Computational Approach to Estimating AR(1) Process with SMM:

1. Housekeeping:

- Initialize the algorithm. Set appropriate parameters (true values of ρ and σ , T, H).
- Simulate data under "true" parameters. Set a seed. Using true values of ρ and σ , simulate a time series of data of length T. Calculate moments from the time series. These are your "data" moments.
- Draw and store a set of shocks that are of H vectors of length T. Just like the Krusell-Smith problem set, we want to make sure we are using the *same* shocks when we simulate our model.
- 2. 1st Step Estimate of Parameters: The parameters that minimize your objective function J are your first step estimates.
 - Set the weight matrix to be the identity matrix.
 - Define an objective function J to minimize. This will be a function of your parameters that compares simulated moments to the "data" moments computed in (1).
 - Calculate standard errors.

3. 2nd Step Estimate of Parameters:

- Calculate optimal weight matrix using Newey-West.
- Minimize the same objective function defined in (2) using the new weight matrix.
- Calculate standard errors. Confirm that the standard are smaller under the optimal weight matrix.

Algorithm 1 Estimate AR(1) Model via SMM

- 1: procedure SMM
- 2: call GetMoments($\rho_0, \sigma_0, T, H = 1$)
- return $m(\rho_0, \sigma_0)$ 3:

return $\hat{\rho}_1, \hat{\sigma}_1$

W = I4:

6:

call MinObjFunc $(W, \tilde{\rho}, \tilde{\sigma}, m(\rho_0, \sigma_0))$ 5:

- $\triangleright \{\tilde{\rho}, \tilde{\sigma}\}$ is some initial guess
 - ⊳ First Step Estimator

- call ComputeSE $(\hat{\rho}_1, \hat{\sigma}_1)$ 7:
- return $diag(\Sigma_1(\hat{\rho}_1, \hat{\sigma}_1))$ 8:

▷ See equation (8)

- call NeweyWest $(m(\rho_0, \sigma_0), m(\hat{\rho}_1, \hat{\sigma}_1))$ 9:
- return S10:
- $W^* = S^{-1}$ 11:
- call MinObjFunc($W^*, \tilde{\rho}, \tilde{\sigma}, m(\rho_0, \sigma_0)$) 12:
- return $\hat{\rho}_2, \hat{\sigma}_2$ 13:

 \triangleright Second Step Estimator

- call ComputeSE $(\hat{\rho}_2, \hat{\sigma}_2)$ 14:
- return $diag(\Sigma_2(\hat{\rho}_2, \hat{\sigma}_2))$ 15:
- 16: end procedure

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function GetMoments(\rho, \sigma, T, H)
     - Simulate H AR(1) processes of length T.
     - Compute appropriate moments for each of the T simulations.
     - Average across H individual moments: m(\rho, \sigma) = 1/H \sum_{H} m_h(\rho, \sigma)
     return m(\rho, \sigma)
end function
function MinObjFunc(W, \hat{\rho}, \hat{\sigma}, m(\rho_0, \sigma_0))
      minimize ObjFunc(W, \hat{\rho}, \hat{\sigma}, m(\rho_0, \sigma_0))
     return \{\hat{\rho}, \hat{\sigma}\}
end function
function ObjFunc(W, \hat{\rho}, \hat{\sigma}, m(\rho_0, \sigma_0))
      GETMOMENTS(\hat{\rho}, \hat{\sigma}, T, H) return m(\hat{\rho}, \hat{\sigma})
     J = [m(\hat{\rho}, \hat{\sigma}) - m(\rho_0, \sigma_0)]'W[m(\hat{\rho}, \hat{\sigma}) - m(\rho_0, \sigma_0)]
     return J
end function
function ComputeSE(\hat{\rho}, \hat{\sigma})
      \varepsilon = 1e^{-10}
      GETMOMENTS(\hat{\rho}, \hat{\sigma}, T, H) return m(\hat{\rho}, \hat{\sigma})
      GETMOMENTS(\hat{\rho} - \varepsilon, \hat{\sigma}, T, H) return m(\hat{\rho} - \varepsilon, \hat{\sigma})
      GETMOMENTS(\hat{\rho}, \hat{\sigma} - \varepsilon, T, H) return m(\hat{\rho}, \hat{\sigma} - \varepsilon)
      \nabla_{\rho} = [m(\hat{\rho} - \varepsilon, \hat{\sigma}) - m(\hat{\rho}, \hat{\sigma})]/\varepsilon and \nabla_{\sigma} = [m(\hat{\rho}, \hat{\sigma} - \varepsilon) - m(\hat{\rho}, \hat{\sigma})]/\varepsilon
      \nabla_b = [\nabla_\rho, \nabla_\sigma]
     return diag(\Sigma(\hat{\rho}, \hat{\sigma}))
                                                                                                                     \triangleright See equation (8)
end function
function NeweyWest(m(\rho_0, \sigma_0), m(\hat{\rho}, \hat{\sigma}))
     See HelpfulFunction.jl
     return S
end function
```