Python code of the simulations

1 Source for: Island model – monogenic selection

```
#! /usr/bin/python
#Object: Simulation of an island model model (monogenic version)
#Aim: Simulate data in order to assess BayeScEnv performances
#Author: Pierre de Villemereuil
#Decembre 2013
#-----Header------
print("Importing SimuPOP")
#Setting allele size
from simuOpt import setOptions
setOptions(alleleType='binary') #Binary for two alleles states (SNPs)
\#Importing\ simuPOP
import simuPOP as sim
from simuPOP.utils import saveCSV
from simuPOP.sampling import drawRandomSample
from simuPOP.utils import migrIslandRates
print("Importing some other libraries")
from decimal import *
from math import *
import numpy
import random
print("Script starting")
#-----Simulation parameters-----
#Population(s) size(s)
popsize=500
#Number of generations
numgen = 400
atgen=100
#Number of chromosomes
numchrom=10
#Number of loci per chrom.
numloc = 500
#Create vector of loci numbers
vecloc=[numloc]*numchrom
#Initializing ancestral population frequencies
p=[0]*numloc*numchrom
for i in range(numloc*numchrom):
       p[i]=0.1+(0.9-0.1)*numpy.random.beta(2,2)
                                                   #Almost uniform, but not too much
           weight in the edges plus restricted to (0.1,0.9)
#Migration rate (probability for one individual to disperse)
m = 0.0045
\#Setting\ env\ values
vec_env=[0]*16
j=float(-1.5)
for i in range(16):
       vec_env[i]=round(j,1)
       j = j + 0.2
#-----Custom functions-----
#Function naming the alleles
#Depends on the number of chromosomes (chrom) and loci (loc)
def allele_naming(chrom,loc):
       res=[]
       for i in range(chrom):
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for j in range(1,loc+1):
                       res.append(chr(65+i)+str(j))
       return res
#Function to attribute env values to individuals' info fields
def env_set(pop):
       sim.stat(pop,popSize=True)
       subsize=pop.dvars().subPopSize
       numpop=len(subsize)
       for i in range(numpop):
               pop.setIndInfo(vec_env[i],'env',subPop=i)
       return True
#Link function between env value and selection
#Modified 'logitistic' function
def fit_func(x):
       res = (1 - exp(-x))/(1 + exp(-x))
       res=res/10
       return res
#Defining fitness according to selection
\#(0,0) is the reference genotype
def fit_env(geno,env):
       s=fit_func(env)
       if geno[0]+geno[1]==0:
               w = 1
       if geno[0]+geno[1]==1:
               w = 1 + (s/2)
       if geno[0]+geno[1]==2:
               w = 1 + s
       return w
#Function defining a constant subPop size (popsize) for any number of subpops (demographic
   model)
def demo(pop):
       sim.stat(pop,popSize=True)
       subsize=pop.dvars().subPopSize
#If subsize is of length 1, then it is a integer and len() does not work
       if type(subsize) == type(1):
               numpop=1
       else:
               numpop=len(subsize)
       vecsize=[popsize]*(numpop)
       return vecsize
#-----Setting up the simulator-----
for k in range(1,101):
   print('Simulation number '+str(k))
   \hbox{\it\#Creating initial population of size popsize and "caryotype" vecloc}
   pop=sim.Population(size=[popsize]*16,loci=vecloc,infoFields=['migrate_to','fitness','
       env'],lociNames=allele_naming(numchrom,numloc))
   sim.initSex(pop)
   for i in range(numloc*numchrom):
           for j in range(16):
                   tmp_p=numpy.random.beta(p[i]/0.11,(1-p[i])/0.11)
                   sim.initGenotype(pop,freq=[tmp_p,1-tmp_p],loci=i,subPops=j)
   #-----Main evolving process-----
   pop.evolve(
           sim.PyEval('"Gen: %d" % gen', step=10),
           sim.PyOutput('\n',step=10),
   {\tt\#Migration\ using\ a\ simple\ stepping\ stone\ model}
           sim.Migrator(rate=migrIslandRates(m,16)),
   #Selection process
    #Set environmental value (env infoField) for each individual in the population
   #Takes place at each generation after the first fission
```

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sim.PyOperator(env_set),
#Selection occures at locus 50 according to env information field
         sim.PySelector(fit_env,loci=50,begin=atgen),
#Mating at random (pangamy)
        matingScheme=sim.RandomMating(
#Fixed population size (fixed at 'popsize')
        subPopSize=demo,
#Recombination to avoid selective sweep
         ops=[sim.Recombinator(rates=0.002)]
        ),
        postOps=[
#Mutation rate 10e-6
        sim.SNPMutator(u=0.000001, v=0.000001)
        ],
#Evolve for a number 'numgen' of generations
        gen = numgen
sim.stat(pop,popSize=True)
subsize=pop.dvars().subPopSize
numpop=len(subsize)
for i in range(numpop):
pop.setIndInfo(vec_env[i],'env',subPop=i)
sample = drawRandomSample(pop, sizes=[20]*numpop)
sample.addInfoFields('pop_name')
vecname=[]
for i in range(1,numpop+1):
         vecname = vecname + [i] * 20
sample.setIndInfo(vecname, 'pop_name')
#sim.dump(pop,structure=False)
print('Saving population')
saveCSV(sample,filename="sims/sim"+str(k)+".csv",infoFields=['pop_name','env'],
    sexFormatter=None, affectionFormatter=None, header=False)
```

2 Source for: Island model – polygenic selection

```
#! /usr/bin/python
#Object: Simulation of an island model model (polygenic version)
#Aim: Simulate data in order to assess BayeScEnv performances
#Author: Pierre de Villemereuil
#Decembre 2013
#-----Header-----
print("Importing SimuPOP")
#Setting allele size
from simuOpt import setOptions
setOptions(alleleType='binary') #Binary for two alleles states (SNPs)
#Importing simuPOP
import simuPOP as sim
from simuPOP.utils import saveCSV
{\tt from \ simuPOP.sampling \ import \ drawRandomSample}
from simuPOP.utils import migrIslandRates
print("Importing some other libraries")
from decimal import *
from math import *
import numpy
import random
print("Script starting")
#Population(s) size(s)
popsize=500
```

```
#Number of generations
numgen = 400
atgen=100
#Number of chromosomes
numchrom=10
#Number of loci per chrom.
numloc = 500
#Create vector of loci numbers
vecloc = [numloc] * numchrom
#Initializing ancestral population frequencies
p=[0]*numloc*numchrom
for i in range(numloc*numchrom):
       p[i]=0.1+(0.9-0.1)*numpy.random.beta(2,2)
                                                       #Almost uniform, but not too much
            weight in the edges plus restricted to (0.1,0.9)
#Migration rate (probability for one individual to disperse)
m = 0.0045
#Setting env values
vec_env=[0]*16
j=float(-1.5)
for i in range(16):
       vec_env[i]=round(j,1)
        j = j + 0.2
#Defining the selected loci
locisel
    =[2793,1850,583,4083,3349,860,4785,706,947,939,1819,925,403,2867,2897,97,3102,2618,708,1190,2471,15
#-----Custom functions-----
#Function naming the alleles
\# Depends on the number of chromosomes (chrom) and loci (loc)
def allele_naming(chrom,loc):
       res=[]
        for i in range(chrom):
               for j in range(1,loc+1):
                       res.append(chr(65+i)+str(j))
        return res
#Function to attribute env values to individuals' info fields
def env_set(pop):
        sim.stat(pop,popSize=True)
       subsize=pop.dvars().subPopSize
        numpop=len(subsize)
        for i in range(numpop):
               pop.setIndInfo(vec_env[i], 'env', subPop=i)
        return True
#Link function between env value and selection
#Modified 'logitistic' function
def fit_func(x):
        res=(1-exp(-x))/(1+exp(-x))
        res=res/50
        return res
#Defining fitness according to selection
\#(0,0) is the reference genotype
def fit_env(geno,env):
       N=len(geno)
        s=fit_func(env)
        t11=0
       t00=0
        for i in range (N/2):
                a1=geno[i*2]
                a2=geno[i*2+1]
                if (a1+a2==0):
                       t00=t00+1
                if (a1+a2==2):
```

```
t11=t11+1
       w = ((1+s)**t11)*((1-s)**t00)
       return w
#Function defining a constant subPop size (popsize) for any number of subpops (demographic
   model)
def demo(pop):
       sim.stat(pop,popSize=True)
       subsize=pop.dvars().subPopSize
#If subsize is of length 1, then it is a integer and len() does not work
       if type(subsize) == type(1):
              numpop=1
       else:
              numpop=len(subsize)
       vecsize=[popsize]*(numpop)
       return vecsize
#-----Setting up the simulator-----
for k in range(1,101):
   print('Simulation number '+str(k))
   \hbox{\it\#Creating initial population of size popsize and "caryotype" vecloc}
   pop=sim.Population(size=[popsize]*16,loci=vecloc,infoFields=['migrate_to','fitness','
       env'],lociNames=allele_naming(numchrom, numloc))
   sim.initSex(pop)
   for i in range(numloc*numchrom):
           for j in range(16):
                   tmp_p=numpy.random.beta(p[i]/0.11,(1-p[i])/0.11)
                  sim.initGenotype(pop,freq=[tmp_p,1-tmp_p],loci=i,subPops=j)
   #-----Main evolving process-----
   pop.evolve(
           preOps=[
           sim.PyEval('"Gen: %d" % gen', step=10),
          sim.PyOutput('\n', step=10),
   #Migration using a simple stepping stone model
          sim.Migrator(rate=migrIslandRates(m,16)),
   #Selection process
   #Set environmental value (env infoField) for each individual in the population
   \#Takes place at each generation after the first fission
           sim.PyOperator(env_set),
   \#Selection occures at locus 50 according to env information field
           sim.PySelector(fit_env,loci=locisel,begin=atgen),
   #Mating at random (pangamy)
           matingScheme=sim.RandomMating(
   #Fixed population size (fixed at 'popsize')
           subPopSize=demo,
   #Recombination to avoid selective sweep
           ops=[sim.Recombinator(rates=0.002)]
           ),
           postOps=[
   #Mutation rate 10e-6
           sim.SNPMutator(u=0.000001,v=0.000001),
           ],
    #-----
   #Evolve for a number 'numgen' of generations
           gen = numgen
   sim.stat(pop,popSize=True)
   subsize=pop.dvars().subPopSize
   numpop=len(subsize)
   for i in range(numpop):
          pop.setIndInfo(vec_env[i], 'env', subPop=i)
   sample = drawRandomSample(pop, sizes=[20]*numpop)
   sample.addInfoFields('pop_name')
    vecname=[]
   for i in range(1,numpop+1):
```

3 Source for: Stepping Stone model – polygenic selection

```
#! /usr/bin/python
 #Object: Simulation of stepping stone model (polygenic version)
 #Aim: Simulate data in order to assess BayeScEnv performances
 #Author: Pierre de Villemereuil
 #Decembre 2013
 #-----Header-----Header
 print("Importing SimuPOP")
 #Setting allele size
 from simuOpt import setOptions
 setOptions(alleleType='binary') #Binary for two alleles states (SNPs)
 #Importing simuPOP
 import simuPOP as sim
 from simuPOP.utils import saveCSV
 {\tt from \ simuPOP.sampling \ import \ drawRandomSample}
 from simuPOP.utils import migrIslandRates
 print("Importing some other libraries")
 from decimal import *
 from math import *
 import numpy
 import random
 print("Script starting")
 #Population(s) size(s)
 popsize=500
 #Number of generations
 numgen = 1000
 atgen=700
 #Number of chromosomes
 numchrom=10
 #Number of loci per chrom.
 numloc = 500
 #Create vector of loci numbers
 vecloc=[numloc]*numchrom
 #Initializing ancestral population frequencies
 p=[0]*numloc*numchrom
 for i in range(numloc*numchrom):
                   p[i]=0.1+(0.9-0.1)*numpy.random.beta(2,2)
                                                                                                                           #Almost uniform, but not too much
                            weight in the edges plus restricted to (0.1,0.9)
 #Migration rate (probability for one individual to disperse)
 m = 0.01
 #Defining environment
 vec_env
          = [-0.5198065, -0.8112612, -0.2124103, 0.3941398, 1.1103401, 1.5185518, 0.6871753, 0.0878634, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.0821056, -0.082
 #Defining the selected loci
          =[2793,1850,583,4083,3349,860,4785,706,947,939,1819,925,403,2867,2897,97,3102,2618,708,1190,2471,15
```

```
-----Custom functions-----
#Function naming the alleles
#Depends on the number of chromosomes (chrom) and loci (loc)
def allele_naming(chrom,loc):
       res=[]
       for i in range(chrom):
               for j in range(1,loc+1):
                       res.append(chr(65+i)+str(j))
       return res
#Function to attribute env values to individuals' info fields
def env_set(pop):
       sim.stat(pop,popSize=True)
       subsize=pop.dvars().subPopSize
       numpop=len(subsize)
       for i in range(numpop):
               pop.setIndInfo(vec_env[i], 'env', subPop=i)
#Link function between env value and selection
#Modified 'logitistic' function
def fit func(x):
       res = (1 - exp(-x))/(1 + exp(-x))
       res=res/50
       return res
#Defining fitness according to selection
\#(0,0) is the reference genotype
def fit_env(geno,env):
       N=len(geno)
       s=fit_func(env)
       t11=0
       +.00 = 0
       for i in range(N/2):
               a1=geno[i*2]
               a2=geno[i*2+1]
               if (a1+a2==0):
                       t00=t00+1
               if (a1+a2==2):
                       t11=t11+1
       w = ((1+s)**t11)*((1-s)**t00)
#Function defining a constant subPop size (popsize) for any number of subpops (demographic
   model)
def demo(pop):
       sim.stat(pop,popSize=True)
       subsize=pop.dvars().subPopSize
#If subsize is of length 1, then it is a integer and len() does not work
       if type(subsize) == type(1):
               numpop=1
       else:
               numpop=len(subsize)
       vecsize=[popsize]*(numpop)
       return vecsize
#-----Setting up the simulator-----
for k in range (1,101):
   print('Simulation number '+str(k))
   #Creating initial population of size popsize and "caryotype" vecloc
   pop=sim.Population(size=[popsize]*16,loci=vecloc,infoFields=['migrate_to','fitness','
       env'],lociNames=allele_naming(numchrom,numloc))
   sim.initSex(pop)
   sim.initSex(pop)
   for i in range(numloc*numchrom):
           for j in range(16):
                   tmp_p=numpy.random.beta(p[i]/0.11,(1-p[i])/0.11)
```

```
sim.initGenotype(pop,freq=[tmp_p,1-tmp_p],loci=i,subPops=j)
#-----Main evolving process----
pop.evolve(
       preOps=[
       sim.PyEval('"Gen: %d" % gen', step=10),
       sim.PyOutput('\n',step=10),
#Migration using a simple stepping stone model
       sim.Migrator(rate=migrSteppingStoneRates(m,16,circular=False)),
#Selection process
#Set environmental value (env infoField) for each individual in the population
#Takes place at each generation after the first fission
        sim.PyOperator(env_set),
#Selection occures at locus 50 according to env information field
       sim.PySelector(fit_env,loci=locisel,begin=atgen),
       1.
#Mating at random (pangamy)
       matingScheme=sim.RandomMating(
#Fixed population size (fixed at 'popsize')
       subPopSize=demo,
#Recombination to avoid selective sweep
       ops=[sim.Recombinator(rates=0.002)]
       ),
       postOps=[
#Mutation rate 10e-6
       sim.SNPMutator(u=0.000001,v=0.000001)
       ],
#-----
#Evolve for a number 'numgen' of generations
       gen = numgen
sim.stat(pop,popSize=True)
subsize=pop.dvars().subPopSize
numpop=len(subsize)
for i in range(numpop):
       pop.setIndInfo(vec_env[i], 'env', subPop=i)
sample = drawRandomSample(pop, sizes=[20]*numpop)
sample.addInfoFields('pop_name')
vecname=[]
for i in range(1,numpop+1):
       vecname=vecname+[i]*20
sample.setIndInfo(vecname, 'pop_name')
#sim.dump(pop,structure=False)
print('Saving population')
saveCSV(sample,filename="sims/sim"+str(k)+".csv",infoFields=['pop_name','env'],
    sexFormatter=None, affectionFormatter=None, header=False)
```

4 Source for: Hierarchically structured model – polygenic selection

```
print("Importing some other libraries")
from decimal import *
from math import *
import numpy
import random
print("Script starting")
#-----Simulation parameters-----
#Population(s) size(s)
popsize=500
#Number of generations
numgen=600
atgen=[50,150,200,300]
#Number of chromosomes
numchrom=10
#Number of loci per chrom.
numloc = 500
#Create vector of loci numbers
vecloc=[numloc]*numchrom
#Migration rate (probability for one individual to disperse)
m = 0.0045
#-----Initializing some variables-------
#Environmental values for the two first populations
vec_env
        = [-0.5198065, -0.8112612, -0.2124103, 0.3941398, 1.1103401, 1.5185518, 0.6871753, 0.0878634, -0.0821056, -0.8112612, -0.0821056, -0.8112612, -0.0821056, -0.8112612, -0.0821056, -0.8112612, -0.0821056, -0.8112612, -0.0821056, -0.8112612, -0.0821056, -0.8112612, -0.0821056, -0.8112612, -0.0821056, -0.8112612, -0.0821056, -0.8112612, -0.0821056, -0.8112612, -0.0821056, -0.8112612, -0.0821056, -0.8112612, -0.0821056, -0.8112612, -0.0821056, -0.8112612, -0.0821056, -0.8112612, -0.0821056, -0.8112612, -0.0821056, -0.8112612, -0.0821056, -0.8112612, -0.0821056, -0.8112612, -0.0821056, -0.8112612, -0.0821056, -0.8112612, -0.0821056, -0.8112612, -0.0821056, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.8112612, -0.811
#Defining the loci under selection
locisel
        =[2793,1850,583,4083,3349,860,4785,706,947,939,1819,925,403,2867,2897,97,3102,2618,708,1190,2471,15
#Function naming the alleles
#Depends on the number of chromosomes (chrom) and loci (loc)
def allele_naming(chrom,loc):
                res=[]
                 for i in range(chrom):
                                  for j in range(1,loc+1):
                                                   res.append(chr(65+i)+str(j))
                 return res
#Function operating the migration for each generation
def migration(pop):
#Extract the number of populations 'numpop'
                sim.stat(pop,popSize=True)
                 subsize=pop.dvars().subPopSize
                numpop=len(subsize)
#First iteration step is 2
#i reflects the number of pops (2,4,8,16...)
                i=2
\#j reflects the number of steps (1,2,3...)
                j=1
{\tt\#Seeding\ iterative\ process\ to\ construct\ migration\ rate\ matrix}
#a is a 2x2 matrix
                 a=zeros((2,2))
                 a[1][0]=1
                 a[0][1]=1
                while i<numpop:
\hbox{\it \#while the number of pops is not reached}
\#note that for numpop=2, the loop is not activate
                 #i doubles
                                 i = 2 * i
                 #incrementing j
                                 j+=1
```

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#tmp is a submatrix 2x2 containing the 'migration coefficient' i/2 (1/2 for 4 pops,
             1/4 for 8 pops, etc..)
                tmp=zeros(((i/2),(i/2)))+(float(1)/(i/2))
        #a is updated to contain the coefficients in anti-diag submatrices
        #a is now 4x4, then 8x8 for 16 pops
                a=hstack((vstack((a,tmp)),vstack((tmp,a))))
#End while
#the matrix needs to be scaled to sum to 1
#the sum to 1 is needed for m (migration rate) to be a relevant biological parameter
#Scaled by the number of steps j (each step adding exactly 1 to each row (1, or 2*1/2 or
    4*1/4, etc...))
       res=(a/j)*m
#res is an array, we need a nested list
        A=res.tolist()
#And now migration finally happens !
       sim.migrate(pop,rate=A)
        return True
#Function to update the environmental values for each population
#Env values are stored in the global variable vec_env
#New values are drawn from a normal distribution
#with the env value of the mother pop as mean
def env_update(pop):
        global vec_env
        sim.stat(pop,popSize=True)
        subsize=pop.dvars().subPopSize
       numpop=len(subsize)
#Already fixed for numpop == 2
        if numpop > 2:
        #k is the number to create the two new values (x=x0+k \text{ ou } x=x0-k)
                k=1.6/float(numpop)
        #tmp will recieve the new env values
                tmp = [0] * numpop
                for i in range(numpop):
                        #if we are left to the old value (x0)
                        if (i\%2==0):
                                 \#i/2 is the result of an euclidian division
                                 tmp[i]=round(vec_env[i/2]-k,1)
                        #else, we are right to the old value (x0)
                        else:
                                 tmp[i] = round(vec_env[i/2]+k,1)
                vec_env=tmp
        return True
#Function to attribute env values to individuals' info fields
def env_set(pop):
       sim.stat(pop,popSize=True)
        subsize=pop.dvars().subPopSize
        numpop=len(subsize)
        for i in range(numpop):
               pop.setIndInfo(vec_env[i], 'env', subPop=i)
        return True
\#Link function between env value and selection
#Modified 'logitistic' function
def fit func(x):
        res = (1 - exp(-x))/(1 + exp(-x))
        res=res/50
        return res
#Defining fitness according to selection
\#(0,0) is the reference genotype
def fit_env(geno,env):
       N=len(geno)
       s=fit_func(env)
        +.11 = 0
        t00 = 0
        for i in range(N/2):
                a1=geno[i*2]
                a2=geno[i*2+1]
                if (a1+a2==0):
```

```
t00=t00+1
                if (a1+a2==2):
                        t11 = t11 + 1
        w = ((1+s)**t11)*((1-s)**t00)
        return w
#Function defining a constant subPop size (popsize) for any number of subpops (demographic
   model)
def demo(pop):
       sim.stat(pop,popSize=True)
       subsize=pop.dvars().subPopSize
#If subsize is of length 1, then it is a integer and len() does not work
       if type(subsize) == type(1):
               numpop=1
        else:
               numpop=len(subsize)
        vecsize=[popsize]*(numpop)
        return vecsize
#-----Setting up the simulator--------
for k in range(1,101):
        #Environmental values for the two first populations print('Simulation number '+str(k))
        #Creating initial population of size popsize and "caryotype" vecloc
        pop=sim.Population(size=[popsize],loci=vecloc,infoFields=['migrate_to','fitness','
           env'],lociNames=allele_naming(numchrom, numloc))
                  -----Main evolving process-----
        pop.evolve(
        #Initializing sex and genotype
               initOps=[
                       sim.InitSex(),
                        sim.InitGenotype(freq=[0.5,0.5]),
                preOps=[
        #Splitting each population into two at 'atgen' generations
                sim.SplitSubPops(proportions=[0.5,0.5], at=atgen),
                sim.PyOutput('----*-Split-*----\n',at=atgen),
                sim.PyEval('"Gen: %d" % gen', step=10),
               sim.PyOutput('\n', step=10),
        #Calling function 'migration' for individuals migration
        #(Operator Migrator does not allow for varying number of subpopulations)
                sim.PyOperator(migration, begin=atgen[0]),
        #Selection process
        #Set environmental value (env infoField) for each individual in the population
        #Takes place at each generation after the first fission
                sim.PyOperator(env_set,begin=atgen[3]),
        #Selection occures at locus locisel according to env information field
                sim.PySelector(fit_env,loci=locisel,begin=atgen[3]),
               ],
        #Mating at random (pangamy)
                matingScheme=sim.RandomMating(
        #Fixed population size (fixed at 'popsize')
                subPopSize=demo,
        #Recombination to avoid selective sweep
                ops=[sim.Recombinator(rates=0.002)]
                ),
                postOps=[
        \#Mutation rate 10e-6
               sim.SNPMutator(u=0.000001, v=0.000001)
        #Evolve for a number 'numgen' of generations
               gen = numgen
        sim.stat(pop,popSize=True)
        subsize=pop.dvars().subPopSize
        numpop=len(subsize)
        for i in range(numpop):
               pop.setIndInfo(vec_env[i], 'env', subPop=i)
```