## R SCRIPT

# various utility functions for parameter recovery routines

# density function for Wald (unshifted) dwald = function(x, gamma, alpha) { return( $(alpha/(sqrt(2*pi*x^3)))*exp(-(alpha-gamma*x)^2/(2*x)))$ # function to generate random shifted Wald data # adapted from pp. 79-80, Dagpunar, J. (1988). Principles of Random Variate Generation. Clarendon Press, Oxford. # code modified from Heathcote (2004) rwald = function(n, gamma, alpha, theta) { y2 = rchisq(n, 1)y2onm = y2/gammau = runif(n)r1 = (2\*alpha + y2onm - sqrt(y2onm\*(4\*alpha+y2onm)))/(2\*gamma) $r2 = (a1pha/gamma)^2/r1$ ifelse(u < alpha/(alpha+gamma\*r1), theta+r1, theta+r2) # basic function that returns Root Mean Squared Error rmse = function(error) { sgrt (mean (error^2)) } # negative log likelihood for shifted Wald nll. wald = function (par, dat) { return(-sum(log(dwald(dat-par[3], gamma=par[1], alpha=par[2])))) # Start point estimate for SW, based on first two moments # assumes s = p\*min(x), where x is a data vector # from Heathcote (2004) waldinit = function (x, p = 0.9) { theta = p\*min(x)x = x - thetagamma = sqrt(mean(x)/var(x))alpha = gamma\*mean(x)

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return (c (gamma, alpha, theta))
#Running simulations for Fit Functions
# define simulation parameters
nSub = 20
nTrials = 20
# parent distribution means and sds (from F, Vick, Bowman, 2018)
G = 3.91 \# gamma
G. sd = 0.70
A = 0.92 \# alpha
A. sd = 0.17
H = 0.32 \# theta
H. sd = 0.05
# each simulation run starts HERE
# build simulated RT matrix
# rows = subjects, columns = trials
rts = matrix(0, nrow=nSub, ncol=nTrials)
# build matrix to store "target" parameters
# rows = subjects, columns = gamma, alpha, theta
targets = matrix(0, nrow=nSub, ncol=3)
# randomly draw target shifted Wald parameters from parent distribution
# (one unique parameter value for each subject)
for (i in 1:nSub) {
  targets[i,1] = rnorm(1, mean=G, sd=G.sd) # draw random gamma
  targets[i,2] = rnorm(1, mean=A, sd=A.sd) # draw random alpha
  targets[i,3] = rnorm(1, mean=H, sd=H.sd) # draw random theta
# from these target SW values, generate the distribution of observed RTs
for each subject
for (i in 1:nSub) {
  rts[i,] = rwald(nTrials,
                  gamma = targets[i, 1],
                  alpha = targets[i, 2],
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theta = targets[i, 3])

# now let's use CMLE to fit shifted Wald distribution to each row of observed RTs
# first, we'll store these estimates in a matrix
predictions = matrix(0, nrow=nSub, ncol=3)

# now do the fits (one per subject)
for (i in 1:nSub) {
   fit = optim(waldinit(rts[i,]), nll.wald, dat=rts[i,])
   predictions[i,] = fit$par
}
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