# Markov Chain Simulation

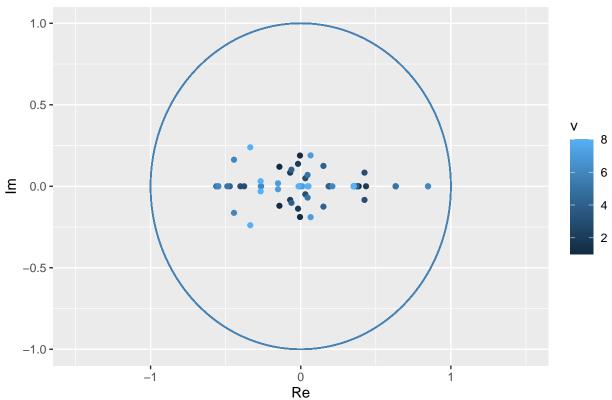
# Generating Random Matrices

```
\# generates rows of size P which are valid probability distributions
r0 <- function(M){
  prob <- runif(M,0,1)</pre>
  prob/sum(prob) # return normalized random row vector
r1 <- function(M){</pre>
  prob <- runif(M,0,1)</pre>
  num_zeros <- sample(1:(M-1),1)</pre>
  choices <- sample(1:M, num_zeros)</pre>
  prob[choices] <- 0</pre>
  prob/sum(prob) # return normalized random row vector
# initialize random P
rand_M <- function(M,row_fxn){</pre>
  P <- matrix(rep(NA, M * M), ncol = M) # create transition matrix
  for(i in 1:M){P[i,] = row_fxn(M)}
}
M <- 8
P <- rand_M(M,r1)</pre>
```

### **Eigenvectors**

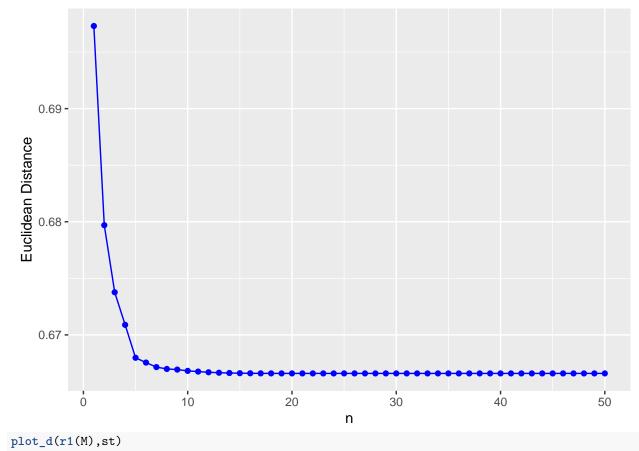
```
eig_P <- eigen(P)</pre>
eig_vectors <- eig_P[2]</pre>
evec <- data.frame(eig_vectors)</pre>
cols \leftarrow 3 \text{ \# set 3 to hold (re,im) pair and whose row it belongs to}
complex <- matrix(rep(NA,cols*M*M), ncol = cols)</pre>
colnames(complex) <- c("Re","Im","v")</pre>
for(i in 1:M){
  for(j in 1:M){
    curr <- evec[i,j]</pre>
    complex[M*(i-1) + j, ] \leftarrow c(Re(curr), Im(curr), i)
  }
}
r < -1
ep <- 0.5
ggplot(complex) +
  geom_point(aes(x = Re, y = Im, color = v)) +
  labs(x = "Re", y = "Im", title = "Distribution of Eigenvectors in the Complex Plane") +
  xlim(-(r+ep),r+ep) + ylim(-r,r) +
  ggforce::geom_circle(aes(x0=0,y0=0,r=r), color = "steelblue")
```

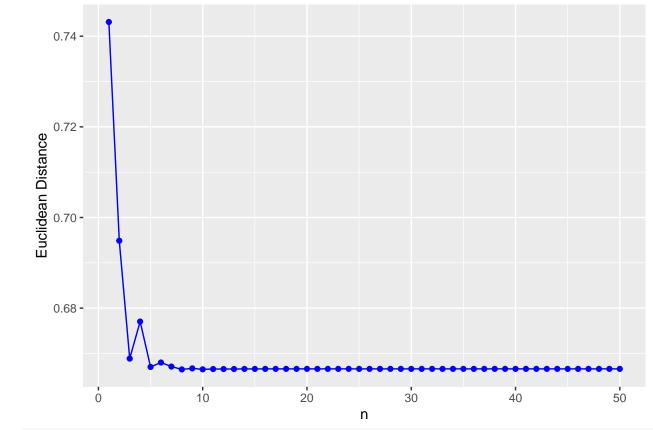
# Distribution of Eigenvectors in the Complex Plane

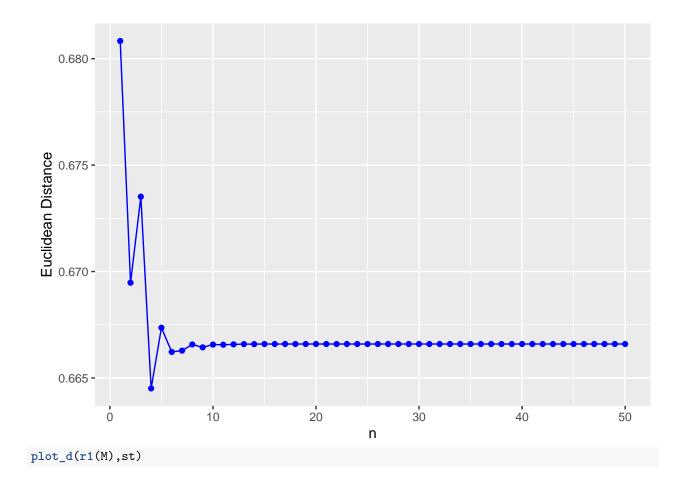


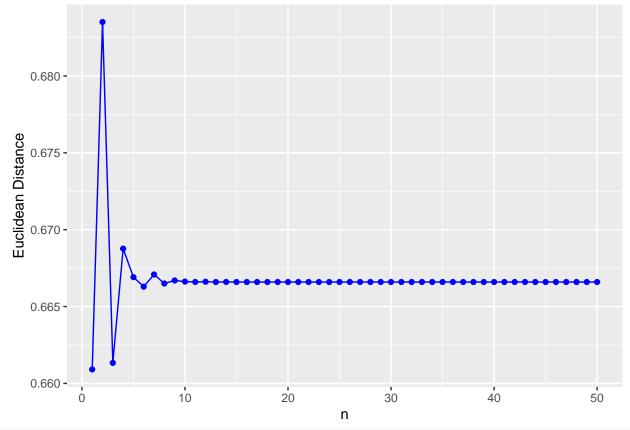
#### Convergence

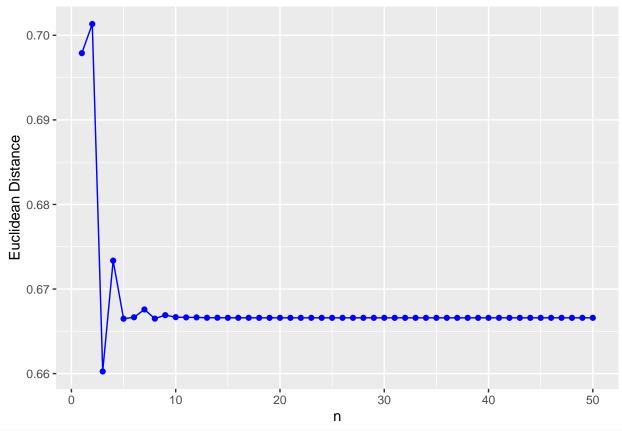
```
set.seed(23)
it <- 50 # set number of iterations of transition matrix
evolve <- function(pi){</pre>
  # simulate and record evolution of pi
  vals <- matrix(rep(NA, (M+1) * it), ncol = (M+1))
  # rename the columns
  str_vec <- rep(NA, M)
  for(i in 1:M){str_vec[i] = paste("x",i,sep="")}
  colnames(vals) <- c("n",str_vec)</pre>
  # evolve pi
  for(i in 1:it){
    vals[i, ] = c(i, pi %*% matrix.power(P,i))
  #store the values in a dataframe
  vals_ <- data.frame(vals) # store indices as base df in case they are needed
  vals <- subset(vals_, select = -c(n))</pre>
  vals
}
piO <- r1(M) # create some initial distribution
st <- evec[1,1] # choose reference vector to find distance from
distance <- function(pi,ref_dist){</pre>
  #plot difference from a reference/stationary distribution
  diff <- rbind(evolve(pi),ref_dist)</pre>
  dist_vec <- rep(0, it)</pre>
  for(i in 1:it){
    curr_dist <- stats::dist(diff[c(i,it+1),], method = "euclidean")</pre>
    dist_vec[i] <- curr_dist</pre>
  data.frame(dist_vec)
plot_d <- function(init,ref){</pre>
 dist_vec <- distance(init,ref)</pre>
  dist_plot <- ggplot(dist_vec, mapping = aes(x = 1:it, y = dist_vec)) +</pre>
    geom_point(color = col_str) + geom_line(color = col_str) +
    labs(x = "n", y = "Euclidean Distance")
  dist_plot
plot_d(r1(M),st)
```











```
col_str <- "red"
plot_d <- function(init,ref){
    dist_vec <- distance(init,ref)
    dist_plot <- ggplot(dist_vec, mapping = aes(x = 1:it, y = dist_vec)) +
        geom_point(color = col_str) + geom_line(color = col_str) +
        labs(x = "n", y = "Euclidean Distance")
    dist_plot
}
plot_d(r0(M),st)</pre>
```

