

oneFill_Bottleneck

December 20, 2017

1 Serial founder events with migration, founder events.

In these simulations, the urban environment was colonized through serial founder events beginning from a single rural population at carrying capacity. We performed simulations under 3 migration rates: 0, 0.01, 0.05. We used 10 founding proportions: 0.01; 0.02; 0.035; 0.05; 0.075; 0.1; 0.2; 0.5; 0.75; 1.0. The parameter combinations were as follows (all varied parameters crossed factorially, total simulations = 30):

- Number of simulations: 1000
- Number of generations: 500 (following filling of matrix)
- Max Creation probability: 1.0
- Bottleneck proportion: [0.01; 0.02; 0.035; 0.05; 0.075; 0.1; 0.2; 0.5; 0.75; 1.0]
- pA and pB: 0.5
- Migration rate: [0; 0.01; 0.05]
- Maximum K (rural): 1000
- Minimum K (urban): 1000

```
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.Clones/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)

```

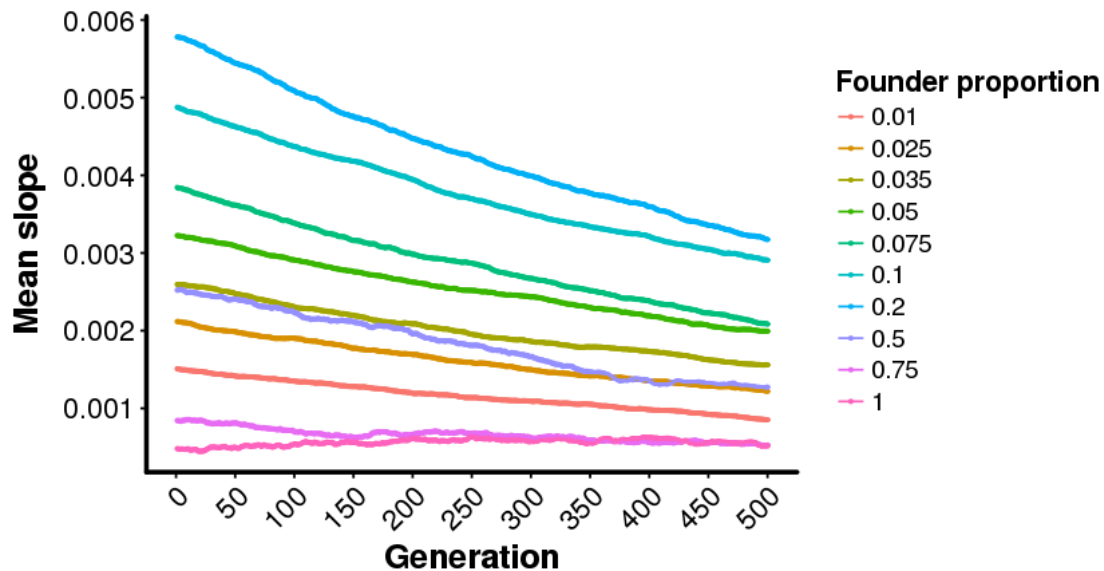
1.1 Mean slope with founder proportion under varying migration rates

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.

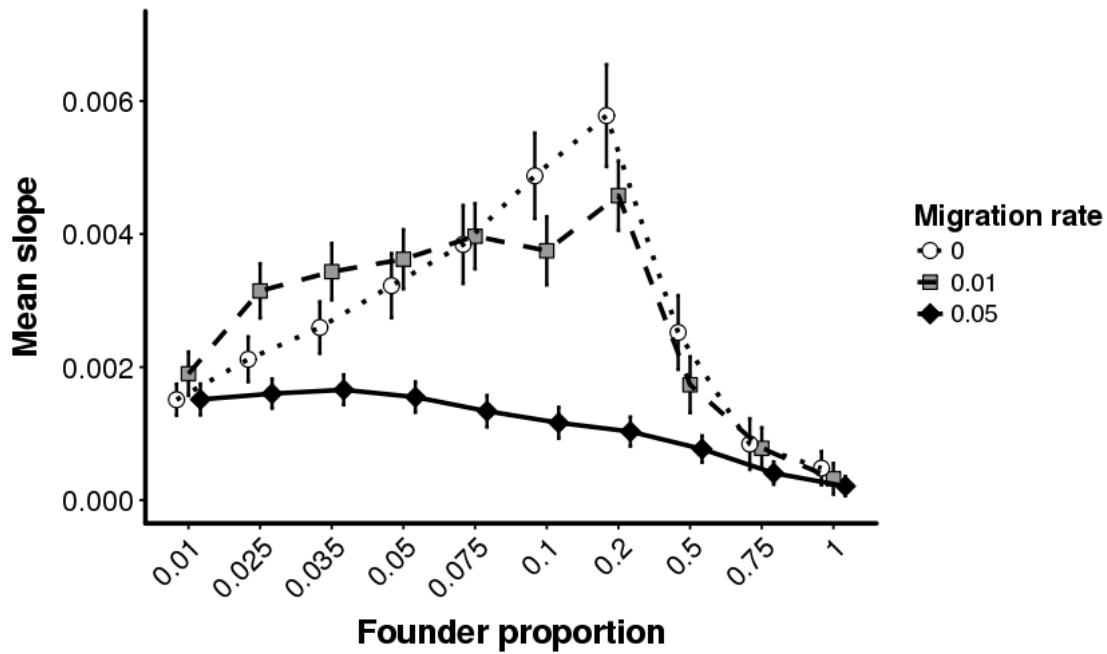
```

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") + ng1.45
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
```



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

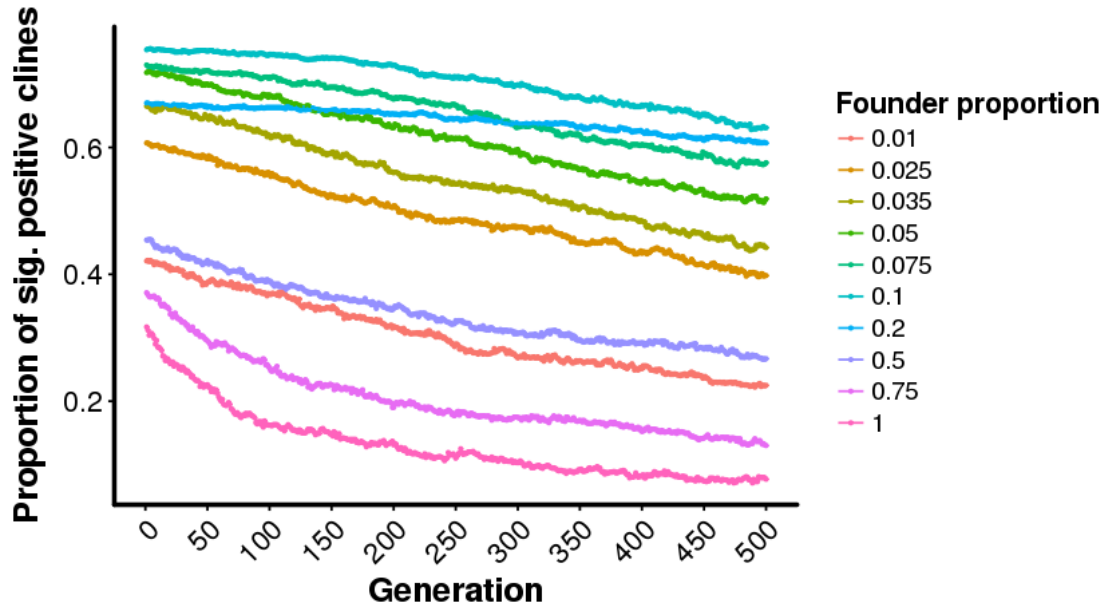
Here I look at how founder effects and migration affect the proportion of significantly negative and positive clines in the *Ac* and *Li* alleles and in HCN clines.

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)

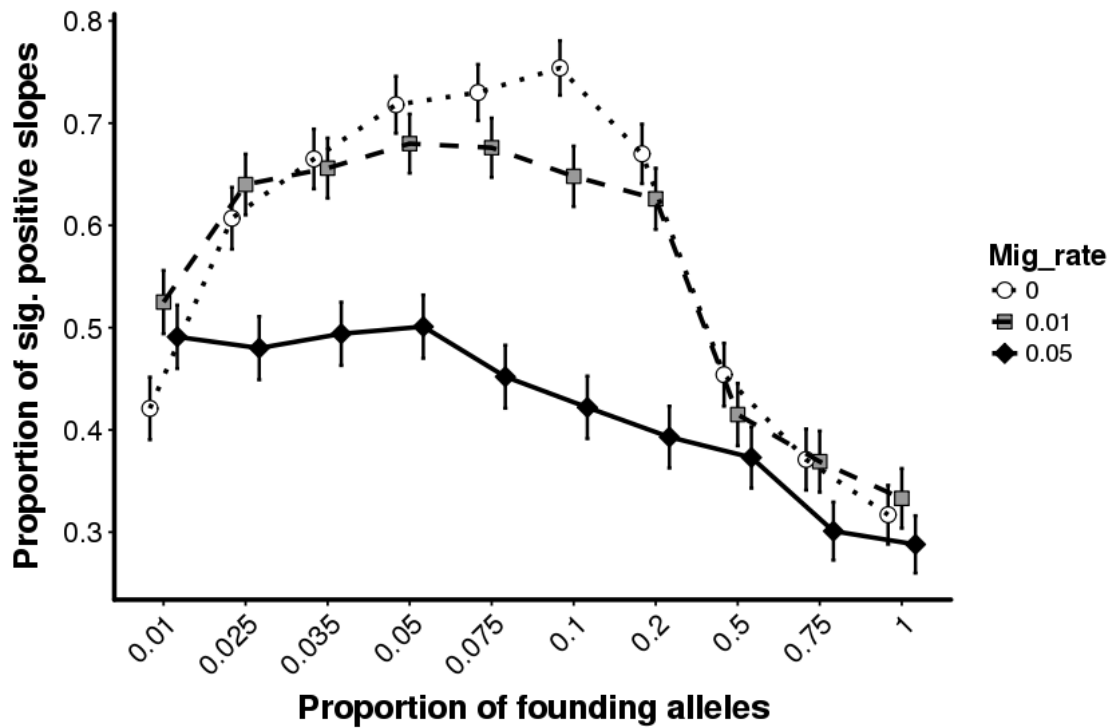
```
In [21]: PropSigPos_Cyan_byGen <- ggplot(datSlopes_NoMig, aes(x = seq, y = prop_sigPos, group =
  bot, color = bot)) +
  ylab("Proportion of sig. positive clines") + xlab("Generation") + geom_point(size =
  0.5, aes(color = bot)) +
  geom_line(size = 0.5, aes(color = bot)) +
  coord_cartesian(xlim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



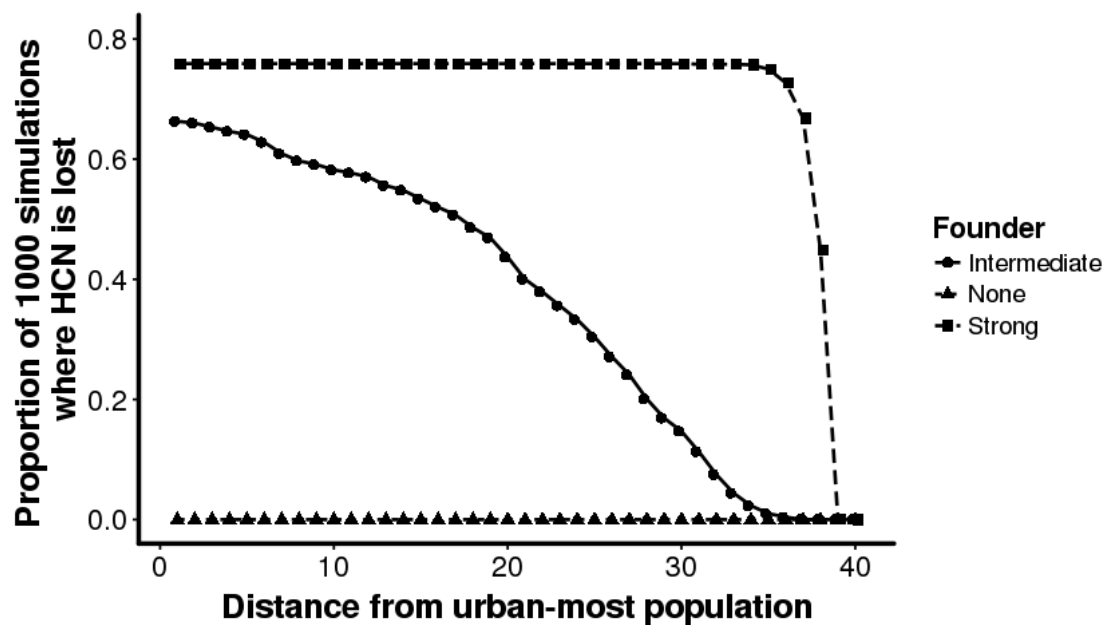
1.3 Dynamics of HCN frequency change during colonization

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 **

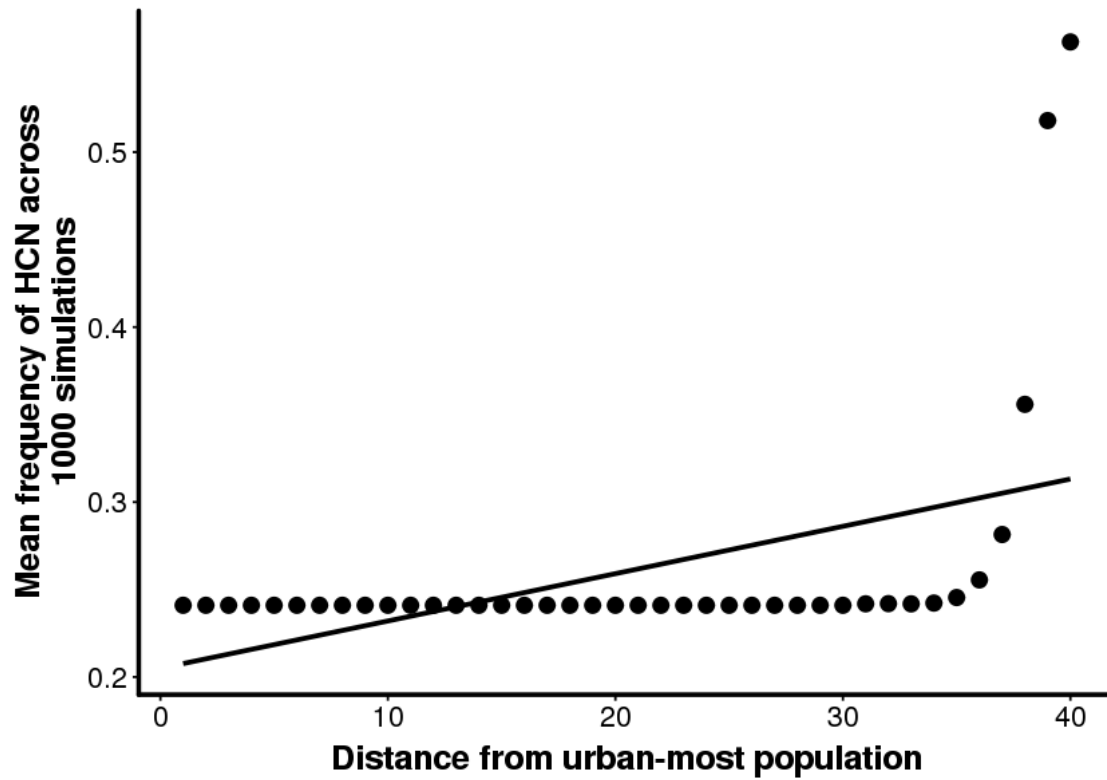
Signif. codes: 0 *** 0.001 ** 0.01 * 0.05 . 0.1 1

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 ***
```

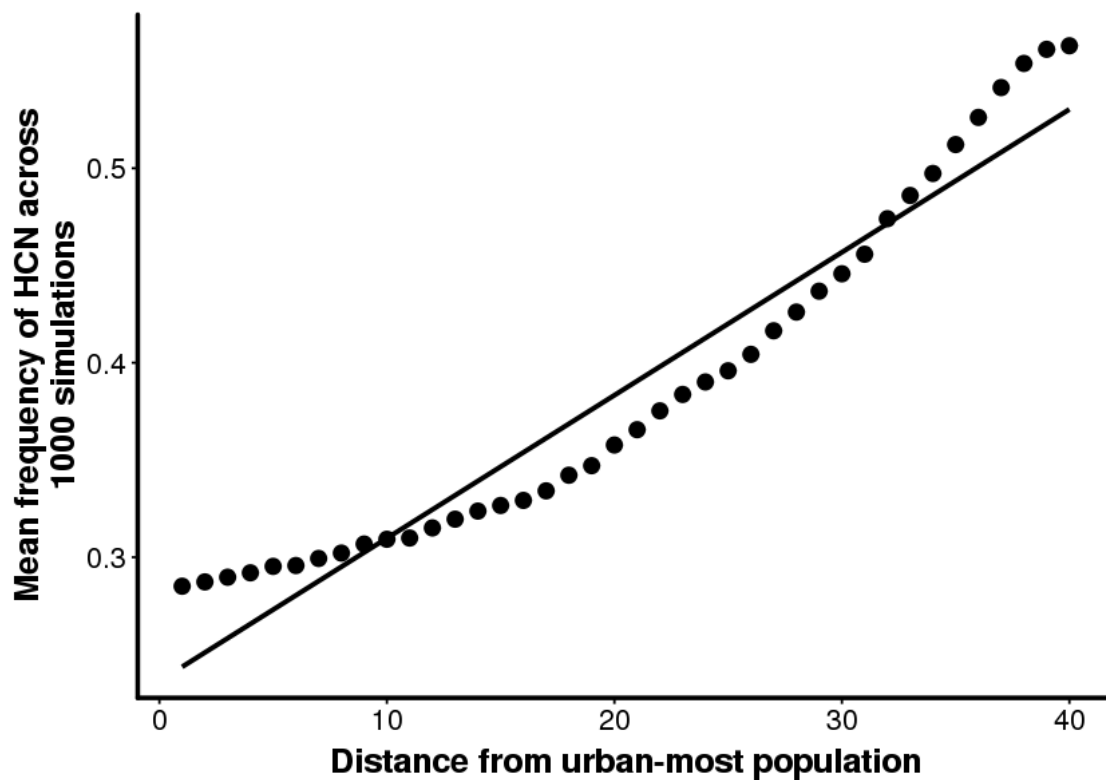
Signif. codes: 0 *** 0.001 ** 0.01 * 0.05 . 0.1 1

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 ***

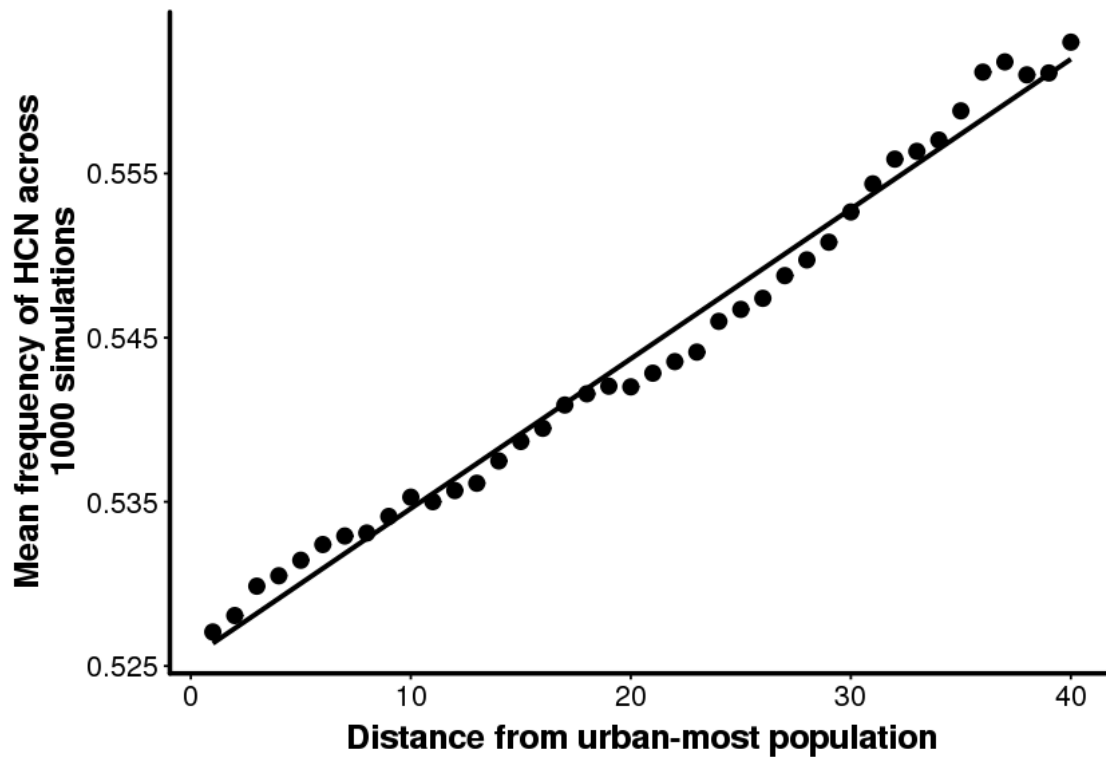
Signif. codes: 0 *** 0.001 ** 0.01 * 0.05 . 0.1 1

Residual standard error: 0.001307 on 38 degrees of freedom

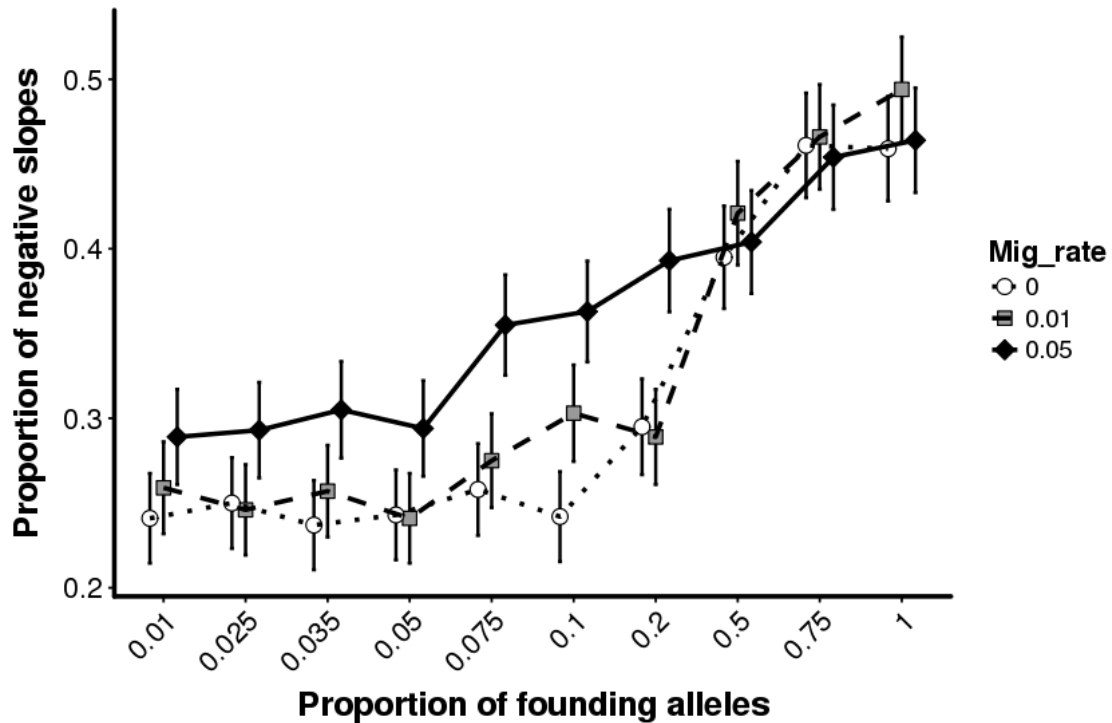
Multiple R-squared: 0.9856, Adjusted R-squared: 0.9852

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)

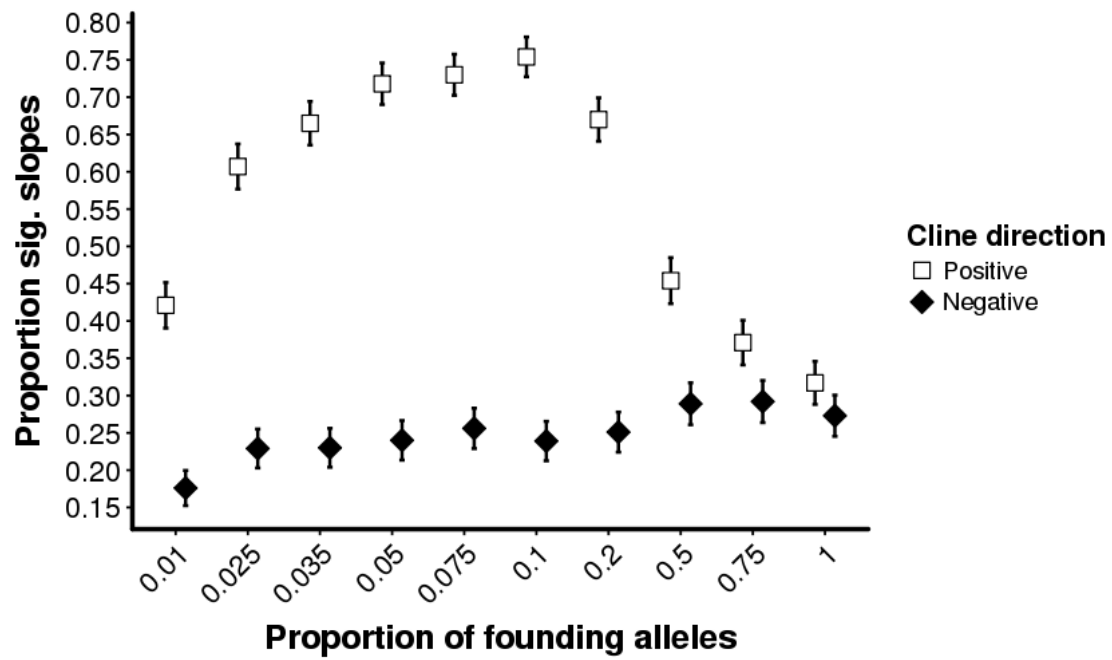
#Melt dataframe
dm1 <- melt(datSlopes_GenOne_NoMig[,c("bot", "prop_sigPos", "ci_sigPos")],
id=c("bot", "ci_sigPos"))
dm2 <- melt(datSlopes_GenOne_NoMig[,c("bot", "prop_sigNeg", "ci_sigNeg")],
id=c("bot", "ci_sigNeg"))

#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
```

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