```
# ... represents all the arguments taken in by the rdist function
RM_explicit <- function(rdist){</pre>
  function(N, ..., symm = FALSE){
    # Create an [N x N] matrix sampling the rows from rdist, passing ... to rdist
    P \leftarrow matrix(rdist(N^2, ...), nrow = N)
    # Make symmetric if prompted
    if(symm){P <- .makeHermitian(P)}</pre>
    # Return P
  }
# A version where we add an imaginary component
RM_explicit_cplx <- function(rdist){</pre>
  RM_dist <- function(N, ..., symm = FALSE, cplx = FALSE, herm = FALSE){</pre>
    # Create an [N \times N] matrix sampling the rows from rdist, passing ... to rdist
    P \leftarrow matrix(rdist(N^2, ...), nrow = N)
    # Make symmetric/hermitian if prompted
    if(symm | herm){P <- .makeHermitian(P)}</pre>
    # Returns a matrix with complex (and hermitian) entries if prompted
    if(cplx){
      # Recursively add imaginary components as 1i * instance of real-valued matrix.
      Im P <- (1i * RM dist(N, ...))</pre>
      # Make imaginary part Hermitian if prompted
      if(herm){P <- P + .makeHermitian(Im P)}</pre>
      else{P \leftarrow P + Im P}
    P # Return the matrix
```

Beta Matrices

```
# Generate a Hermite beta matrix using Dumitriu's Matrix Model
RM_beta <- function(N, beta){
    # Set the diagonal ~ N(0,2)
    P <- diag(rnorm(n = N, mean = 0, sd = sqrt(2)))
    # Get degrees of freedom sequence for offdigonal
    df_seq <- beta * (N - seq(1, N-1))
    # Set the off-1 diagonals as chi squared variables with df(beta_i)
    P[row(P) - col(P) == 1] <- P[row(P) - col(P) == -1] <- sqrt(rchisq(N-1, df_seq))
    # Rescale the entries by 1/sqrt(2)
    P <- P/sqrt(2)
    # Return the beta matrix
    P
}</pre>
```

Implicitly Distributed

Stochastic Matrices

```
# Generates stochastic rows of length N
.stoch_row <- function(N){</pre>
  # Sample a vector of probabilities
 row \leftarrow runif(n = N, min = 0, max = 1)
  # Return the normalized row (sums to one)
 row / sum(row)
# Generates same rows as in .stoch_{row}(N), but with randomly introduced sparsity
.stoch row zeros <- function(N){</pre>
  # Sample a vector of probabilities
 row \leftarrow runif(n = N, min = 0, max = 1)
  # Sample a vertex degree of at least one (as to ensure row is stochastic)
 degree_vertex <- sample(x = 1:(N-1), size = 1)</pre>
  # Sever a random selection of edges to set the vertex degree
 row[sample(1:N, size = N - degree_vertex)] <- 0</pre>
  # Return normalized row
 row / sum(row)
}
RM_stoch <- function(N, symm = F, sparsity = F){</pre>
  # Choose row function depending on sparsity argument
  if(sparsity){row_fxn <- .stoch_row_zeros} else {row_fxn <- .stoch_row}</pre>
  \# Generate the [N x N] stochastic matrix stacking N stochastic rows
 P <- do.call("rbind", lapply(X = rep(N, N), FUN = row_fxn))
  # Make symmetric (if prompted)
 if(symm){ P <- .makeStochSymm(P) }</pre>
  # Return the matrix
}
```

Erdos-Renyi Matrices

```
# Generates a stochastic row with parameterized sparsity of p
.stoch_row_erdos <- function(N, p) {
    # Sample a vector of probabilities
    row <- runif(n = N, min = 0, max = 1)
    # Sample the vertex degree so that it is ~ Bin(n,p)
    degree_vertex <- rbinom(n = 1, size = N, prob = 1 - p)
    # Sever a random selection of edges to set the vertex degree
    row[sample(1:N, degree_vertex)] <- 0
    # Return normalized row only if non-zero (cannot divide by 0)
    if(sum(row) != 0) {
        row / sum(row)
    } else {
            .stoch_row_erdos(N, p) # Otherwise, try again
    }
}</pre>
```

```
RM_erdos <- function(N, p, stoch = T){</pre>
  # Generate an [N x N] Erdos-Renyi stochastic matrix by stacking N p-stochastic rows
 P <- do.call("rbind", lapply(X = rep(N, N), FUN = .stoch_row_erdos, p = p))
  # Return the Erdos-Renyi transition matrix
  Р
}
# Returns a Hermitian version of a matrix by manual assignment
.makeHermitian <- function(P){</pre>
  for(i in 1:nrow(P)){
    for(j in 1:ncol(P)){
      \# Select the entries in the upper triangle (i < j)
      if(i < j){</pre>
        # Make the upper triangle equal to the conjugate transpose of the lower triangle
        P[i,j] <- Conj(P[j,i])</pre>
      }
    }
  }
  # Return the Hermitian Matrix
}
\# Return the off-diagonal entries of row i
.offdiagonalEntries <- function(row, row_index){row[which(1:length(row) != row_index)]}
```

Ensemble Extensions

```
# Extends a RM_dist function to its RME_dist ensemble counterpart
RME_extender <- function(RM_dist){
    # Function returns a list of replicates of the RM_dist function with '...' as arguments
    function(N, ..., size){
        lapply(X = rep(N, size), FUN = RM_dist, ...)
    }
}

RME_unif <- RME_extender(RM_unif)
RME_norm <- RME_extender(RM_norm)
RME_beta <- RME_extender(RM_beta)
RME_stoch <- RME_extender(RM_stoch)
RME_erdos <- RME_extender(RM_erdos)</pre>
```

Spectral Statistics

Spectrum

```
spectrum <- function(array, components = T, norm_order = T, singular = F, order = NA){</pre>
  # Digits to round values to
  digits <- 4
  # Get the type of array
  array_class <- .arrayClass(array)</pre>
  # For ensembles, iteratively rbind() each matrix's spectrum
  if(array_class == "ensemble"){
    map_dfr(array, .spectrum_matrix, components, norm_order, singular, order, digits)
  }
  # From matrices, call the function returning the ordered spectrum for a singleton matrix
  else if(array class == "matrix"){
    .spectrum_matrix(array, components, norm_order, singular, order, digits)
  }
}
# Helper function returning tidied eigenvalue array for a matrix
.spectrum_matrix <- function(P, components, norm_order, singular, order, digits = 4){</pre>
  # For singular values, take P as product of the itself and its tranpose
  if(singular){P <- P %*% t(P)}</pre>
  # Get the eigenvalues of P
  eigenvalues <- eigen(P, only.values = TRUE)$values</pre>
  # Take the square root of the eigenvalues to obtain singular values
  if(singular){eigenvalues <- sqrt(eigenvalues)}</pre>
  # Sort the eigenvalues to make it an ordered spectrum
  eigenvalues <- .sortValues(eigenvalues, norm_order)</pre>
  # If uninitialized, select all orders; otherwise, use c() so singletons => vectors
  if(class(order) == "logical"){order <- 1:nrow(P)} else{order <- c(order)}</pre>
  # Return the spectrum of the matrix
  purrr::map_dfr(order, .resolve_eigenvalue, eigenvalues, components, digits)
# Read and parse an eigenvalue from a sorted eigenvalue array
.resolve_eigenvalue <- function(order, eigenvalues, components, digits){</pre>
  # Read from a sorted eigenvalue array at that order
  eigenvalue <- eigenvalues[order]</pre>
  # Get norm and order columns
  features <- data.frame(Norm = abs(eigenvalue), Order = order)</pre>
  if(components){
    # If components are sought, resolve the eigenvalue into seperate columns first
    res <- cbind(data.frame(Re = Re(eigenvalue), Im = Im(eigenvalue)), features)
  } else{
    # Otherwise, don't resolve the eigenvalue components
    res <- cbind(data.frame(Eigenvalue = eigenvalue), features)
  # Round entries and return the resolved eigenvalue
 res <- round(res, digits)</pre>
  return(res)
```

Helper Functions

```
# Parses an array to see classify it as a matrix or an ensemble of matrices.
.arrayClass <- function(array){</pre>
  # Sample an element from the array and get its class
  elem <- array[[1]]</pre>
  types <- class(elem)</pre>
  # Classify it by analyzing the element class
  if("numeric" %in% types || "complex" %in% types){
    return("matrix")
  else if("matrix" %in% types){
    return("ensemble")
  }
}
# Sort an array of numbers by their norm (written for eigenvalue sorting)
.sortValues <- function(vals, norm_order){</pre>
  values <- data.frame(value = vals)</pre>
  # If asked to sort by norms, arrange by norm and return
  if(norm_order){
    values$norm <- abs(values$value)</pre>
    values <- values %>% arrange(desc(norm))
    # Return the norm-sorted values
    values$value
  # Otherwise, sort by sign and return
  else{ sort(vals, decreasing = TRUE) }
```

Dispersions

```
# Compute the dispersion of a matrix or matrix ensemble
dispersion <- function(array, pairs = NA, norm_order = T, singular = F, pow_norm = 1){</pre>
  # Digits to round values to
 digits <- 4
  # Get the type of array
  array_class <- .arrayClass(array)</pre>
  # Parse input and generate pair scheme (default NA), passing on array for dimension
  pairs <- .parsePairs(pairs, array, array_class)</pre>
  # For ensembles; iteratively rbind() each matrix's dispersion
  if(array_class == "ensemble"){
    map_dfr(array, .dispersion_matrix, pairs, norm_order, singular, pow_norm, digits)
  # Array is a matrix; call function returning dispersion for singleton matrix
  else if(array class == "matrix"){
    .dispersion_matrix(array, pairs, norm_order, singular, pow_norm, digits)
  }
}
# Find the eigenvalue dispersions for a given matrix
.dispersion_matrix <- function(P, pairs, norm_order, singular, pow_norm, digits = 4){</pre>
  # Get the ordered spectrum of the matrix
  eigenvalues <- spectrum(P, norm_order = norm_order, singular = singular)</pre>
  # Generate norm function to pass along as argument (Euclidean or Beta norm)
 norm_fn <- function(x){ (abs(x))^pow_norm }</pre>
  # Compute and return the dispersion
 map2_dfr(pairs[["i"]], pairs[["j"]], .resolve_dispersion, eigenvalues, norm_fn, digits)
}
# Read and parse a dispersion observation between eigenvalue i and j.
.resolve_dispersion <- function(i, j, eigenvalues, norm_fn, digits){</pre>
  # Initialize dispersion dataframe by adding order of eigenvalues compared
  disp \leftarrow data.frame(i = i, j = j)
  # Add the eigenvalues
  disp$eig_i <- .read_eigenvalue(i, eigenvalues)</pre>
  disp$eig_j <- .read_eigenvalue(j, eigenvalues)</pre>
  # Get the identity difference
  disp$id_diff <- disp$eig_j - disp$eig_i</pre>
  # Compute norm of the identity difference (standard norm metric)
  disp$id_diff_norm <- norm_fn(disp$id_diff)</pre>
  # Compute the difference of absolutes
  disp$abs_diff <- norm_fn(disp$eig_j) - norm_fn(disp$eig_i)</pre>
  # Round digits
  disp <- round(disp, digits)</pre>
  # Get the ranking difference
  disp$diff_ij <- disp$i - disp$j</pre>
  # Return the resolved dispersion observation
  disp
}
```

Helper Functions

```
# Parse a string argument for which pairing scheme to utilize
.parsePairs <- function(pairs, array, array_class){</pre>
 # Valid schemes for printing if user is unaware of options
 valid_schemes <- c("largest", "lower", "upper", "consecutive", "all")</pre>
 # Set default to be the consecutive pair scheme
 if(class(pairs) == "logical"){pairs <- "consecutive"}</pre>
 # Stop function call if the argument is invalid
 if(!(pairs %in% valid_schemes)){
   scheme_list <- paste(valid_schemes, collapse = ", ")</pre>
   stop(paste("Invalid pair scheme. Try one of the following: ", scheme_list, ".", ""))
 # // Once we verify that we have a valid pair scheme string, try to parse it.
 # First, obtain a matrix by inferring array type; if ensemble take first matrix
 if(array class == "ensemble") { P <- array[[1]] }</pre>
 else if(array_class == "matrix") { P <- array }</pre>
 # Obtain the dimension of the matrix
 N \leftarrow nrow(P)
 # Parse the pair string and evaluate the pair scheme
 if(pairs == "largest"){pair_scheme <- data.frame(i = 2, j = 1)}</pre>
 else if(pairs == "consecutive"){pair_scheme <- .consecutive_pairs(N)}</pre>
 else if(pairs == "lower"){pair_scheme <- .unique_pairs_lower(N)}</pre>
 else if(pairs == "upper"){pair_scheme <- .unique_pairs_upper(N)}</pre>
 else if(pairs == "all"){pair_scheme <- .all_pairs(N)}</pre>
 # Return pair scheme
 return(pair_scheme)
```

Pairing Schema

```
#-----#
# The trivial pairing scheme:
# Enumerate all possible pairs.
.all_pairs <- function(N){</pre>
 purrr::map_dfr(1:N, function(i, N){data.frame(i = rep(i, N), j = 1:N)}, N)
# The consecutive pairing scheme:
# Enumerate all possible consecutive/neighboring pairs. Ensures no linear combiantions.
.consecutive_pairs <- function(N){</pre>
 purrr::map_dfr(2:N, function(i){data.frame(i = i, j = as.integer(i - 1))})
#-----#
# The lower-triangular pairing scheme:
# Enumerate the pair combinations given N items with i > j.
.unique_pairs_lower <- function(N){</pre>
 is <- do.call("c", purrr::map(1:N, function(i){rep(i,N)}))</pre>
 js <- rep(1:N, N)
 # Helper function: selects elements only if they are lower triangular
 .isLowerTri \leftarrow function(i, j){if(i > j){ c(i = i, j = j) }}
 pairs <- do.call("rbind",purrr::map2(is, js, .f = .isLowerTri))</pre>
 data.frame(pairs)
}
               -----
# The upper-triangular pairing scheme:
# Enumerate the pair combinations given N items with i < j.
.unique_pairs_upper <- function(N){</pre>
 is <- do.call("c", purrr::map(1:N, function(i){rep(i,N)}))</pre>
 js <- rep(1:N, N)
 # Helper function: selects elements only if they are lower triangular
 .isUpperTri <- function(i, j){if(i < j){ c(i = i, j = j) }}
 pairs <- do.call("rbind",purrr::map2(is, js, .f = .isUpperTri))</pre>
 data.frame(pairs)
```

Parallel Extensions

Spectrum

```
spectrum_parallel <- function(array, components = TRUE, sort_norms = TRUE, singular = FALSE, order = NA
    digits <- 4 # Digits to round values to
    # Array is a matrix; call function returning eigenvalues for singleton matrix
    if(class(array) == "matrix"){
        .spectrum_matrix(array, components, sort_norms, singular, order, digits)
    }
    # Array is an ensemble; recursively row binding each matrix's eigenvalues
    else if(class(array) == "list"){
        furrr::future_map_dfr(array, .spectrum_matrix, components, sort_norms, singular, order, digits)
}</pre>
```

Dispersion

```
dispersion_parallel <- function(array, pairs = NA, sort_norms = TRUE, singular = FALSE, norm_pow = 1){
    digits <- 4 # Digits to round values to
    pairs <- .parsePairs(pairs, array) # Parse input and generate pair scheme (default NA), passing on ar
    # Array is a matrix; call function returning dispersion for singleton matrix
    if(class(array) == "matrix"){
        .dispersion_matrix(array, pairs, sort_norms, singular, norm_pow, digits)
    }
    # Array is an ensemble; recursively row binding each matrix's dispersions
    else if(class(array) == "list"){
        furrr::future_map_dfr(array, .dispersion_matrix, pairs, sort_norms, singular, norm_pow, digits)
    }
}</pre>
```