

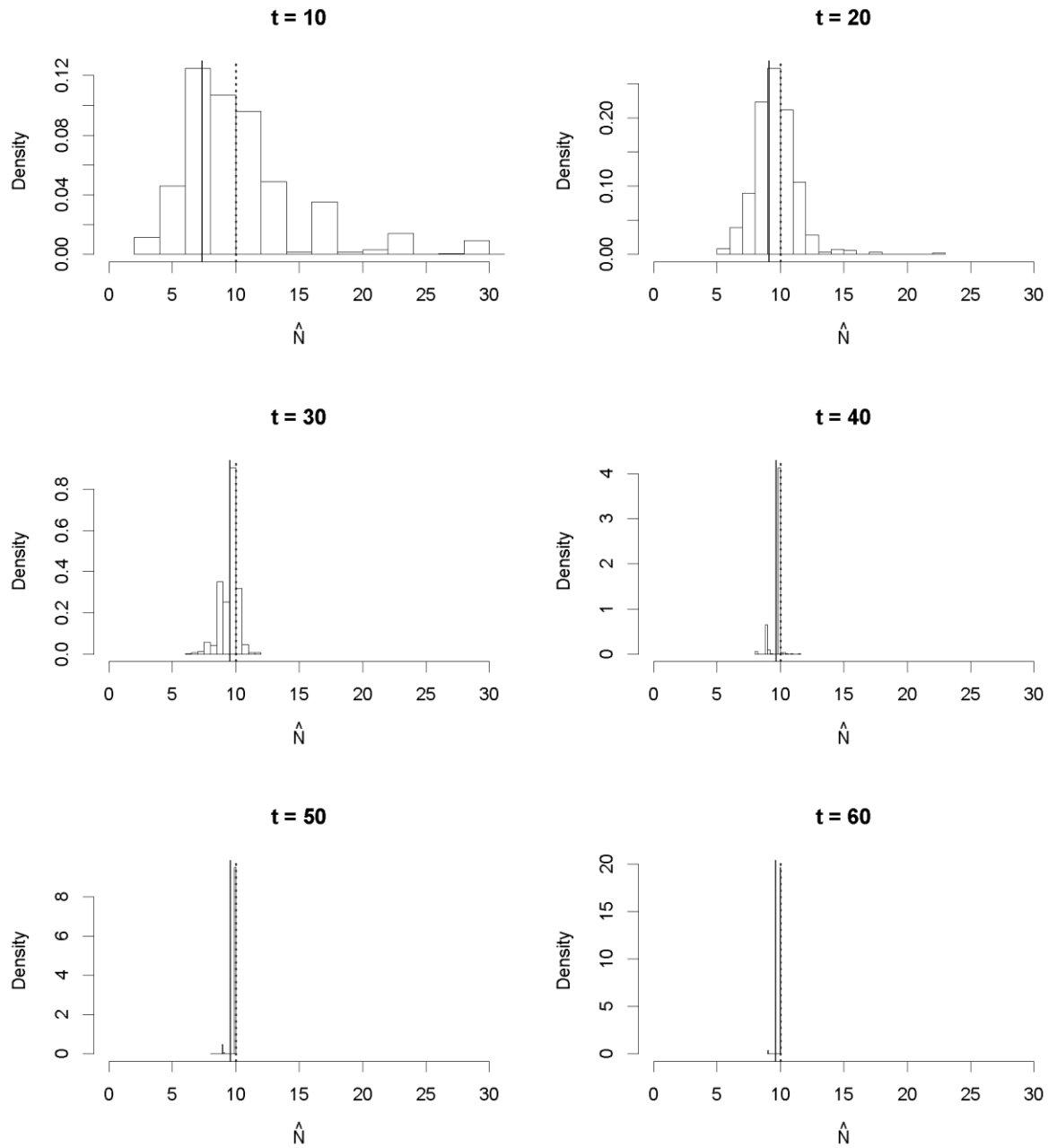


Fig. S2 Empirical distributions of abundance estimates ( $\hat{N}$ ) using the  $M_0$  (null, constant  $p$ ) closed-capture recapture estimator and simulated heterogeneous data where A)  $\text{logit}(p_i) = N(\text{expit}(\mu = 0.05), \sigma = 0.1)$  and B)  $\text{logit}(p_i) = N(\text{expit}(\mu = 0.10), \sigma = 0.1)$  over sampling occasions ( $t$ ) of 10-60; the dotted line at  $N = 10$  indicates true abundance and the solid line is  $E[\hat{N}]$ .



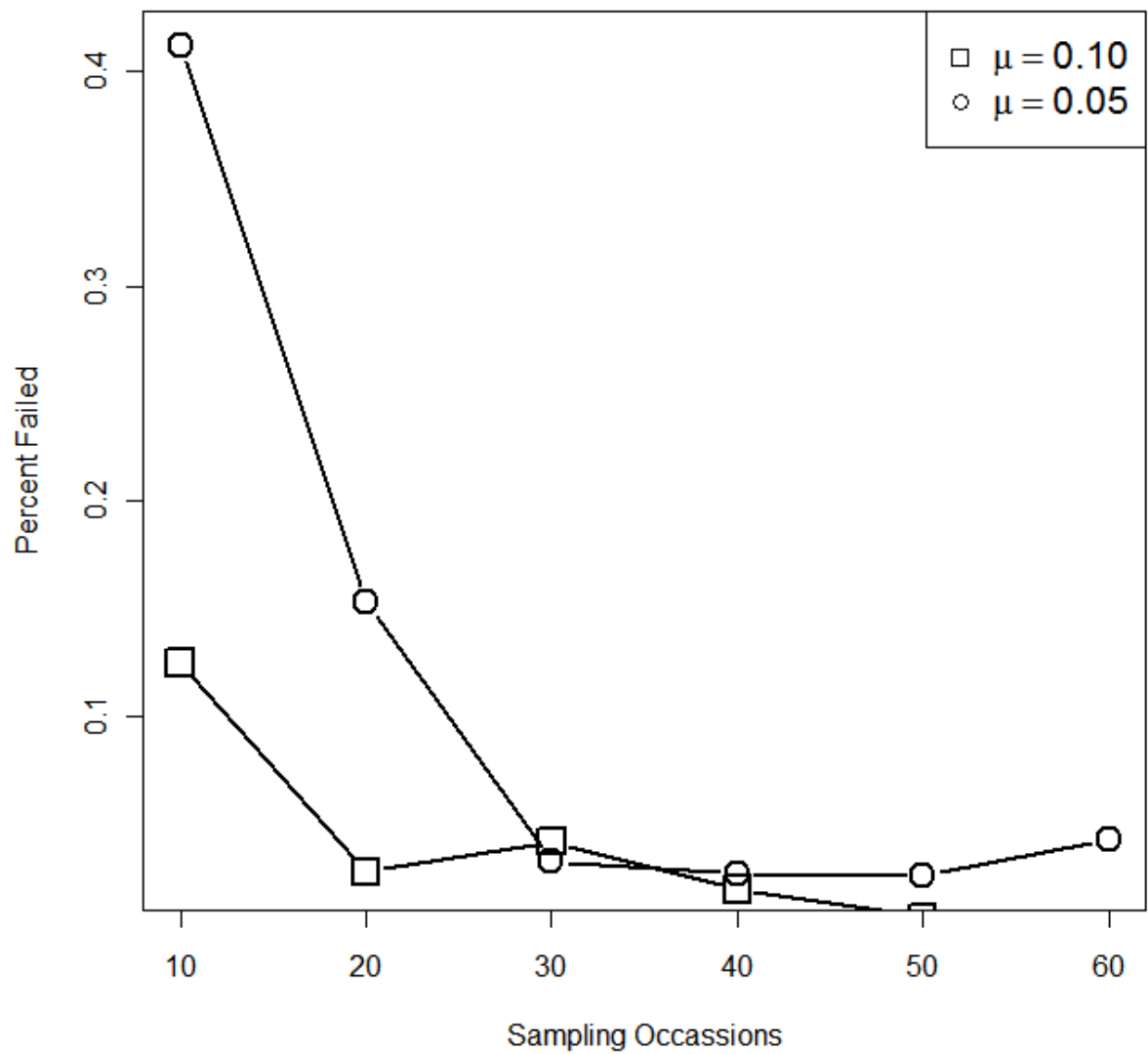


Fig. S3 Percentage of failed point or standard error estimates from simulated heterogeneous capture-recapture data where  $\text{logit}(p_i) = N(\text{expit}(\mu), \sigma)$  where  $\mu = 0.05$  or  $0.1$  and  $\sigma = 0.1$  over sampling occasions of 10-60.

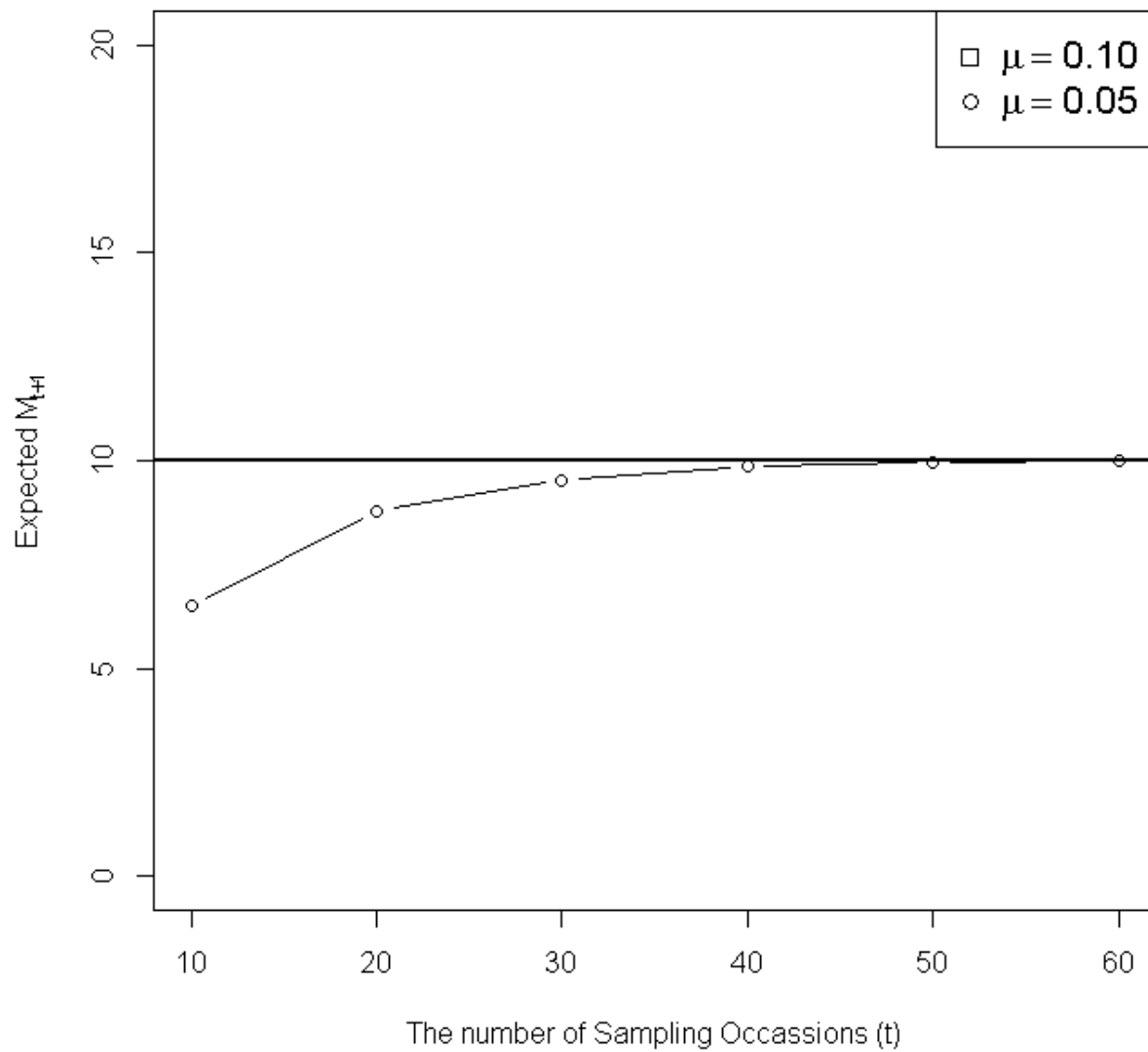


Fig. S4 The expected number of unique individuals detected ( $M_{t+1}$ ) from simulated heterogeneous capture-recapture data where  $\logit(p_i) = N(\text{expit}(\mu), \sigma)$  where  $\mu = 0.05$  or  $0.1$  and  $\sigma = 0.1$  over sampling occasions of 10-60.