

oneFill_Bottleneck

December 20, 2017

0.1 Effects of migration and bottleneck on the strength and formation of cyanogenesis clines

In this notebook, I look at the effects of migration rate and population bottlenecks on the strength and proportion of significant positive and negative clines in cyanogenesis.

Positive clines Less HCN in urban environment.

Negative clines More HCN in urban environment.

```
In [1]: library(plyr)
library(ggplot2)
library(data.table, lib=~ /Rpackages")
library(Rmisc, lib = "~ /Rpackages")
library(dplyr)
library(broom)
```

Loading required package: lattice

Attaching package: dplyr

The following objects are masked from package:data.table:

between, first, last

The following objects are masked from package:plyr:

arrange, count, desc, failwith, id, mutate, rename, summarise, summarize

The following objects are masked from package:stats:

filter, lag

The following objects are masked from package:base:

intersect, setdiff, setequal, union

```
In [82]: #Working directory for datasets varying migration rate and bottleneck proportion
setwd('/scratch/research/projects/trifolium/SEC_Simulation.Evolutionary.Clines/SEC_Data
/drift-migration/1D/Mig_Bot_Vary')

#Load datasets that will be used for analyses
datSlopes <- fread('20171106_SlopeSum_Gen_BotMig-Merged_distRev.csv', header = T)
datFreqFirst <- fread('20171106_FreqFirstGen_BotMig-Merged_distRev.csv', header = T)
datSlopes$bot <- as.factor(as.character(datSlopes$bot))
datSlopes$Mig_rate <- as.factor(as.character(datSlopes$Mig_rate))
```

```

datFreqFirst$bot <- as.factor(as.character(datFreqFirst$bot))
datFreqFirst$Mig_rate <- as.factor(as.character(datFreqFirst$Mig_rate))

#Data subsets
datSlopes_GenOne <- subset(datSlopes, seq == "1")
datSlopes_NoMig <- subset(datSlopes, Mig_rate == "0")
datSlopes_GenOne_NoMig <- subset(datSlopes, seq == "1" & Mig_rate == "0")
datFreqFirst_StrongBot_NoMig <- subset(datFreqFirst, bot == "0.01" & Mig_rate == "0")
datFreqFirst_InterBot_NoMig <- subset(datFreqFirst, bot == "0.2" & Mig_rate == "0")
datFreqFirst_NoBot_NoMig <- subset(datFreqFirst, bot == "1" & Mig_rate == "0")

#Proportion of simulations with Cyan lost by distance under strong bottlenecks
datPropLost_StrongBot_NoMig <- datFreqFirst_StrongBot_NoMig %>%
  group_by(Distance) %>%
  summarize(n = n(),
            Lost = sum(Cyan == 0) / n,
            Fixed = sum(Cyan == 1) / n) %>%
  mutate(Founder = "Strong")
datPropLost_StrongBot_NoMig <- dplyr::select(datPropLost_StrongBot_NoMig, Distance,
                                           Lost, Founder)

#Proportion of simulations with Cyan lost by distance under intermediate bottlenecks
datPropLost_InterBot_NoMig <- datFreqFirst_InterBot_NoMig %>%
  group_by(Distance) %>%
  summarize(n = n(),
            Lost = sum(Cyan == 0) / n,
            Fixed = sum(Cyan == 1) / n) %>%
  mutate(Founder = "Intermediate")
datPropLost_InterBot_NoMig <- dplyr::select(datPropLost_InterBot_NoMig, Distance,
                                           Lost, Founder)

#Proportion of simulations with Cyan lost by distance under no bottlenecks
datPropLost_NoBot_NoMig <- datFreqFirst_NoBot_NoMig %>%
  group_by(Distance) %>%
  summarize(n = n(),
            Lost = sum(Cyan == 0) / n,
            Fixed = sum(Cyan == 1) / n) %>%
  mutate(Founder = "None")
datPropLost_NoBot_NoMig <- dplyr::select(datPropLost_NoBot_NoMig, Distance,
                                           Lost, Founder)

# Merge datasets from strong and intermediate bottlenecks above
datPropLost_merged <- rbind(datPropLost_StrongBot_NoMig, datPropLost_InterBot_NoMig,
                             datPropLost_NoBot_NoMig)

# Calculate mean frequency in first generation across simulations for each distance
MeanFreqFirstGen_Distance <- datFreqFirst %>%
  group_by(Distance, bot, Mig_rate) %>%
  summarize(Freq = mean(Cyan))

MeanFreqFirstGen_Distance_StrongBot_NoMig <- subset(MeanFreqFirstGen_Distance, bot ==
"0.01" & Mig_rate == "0")
MeanFreqFirstGen_Distance_InterBot_NoMig <- subset(MeanFreqFirstGen_Distance, bot ==
"0.2" & Mig_rate == "0")
MeanFreqFirstGen_Distance_NoBot_NoMig <- subset(MeanFreqFirstGen_Distance, bot == "1" &
Mig_rate == "0")

```

Read 1200000 rows and 12 (of 12) columns from 0.046 GB file in 00:00:03

In [16]: str(datSlopes)

```

Classes data.table and 'data.frame': 15000 obs. of 20 variables:
 $ bot      : Factor w/ 10 levels "0.01","0.025",...: 1 1 1 1 1 1 1 1 1 1 ...
 $ Mig_rate : Factor w/ 3 levels "0","0.01","0.05": 1 1 1 1 1 1 1 1 1 1 ...

```

```

$ seq      : int  1 2 3 4 5 6 7 8 9 10 ...
$ mean     : num  0.00151 0.0015 0.0015 0.0015 0.0015 ...
$ sd       : num  0.00381 0.00381 0.00381 0.0038 0.0038 ...
$ n        : int  1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 ...
$ se       : num  0.000121 0.00012 0.00012 0.00012 0.00012 ...
$ ci_mean  : num  0.000236 0.000236 0.000236 0.000236 0.000236 ...
$ prop_sigPos: num  0.421 0.422 0.42 0.422 0.42 0.42 0.415 0.418 0.419 0.416 ...
$ prop_pos  : num  0.74 0.74 0.739 0.739 0.739 0.739 0.739 0.739 0.739 0.739 ...
$ se_pos    : num  0.0139 0.0139 0.0139 0.0139 0.0139 ...
$ ci_pos    : num  0.0272 0.0272 0.0272 0.0272 0.0272 ...
$ se_sigPos : num  0.0156 0.0156 0.0156 0.0156 0.0156 ...
$ ci_sigPos : num  0.0306 0.0306 0.0306 0.0306 0.0306 ...
$ prop_sigNeg: num  0.176 0.177 0.176 0.175 0.175 0.177 0.173 0.17 0.175 0.174 ...
$ prop_neg  : num  0.241 0.241 0.241 0.241 0.241 0.241 0.241 0.241 0.241 0.241 ...
$ se_neg    : num  0.0135 0.0135 0.0135 0.0135 0.0135 ...
$ ci_neg    : num  0.0265 0.0265 0.0265 0.0265 0.0265 ...
$ se_sigNeg : num  0.012 0.0121 0.012 0.012 0.012 ...
$ ci_sigNeg : num  0.0236 0.0237 0.0236 0.0236 0.0236 ...
- attr(*, ".internal.selfref")=<externalptr>

```

In [4]: str(datFreqFirst)

Classes data.table and 'data.frame': 1200000 obs. of 12 variables:

```

$ x        : int  0 0 0 0 0 0 0 0 0 0 ...
$ y        : int  0 0 0 0 0 0 0 0 0 0 ...
$ bot      : num  0.01 0.01 0.01 0.01 0.01 0.01 0.01 0.01 0.01 0.01 ...
$ Sim      : int  0 0 0 1 1 1 2 2 2 3 ...
$ Generation: int  1 1 1 1 1 1 1 1 1 1 ...
$ Cyan     : num  0.563 0.563 0.563 0.563 0.563 0.563 0.563 0.563 0.563 0.563 ...
$ Mat_full : int  0 0 0 0 0 0 0 0 0 0 ...
$ Pop_size : int  1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 ...
$ Mig_rate : num  0 0.01 0.05 0 0.01 0.05 0 0.01 0.05 0 ...
$ pA       : num  0.5 0.5 0.5 0.5 0.5 0.5 0.5 0.5 0.5 0.5 ...
$ pB       : num  0.5 0.5 0.5 0.5 0.5 0.5 0.5 0.5 0.5 0.5 ...
$ Distance : int  40 40 40 40 40 40 40 40 40 40 ...
- attr(*, ".internal.selfref")=<externalptr>

```

In [5]: *#Themes used for plotting*

```

ng1=theme(aspect.ratio=0.7,panel.background = element_blank(),
          panel.grid.major = element_blank(),
          panel.grid.minor = element_blank(),
          panel.border=element_blank(),
          axis.line.x = element_line(color="black",size=1),
          axis.line.y = element_line(color="black",size=1),
          axis.ticks=element_line(color="black"),
          axis.text=element_text(color="black",size=15),
          axis.title=element_text(color="black",size=1),
          axis.title.y=element_text(vjust=2,face="bold",size=15),
          axis.title.x=element_text(vjust=0.1,face="bold",size=15),
          axis.text.x=element_text(size=13),
          axis.text.y=element_text(size=13),
          legend.position = "right", legend.direction="vertical",
          legend.text=element_text(size=11), legend.key = element_rect(fill = "white"),
          legend.title = element_text(size=13,face="bold"),legend.key.size = unit(0.5,
"cm"))

ng1.45=theme(aspect.ratio=0.7,panel.background = element_blank(),
            panel.grid.major = element_blank(),
            panel.grid.minor = element_blank(),
            panel.border=element_blank(),

```

```

axis.line.x = element_line(color="black",size=1),
axis.line.y = element_line(color="black",size=1),
axis.ticks=element_line(color="black"),
axis.text=element_text(color="black",size=15),
axis.title=element_text(color="black",size=1),
axis.title.y=element_text(vjust=2,face="bold",size=15),
axis.title.x=element_text(vjust=0.1,face="bold",size=15),
axis.text.x=element_text(size=13,angle=45,hjust=1),
axis.text.y=element_text(size=13),
legend.position = "right", legend.direction="vertical",
legend.text=element_text(size=11), legend.key = element_rect(fill = "white"),
legend.title = element_text(size=13,face="bold"),legend.key.size = unit(0.5,
"cm"))

```

```

In [10]: path = "/scratch/research/projects/trifolium/SEC_Simulation.Evolutionary.Clines/SEC_Sync
/SEC_Figures/Drift.Migration/Mig_Bot_Vary"

```

```

# ggsave("Mean-slope_BotMig.pdf", plot = MeanSlope_BotMig, device = "pdf", width = 6.0,
height = 6.0, path = path, dpi = 600)
# ggsave("PropSigPos_BotMig.pdf", plot = PropSigPos_BotMig, device = "pdf", width = 6.0,
height = 6.0, path = path, dpi = 600)
# ggsave("propLost_bot_NoMig.pdf", plot = propLost_bot_NoMig, device = "pdf", width =
6.0, height = 6.0, path = path, dpi = 600)
# ggsave("Regression_LostAtGen2_Bot001_NpMig.pdf", plot = LostAt2_bot001_NoMig, device =
"pdf", width = 6.0, height = 6.0, path = path, dpi = 600)
# ggsave("Regression_NotLost_Bot02_NoMig.pdf", plot = NotLost_bot02_NoMig, device =
"pdf", width = 6.0, height = 6.0, path = path, dpi = 600)
# ggsave("PropNeg_BotMig.pdf", plot = PropNeg_BotMig, device = "pdf", width = 6.0,
height = 6.0, path = path, dpi = 600)
ggsave("Mean-slope_Bot_HighMig.pdf", plot = MeanSlope_Bot_HighMig, device = "pdf", width
= 6.0, height = 6.0, path = path, dpi = 600)
ggsave("PropSigPos_Bot_HighMig.pdf", plot = PropPos_Bot_HighMig, device = "pdf", width =
6.0, height = 6.0, path = path, dpi = 600)

```

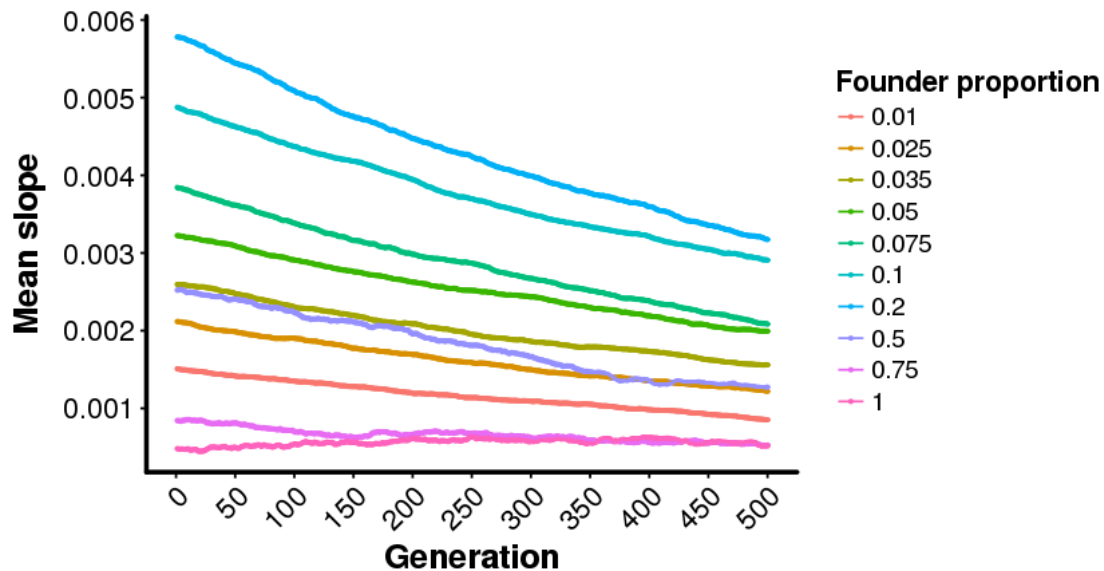
0.2 Mean slope with founder proportion under varying migration rates

Here I look at the mean slope across simulations under varying founding proportions and migration rates. Mean slope is calculated in the first generation following the filling of the landscape matrix. I also plot the change in the mean slope with increasing generations.

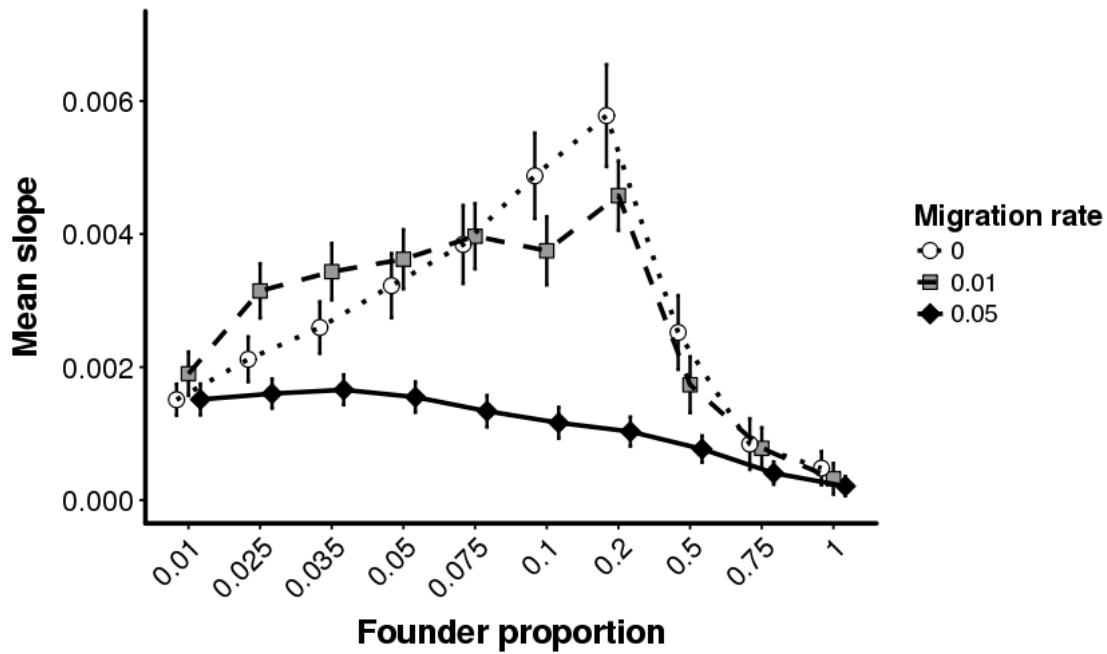
```

In [18]: MeanSlope_Cyan_byGen <- ggplot(datSlopes_NoMig, aes(x = seq, y = mean, group = bot,
color = bot)) +
  ylab("Mean slope") + xlab("Generation") + geom_point(size = 0.5, aes(color = bot)) +
  geom_line(size = 0.5, aes(color = bot)) +
  coord_cartesian(xlim = c(0, 500)) + scale_x_continuous(breaks = seq(from = 0, to =
500, by = 50)) +
  labs(color = "Founder proportion") + ngl.45
MeanSlope_Cyan_byGen

```



```
In [9]: MeanSlope_BotMig <- ggplot(datSlopes_GenOne, aes(x = bot, y = mean, group = Mig_rate)) +
  geom_errorbar(aes(ymin = mean - ci_mean, ymax = mean + ci_mean), width=0.15,
    size=0.7,
    position = position_dodge(width = 0.5)) +
  geom_point(size = 3, aes(fill = Mig_rate, shape = Mig_rate), position =
    position_dodge(width = 0.5)) +
  geom_line(size = 1, aes(linetype = Mig_rate), position = position_dodge(width =
    0.5)) +
  scale_shape_manual(labels = c("0", "0.01", "0.05"), values = c(21, 22, 23)) +
  scale_fill_manual(labels = c("0", "0.01", "0.05"), values = c("white", "grey60",
    "black")) +
  scale_linetype_manual(labels = c("0", "0.01", "0.05"), values = c("dotted",
    "dashed", "solid")) +
  coord_cartesian(ylim = c(0, 0.007)) + scale_y_continuous(breaks = seq(from = 0, to =
    0.006, by = 0.002)) +
  ylab("Mean slope") + xlab("Founder proportion") +
  labs(fill = "Migration rate", shape = "Migration rate", linetype = "Migration rate")
+ ng1.45
MeanSlope_BotMig
```



0.3 Effects of founder events and migration on proportion of negative and positive clines

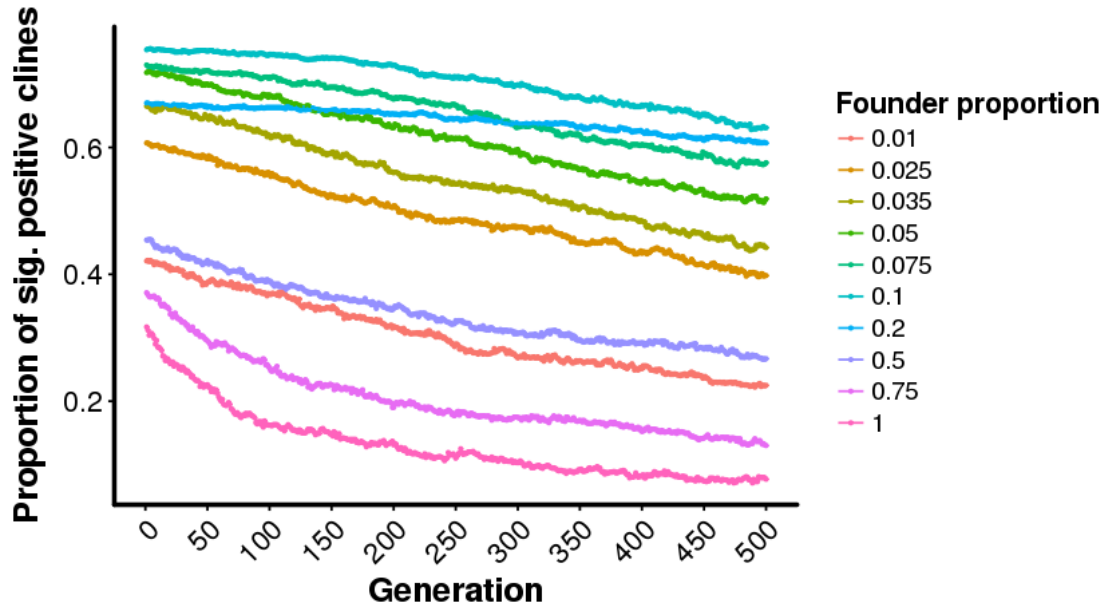
Here I look at how founder effects and migration affect the proportion of significantly negative and positive clines in the *Ac* and *Li* alleles and in HCN clines.

Positive clines: Less HCN in urban environment (i.e. loss of HCN across space)

Negative clines: More HCN in urban environment (i.e. gain in HCN across space)

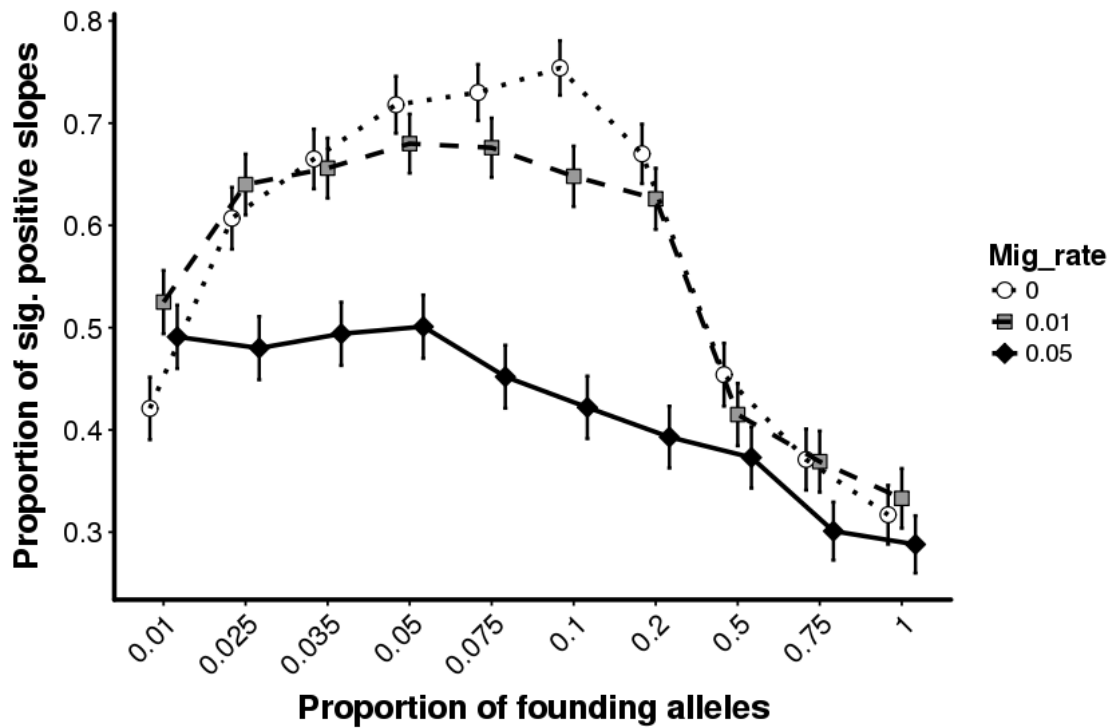
```
In [21]: PropSigPos_Cyan_byGen <- ggplot(datSlopes_NoMig, aes(x = seq, y = prop_sigPos, group =
  bot, color = bot)) +
  ylab("Proportion of sig. positive clines") + xlab("Generation") + geom_point(size =
  0.5, aes(color = bot)) +
  geom_line(size = 0.5, aes(color = bot)) +
  coord_cartesian(xlim = c(0, 500)) + scale_x_continuous(breaks = seq(from = 0, to =
  500, by = 50)) +
```

```
labs(color = "Founder proportion") + ng1.45
PropSigPos_Cyan_byGen
```



```
In [91]: PropSigPos_BotMig <- ggplot(datSlopes_GenOne, aes(x = bot, y = prop_sigPos, group =
  Mig_rate)) +
  geom_errorbar(aes(ymin = prop_sigPos - ci_sigPos, ymax = prop_sigPos + ci_sigPos),
width=0.15, size=0.7,
  position = position_dodge(width = 0.5)) +
  geom_point(size = 3, aes(fill = Mig_rate, shape = Mig_rate), position =
position_dodge(width = 0.5)) +
  geom_line(size = 1, aes(linetype = Mig_rate), position = position_dodge(width =
0.5)) +
  scale_shape_manual(labels = c("0", "0.01", "0.05"), values = c(21, 22, 23)) +
  scale_fill_manual(labels = c("0", "0.01", "0.05"), values = c("white", "grey60",
"black")) +
  scale_linetype_manual(labels = c("0", "0.01", "0.05"), values = c("dotted",
"dashed", "solid")) +
  ylab("Proportion of sig. positive slopes") + xlab("Proportion of founding alleles")
```

```
+ ng1.45
PropSigPos_BotMig
```



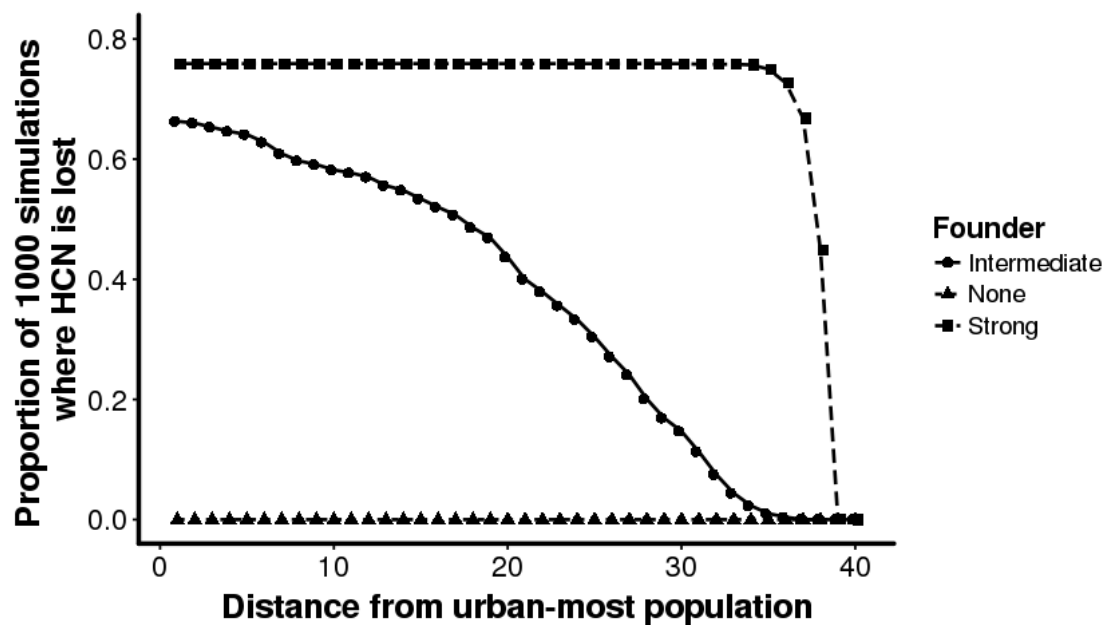
0.4 Dynamics of HCN frequency change during colonization

Here I try to understand the hump-shaped distribution above by exploring how the frequency of HCN changes during the colonization process.

```
In [68]: propLost_bot_NoMig <- ggplot(datPropLost_merged, aes(x = Distance, y = Lost, group =
  Founder)) +
  geom_point(size = 2, aes(fill = Founder, shape = Founder), position =
  position_dodge(width = 0.5)) +
  geom_line(size = 0.75, aes(linetype = Founder)) +
  ylab("Proportion of 1000 simulations
  where HCN is lost") + xlab("Distance from urban-most population") +
  coord_cartesian(ylim = c(0, 0.8)) + scale_y_continuous(breaks = seq(from = 0, to =
```



```
0.8, by = 0.2)) + ng1
propLost_bot_NoMig
```



```
In [72]: lm_FreqFirstGen_StrongBot_NoMig <- lm(Freq ~ Distance, data =
MeanFreqFirstGen_Distance_StrongBot_NoMig)
summary(lm_FreqFirstGen_StrongBot_NoMig)
```

Call:

```
lm(formula = Freq ~ Distance, data = MeanFreqFirstGen_Distance_StrongBot_NoMig)
```

Residuals:

Min	1Q	Median	3Q	Max
-0.05468	-0.03498	-0.01131	0.01506	0.24982

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	0.2049757	0.0196579	10.427	1.05e-12 ***
Distance	0.0027050	0.0008356	3.237	0.0025 **

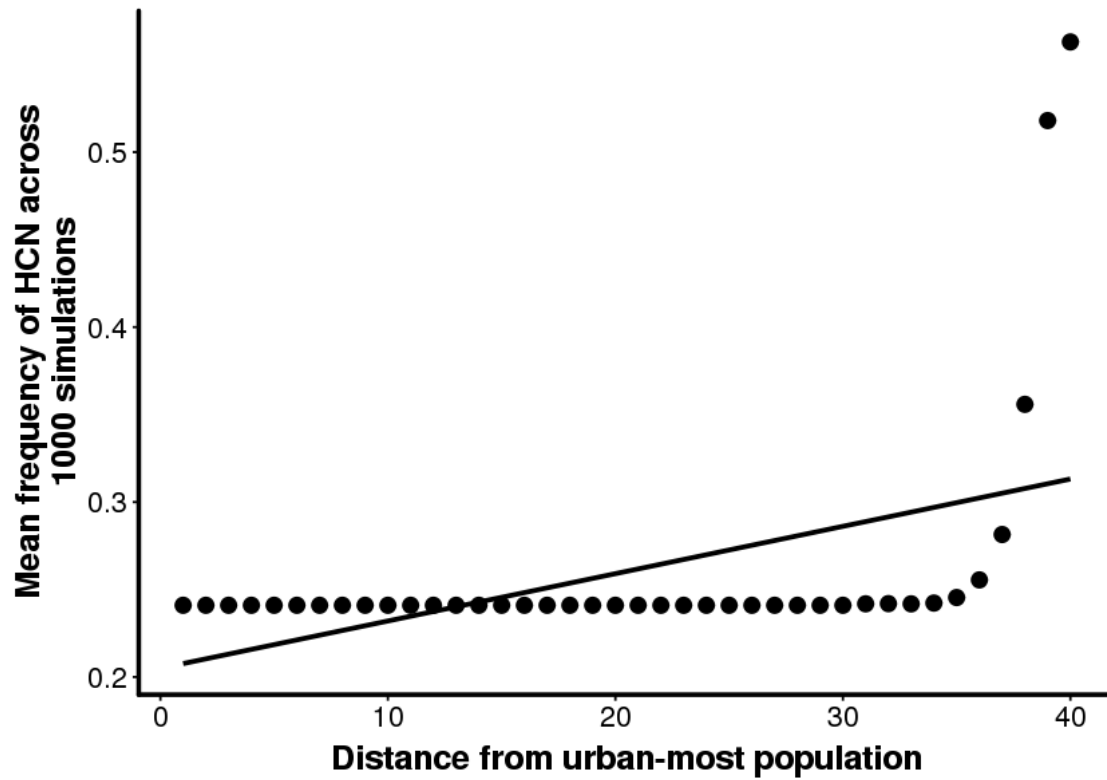
Signif. codes: 0 *** 0.001 ** 0.01 * 0.05 . 0.1 1

Residual standard error: 0.061 on 38 degrees of freedom

Multiple R-squared: 0.2162, Adjusted R-squared: 0.1956

F-statistic: 10.48 on 1 and 38 DF, p-value: 0.002504

```
In [74]: FreqFirstGen_StrongBot_NoMig <- ggplot(MeanFreqFirstGen_Distance_StrongBot_NoMig, aes(x
= Distance, y = Freq)) +
  geom_point(size = 3) +
  geom_smooth(method = "lm", se = F, colour = "black", size = 1.05) +
  ylab("Mean frequency of HCN across
1000 simulations") + xlab("Distance from urban-most population") +
  scale_y_continuous(breaks = seq(from = 0, to = 0.5, by = 0.1)) + ng1
FreqFirstGen_StrongBot_NoMig
```



```
In [76]: lm_FreqFirstGen_InterBot_NoMig <- lm(Freq ~ Distance, data =
MeanFreqFirstGen_Distance_InterBot_NoMig)
summary(lm_FreqFirstGen_InterBot_NoMig)
```

Call:

```
lm(formula = Freq ~ Distance, data = MeanFreqFirstGen_Distance_InterBot_NoMig)
```

Residuals:

Min	1Q	Median	3Q	Max
-0.028870	-0.021871	-0.007804	0.019546	0.041450

Coefficients:

Estimate	Std. Error	t value	Pr(> t)

```
(Intercept) 0.2362970 0.0074957 31.52 <2e-16 ***
Distance    0.0073500 0.0003186 23.07 <2e-16 ***
```

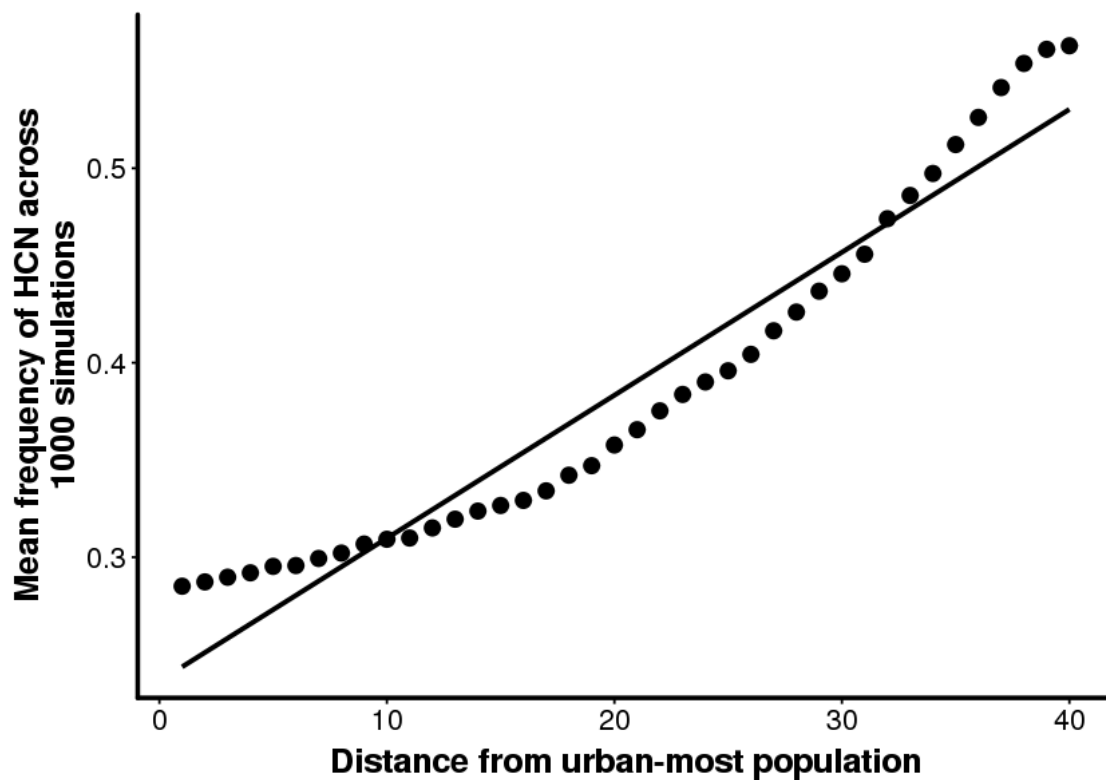
Signif. codes: 0 *** 0.001 ** 0.01 * 0.05 . 0.1 1

Residual standard error: 0.02326 on 38 degrees of freedom

Multiple R-squared: 0.9334, Adjusted R-squared: 0.9316

F-statistic: 532.2 on 1 and 38 DF, p-value: < 2.2e-16

```
In [77]: FreqFirstGen_InterBot_NoMig <- ggplot(MeanFreqFirstGen_Distance_InterBot_NoMig, aes(x =
  Distance, y = Freq)) +
  geom_point(size = 3) +
  geom_smooth(method = "lm", se = F, colour = "black", size = 1.05) +
  ylab("Mean frequency of HCN across
  1000 simulations") + xlab("Distance from urban-most population") + ng1
  FreqFirstGen_InterBot_NoMig + ng1
```



```
In [78]: lm_FreqFirstGen_NoBot_NoMig <- lm(Freq ~ Distance, data =
        MeanFreqFirstGen_Distance_NoBot_NoMig)
        summary(lm_FreqFirstGen_NoBot_NoMig)
```

Call:

```
lm(formula = Freq ~ Distance, data = MeanFreqFirstGen_Distance_NoBot_NoMig)
```

Residuals:

	Min	1Q	Median	3Q	Max
	-2.333e-03	-1.134e-03	-6.130e-06	9.098e-04	2.859e-03

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	0.5254653	0.0004212	1247.42	<2e-16 ***
Distance	0.0009125	0.0000179	50.96	<2e-16 ***

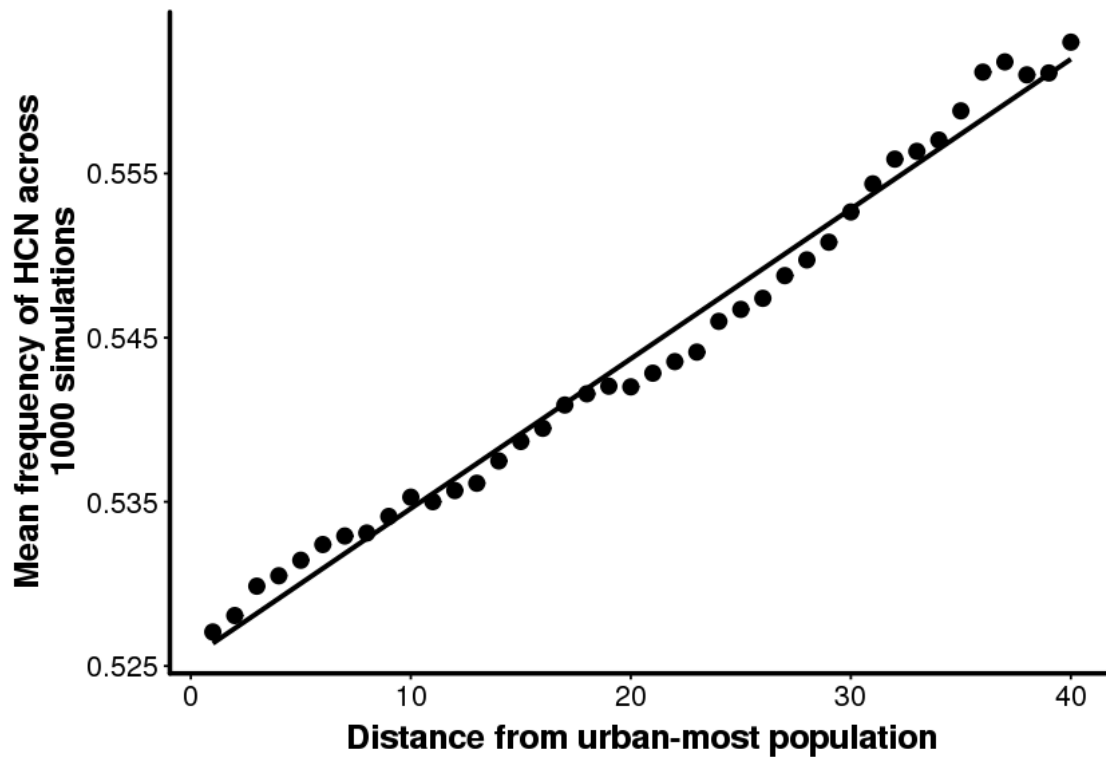
Signif. codes: 0 *** 0.001 ** 0.01 * 0.05 . 0.1 1

Residual standard error: 0.001307 on 38 degrees of freedom

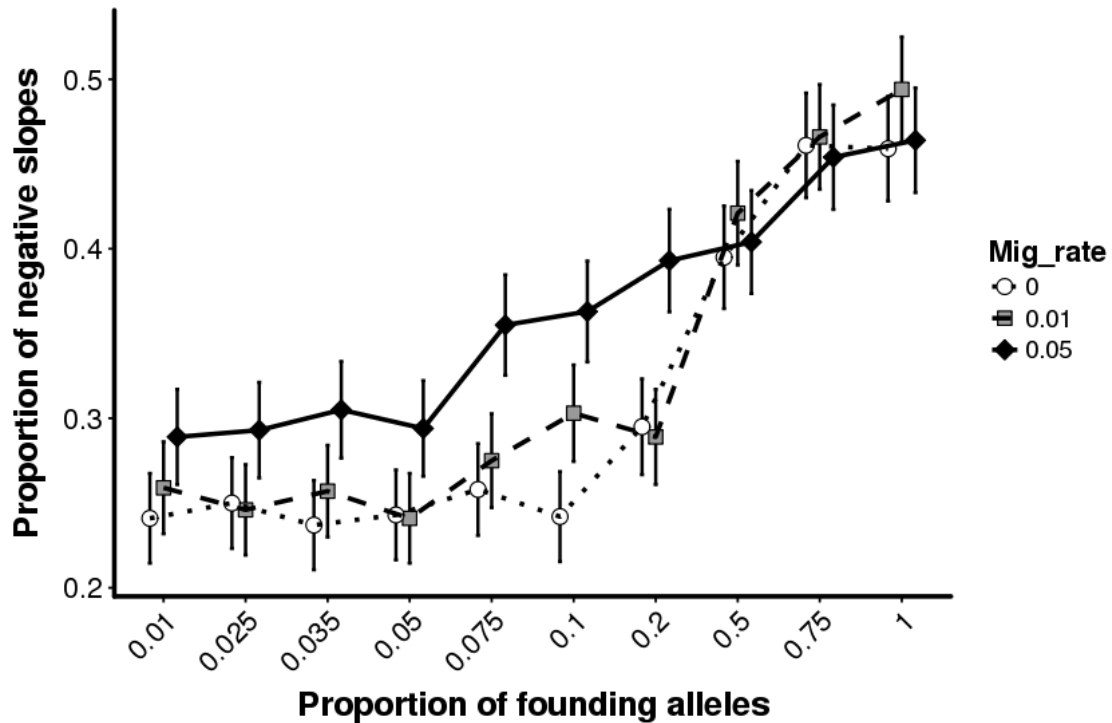
Multiple R-squared: 0.9856, Adjusted R-squared: 0.9852

F-statistic: 2597 on 1 and 38 DF, p-value: < 2.2e-16

```
In [79]: FreqFirstGen_NoBot_NoMig <- ggplot(MeanFreqFirstGen_Distance_NoBot_NoMig, aes(x =
        Distance, y = Freq)) +
        geom_point(size = 3) +
        geom_smooth(method = "lm", se = F, colour = "black", size = 1.05) +
        ylab("Mean frequency of HCN across
        1000 simulations") + xlab("Distance from urban-most population") + ng1
        FreqFirstGen_NoBot_NoMig + ng1
```



```
In [89]: PropNeg_BotMig <- ggplot(datSlopes_GenOne, aes(x = bot, y = prop_neg, group = Mig_rate))
+
  geom_errorbar(aes(ymin = prop_neg - ci_neg, ymax = prop_neg + ci_neg), width=0.15,
size=0.7,
  position = position_dodge(width = 0.5)) +
  geom_point(size = 3, aes(fill = Mig_rate, shape = Mig_rate), position =
position_dodge(width = 0.5)) +
  geom_line(size = 1, aes(linetype = Mig_rate), position = position_dodge(width =
0.5)) +
  scale_shape_manual(labels = c("0", "0.01", "0.05"), values = c(21, 22, 23)) +
  scale_fill_manual(labels = c("0", "0.01", "0.05"), values = c("white", "grey60",
"black")) +
  scale_linetype_manual(labels = c("0", "0.01", "0.05"), values = c("dotted",
"dashed", "solid")) +
  ylab("Proportion of negative slopes") + xlab("Proportion of founding alleles") +
ngl.45
PropNeg_BotMig + ngl.45
```



```
In [87]: #Extra columns that will be used to create melting dataset for plotting proportion of
positive and negative slopes
SlopeSum_Melt_NoMig <- dplyr::select(datSlopes_GenOne_NoMig, bot, prop_sigPos,
ci_sigPos, prop_sigNeg,
ci_sigNeg)

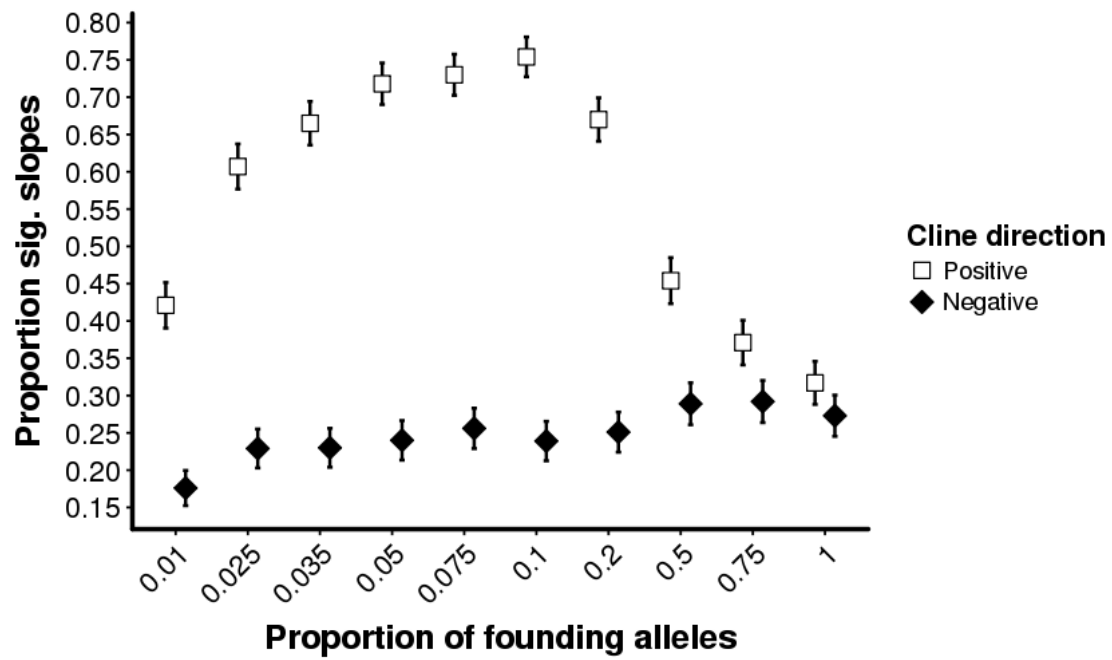
#Melt dataframe
dm1 <- melt(datSlopes_GenOne_NoMig[,c("bot", "prop_sigPos", "ci_sigPos")],
id=c("bot", "ci_sigPos"))
dm2 <- melt(datSlopes_GenOne_NoMig[,c("bot", "prop_sigNeg", "ci_sigNeg")],
id=c("bot", "ci_sigNeg"))

#Rename columns
setnames(dm1, old = "ci_sigPos", new = "ci")
setnames(dm2, old = "ci_sigNeg", new = "ci")

#Merge melted dataframes
SlopeSum_Melt_NoMig <- rbind(dm1, dm2)
SlopeSum_Melt_NoMig
```

bot	ci	variable	value
0.01	0.03060106	prop_sigPos	0.421
0.025	0.03027239	prop_sigPos	0.607
0.035	0.02925427	prop_sigPos	0.665
0.05	0.02788964	prop_sigPos	0.718
0.075	0.02751689	prop_sigPos	0.730
0.1	0.02669373	prop_sigPos	0.754
0.2	0.02914409	prop_sigPos	0.670
0.5	0.03085889	prop_sigPos	0.454
0.75	0.02994114	prop_sigPos	0.371
1	0.02884005	prop_sigPos	0.317
0.01	0.02360348	prop_sigNeg	0.176
0.025	0.02604360	prop_sigNeg	0.229
0.035	0.02608347	prop_sigNeg	0.230
0.05	0.02647089	prop_sigNeg	0.240
0.075	0.02704970	prop_sigNeg	0.256
0.1	0.02643305	prop_sigNeg	0.239
0.2	0.02687409	prop_sigNeg	0.251
0.5	0.02809570	prop_sigNeg	0.289
0.75	0.02818150	prop_sigNeg	0.292
1	0.02761243	prop_sigNeg	0.273

```
In [90]: #Plot proportion of significant slopes by migration rate
PropSig_Bot_NoMig <- ggplot(SlopeSum_Melt_NoMig, aes(x = factor(bot), y = value,
  shape = variable, fill = variable)) +
  geom_errorbar(aes(ymin = value - ci, ymax = value + ci), width=0.15, size = 0.7,
    position = position_dodge(width = 0.55)) +
  ylab("Proportion sig. slopes") + xlab("Proportion of founding alleles") +
  geom_point(size = 3.5, color = "black",
    position = position_dodge(width = 0.55)) +
  scale_shape_manual(labels = c("Positive", "Negative"), values=c(22, 23))+
  scale_fill_manual(labels = c("Positive", "Negative"), values=c("white", "black")) +
  scale_y_continuous(breaks = seq(from = 0, to = 1.0, by = 0.05)) +
  labs(shape = 'Cline direction', fill = 'Cline direction') + ng1.45
PropSig_Bot_NoMig
```

0.5 Do we need to show what's happening to the individual alleles?