Contents

- Step 1: Import CSV Data
- Step 2: R Bridge Implementation
- Step 3a: MDFT Formulation to Calculate Preference Dynamics in Parallel
- Step 3b: MDFT Formulation with State Continuity
- Step 4: Output Results
- Helper Functions

Step 1: Import CSV Data

(reference apolloMain_5 amd apolloMain_6 as example for data manipulation) biasData = readtable('user_choices.csv'); % Replace with the path to your data file disp('User bias data imported successfully.'); taskChoice_Data = readtable('user_choices.csv'); % Replace with the path to your data file disp('User task choice data imported successfully.');

```
robotChoice_Data = readtable('G:\My Drive\myResearch\Research Experimentation\Apollo\apollo\data\Bounding_Overwatch_Data\HumanData_Bounding_Overwatch - 20Split.csv'
% Convert all column headers to lowercase
robotChoice_Data.Properties.VariableNames = lower(robotChoice_Data.Properties.VariableNames);
disp('User robot choice data imported successfully.');
\% Randomly select 10 rows (or all rows if fewer than 10)
numRows = height(robotChoice_Data);
randomIndices = randperm(numRows, min(10, numRows));
robotChoice_Data = robotChoice_Data(randomIndices, :);
% Extract robot state attributes dynamically
robot states = struct();
attributeSuffixes = {'traversability', 'visibility'}; % No leading underscores
for i = 1:3
    for attr = attributeSuffixes
        csvColName = sprintf('robot%d %s', i, attr{1});  % Matches CSV column names
        structFieldName = attr{1};  % Valid field name
        if ismember(csvColName, robotChoice_Data.Properties.VariableNames)
            robot_states.(['robot' num2str(i)]).(structFieldName) = robotChoice_Data.(csvColName);
            warning('Missing attribute column: %s', csvColName);
            robot_states.(['robot' num2str(i)]).(structFieldName) = NaN(height(robotChoice_Data), 1);
        end
    end
% Extract choice data and other metadata
choices = robotChoice_Data.choice;
participant_ids = robotChoice_Data.id;
stake_types = robotChoice_Data.stakes;
time_spent = robotChoice_Data.timeelapsed;
```

User robot choice data imported successfully.

Step 2: R Bridge Implementation

```
disp('Initializing R bridge...');

% Configure paths
rscript_path = 'C:\Program Files\R\R-4.4.2\bin\x64\Rscript.exe';
r_script = 'G:\My Drive\myResearch\Research Experimentation\Apollo\apollo\example\DFT_Bounding_Overwatch.R';
csvFile = 'G:\My Drive\myResearch\Research Experimentation\Apollo\apollo\data\Bounding_Overwatch_Data\HumanData_Bounding_Overwatch - 80Split.csv';
outputDir = 'G:\My Drive\myResearch\Research Experimentation\Apollo\apollo\output_BoundingOverwatch';

% Verify installations
if ~isfile(rscript_path)
    error('Rscript_exe not found at: %s', rscript_path);
elseif ~isfile(r_script)
    error('R script not found at: %s', r_script);
elseif ~isfile(csvFile)
    error('Input CSV not found at: %s', csvFile);
```

```
elseif ~isfolder(outputDir)
    warning('Output folder does not exist, creating: %s', outputDir);
% Execute R with JSON output
    % Use proper argument formatting
    cmd = sprintf(['"%s" "%s" ', ...
'-i "%s" -o "%s"'], ...
               rscript_path, r_script, csvFile, outputDir);
[status,result] = system(cmd);
    if status == 0
        % Handle output path (whether directory or file)
        if isfolder(outputDir)
            jsonFile = fullfile(outputDir, 'DFT_output.json');
           jsonFile = outputDir;
        end
        % Parse JSON output
        if exist(jsonFile, 'file')
            jsonText = fileread(jsonFile);
            params = jsondecode(jsonText);
            \% Extract parameters with validation
            %Boundedphi1, phi2 parameters
            phi1 = min(max(0, validateParam(params, 'phi1', 0.5)),5); % Ensure non-negative
            phi2 = min(max(0, validateParam(params, 'phi2', 0.8)), 0.99); % Constrain 0-1
            %tau = min(1 + exp(validateParam(params, 'timesteps', 0.5)),100); %Constrain to 100
            %Raw phi1, phi2 parameters
            %phi1 = validateParam(params, 'phi1', 0.5);
            %phi2 = validateParam(params, 'phi2', 0.8);
            tau = 1 + exp(validateParam(params, 'timesteps', 0.5));
            error_sd = min(max(0.1, validateParam(params, 'error_sd', 0.1)), 1); % still clip here
            % Extract attribute weights
            beta_weights = [
                params.b_attr1;
                params.b_attr2;
                params.b_attr3;
                params.b_attr4
            ];
            % Get initial preferences from ASCs
            initial P = [
                validateParam(params, 'asc_1', 0);
                validateParam(params, 'asc_2', 0);
                validateParam(params, 'asc_3', 0);
            disp('Estimated Parameters:');
            disp(['phi1: ', num2str(phi1)]);
            disp(['phi2: ', num2str(phi2)]);
            disp(['tau: ', num2str(tau)]);
            disp(['error_sd: ', num2str(error_sd)]);
            disp('Initial Preferences (from ASCs):');
            disp(initial_P');
            error('R output file not found');
    else
        error('R execution failed: %s', result);
    end
    disp('Error during R execution:');
    disp(getReport(ME, 'extended'));
    [phi1, phi2, tau, error_sd] = getFallbackParams();
    beta_weights = [0.3; 0.2; 0.4; 0.5]; % Default weights
    initial_P = zeros(3,1); % Neutral initial preferences
```

Initializing R bridge...

Step 3a: MDFT Formulation to Calculate Preference Dynamics in Parallel

```
%%{
% (MDFT calculations based on estimated parameters)
% Create M matrix from current trial's attributes
% C11-C14 are consequence attributes for Robot 1
% C21-C24 are consequence attributes for Robot 2
```

```
% C31-C34 are consequence attributes for Robot 3
for current_trial = 1:height(robotChoice_Data)
      num_attributes = 4;
       M = [
             robotChoice_Data.c11(current_trial), robotChoice_Data.c12(current_trial), robotChoice_Data.c13(current_trial), robotChoice_Data.c14(current_trial);
             robotChoice\_Data.c21(current\_trial), \ robotChoice\_Data.c22(current\_trial), \ robotChoice\_Data.c23(current\_trial); \\ robotChoice\_Data.c24(current\_trial); \\ robotChoice\_Data
              robotChoice_Data.c31(current_trial), robotChoice_Data.c32(current_trial), robotChoice_Data.c33(current_trial), robotChoice_Data.c34(current_trial)
      % Normalize M values by dividing by 2 and clamping to [0.01, 1]
       M = M / 2:
       M = max(0.01, min(1, M));
      %}
       % --- Global Max Normalization ---
       global_max = max(robotChoice_Data{:, {'c11','c12','c13','c14','c21','c22','c23','c24','c31','c32','c33','c34'}}, [], 'all', 'omitnan');
      if ~isfinite(global_max) || global_max <= 0</pre>
            global_max = 1; % fallback in case of zero or NaN
      M = M / global_max;
                                                              % Normalize by global max
      M = max(0.01, min(1, M));
                                                             % Clamp to [0.01, 1]
       % --- Row-wise Min-Max Normalization ---
       for i = 1:size(M, 1)
             row = M(i, :);
             min_val = min(row);
             max_val = max(row);
              if max val == min val
                   M(i, :) = pmax(0.01, pmin(1, row)); % constant row: clamp only
                    norm_row = (row - min_val) / (max_val - min_val);
                    M(i, :) = max(0.01, min(1, norm_row)); % clamp to [0.01, 1]
       end
       %%}
       attributes = {'C1 - Easy Nav, Low Exposure', 'C2 - Hard Nav, Low Exposure', 'C3 - Easy Nav, High Exposure', 'C4 - Hard Nav, High Exposure'};
       beta = beta_weights ./ sum(abs(beta_weights));
       beta = beta';
       [E_P, V_P, choice_probs, P_tau] = calculateDFTdynamics(...
             phi1, phi2, tau, error_sd, beta, M, initial_P);
       % Display results for the frame
       disp('=== Trial Analysis ===');
       disp(['Trial: ', num2str(current_trial)]);
       disp(['Participant: ', num2str(participant_ids(current_trial))]);
       disp(['Actual Choice: Robot ', num2str(choices(current_trial))]);
       disp('M matrix (alternatives × attributes):');
       disp(arrav2table(M. ...
              'RowNames', {'Robot1', 'Robot2', 'Robot3'}, ...
              'VariableNames', attributes));
       disp('DFT Results:');
       disp(['E_P: ', num2str(E_P', '%.2f ')]);
       disp(['Choice probabilities: ', num2str(choice_probs', '%.3f ')]);
       [~, predicted_choice] = max(choice_probs);
       disp(['Predicted choice: Robot ', num2str(predicted_choice)]);
       disp(['Actual choice: Robot ', num2str(choices(current_trial))]);
      disp(' ');
       if predicted_choice == choices(current_trial)
             disp('√ Prediction matches actual choice');
             disp('X Prediction differs from actual choice');
       end
      % Plot evolution
       figure;
       %plot(0:tau, P_tau);
       % Replace the plotting section with:
       tau_rounded = round(tau); % Ensure integer steps
       if size(P_tau,2) == tau_rounded+1 % Validate dimensions
             plot(0:tau_rounded, P_tau);
             warning('Dimension mismatch: P_tau has %d cols, expected %d',...
                           size(P_tau,2), tau_rounded+1);
```

```
plot(P_tau'); % Fallback plot
    end
    xlabel('Preference Step (\tau)');
   ylabel('Preference Strength');
    legend({'Robot1','Robot2','Robot3'});
    title(sprintf('Preference Evolution (Trial %d)', current_trial));
    grid on;
%%}
=== Trial Analysis ===
Trial: 1
Participant: 141831
Actual Choice: Robot 3
M matrix (alternatives \times attributes):
             C1 - Easy Nav, Low Exposure
                                          C2 - Hard Nav, Low Exposure
                                                                          C3 - Easy Nav, High Exposure
                                                                                                        C4 - Hard Nav, High Exposure
   Robot1
                          1
                                                      0.57827
                                                                                     0.42173
                                                                                                                      0.01
   Robot2
                          1
                                                       0.6312
                                                                                     0.3688
                                                                                                                      0.01
    Robot3
                          1
                                                      0.54961
                                                                                     0.45039
                                                                                                                      0.01
DFT Results:
E P: 0.00 0.02 -0.02
Choice probabilities: 0.339 0.392 0.269
Predicted choice: Robot 2
Actual choice: Robot 3
X Prediction differs from actual choice
=== Trial Analysis ===
Trial: 2
Participant: 141831
Actual Choice: Robot 2
M matrix (alternatives × attributes):
             C1 - Easy Nav, Low Exposure C2 - Hard Nav, Low Exposure
                                                                          C3 - Easy Nav, High Exposure C4 - Hard Nav, High Exposure
                                                      0.44957
                                                                                     0.55043
                                                                                                                       0.01
    Robot1
    Robot2
                                                       0.5533
                                                                                      0.4467
                                                                                                                       0.01
                          1
                                                      0.50618
                                                                                     0.49382
   Robot3
                                                                                                                      0.01
                          1
DFT Results:
E_P: -0.01 0.02 -0.01
Choice probabilities: 0.301 0.399 0.300
Predicted choice: Robot 2
Actual choice: Robot 2
\checkmark Prediction matches actual choice
=== Trial Analysis ===
Participant: 125802
Actual Choice: Robot 3
M matrix (alternatives × attributes):
             C1 - Easy Nav, Low Exposure C2 - Hard Nav, Low Exposure C3 - Easy Nav, High Exposure C4 - Hard Nav, High Exposure
   Robot1
                          1
                                                      0.49468
                                                                                     0.50532
                                                                                                                      0.01
   Robot2
                          1
                                                      0.46147
                                                                                     0.53853
                                                                                                                      0.01
    Robot3
                          1
                                                      0.52825
                                                                                     0.47175
                                                                                                                      0.01
DFT Results:
E P: 0.01 -0.00 -0.00
Choice probabilities: 0.350 0.321 0.329
Predicted choice: Robot 1
Actual choice: Robot 3
X Prediction differs from actual choice
=== Trial Analysis ===
Trial: 4
Participant: 141831
Actual Choice: Robot 3
M matrix (alternatives × attributes):
                                                                                                         C4 - Hard Nav, High Exposure
             C1 - Easy Nav, Low Exposure
                                            C2 - Hard Nav, Low Exposure
                                                                          C3 - Easy Nav, High Exposure
                                                      0.50762
                                                                                     0.49238
                                                                                                                       0.01
    Robot1
                          1
   Robot2
                                                      0.54112
                                                                                     0.45888
                                                                                                                      0.01
                          1
   Robot3
                          1
                                                      0.52591
                                                                                     0.47409
                                                                                                                      0.01
DFT Results:
E_P: 0.00 0.01 -0.01
Choice probabilities: 0.333 0.366 0.301
Predicted choice: Robot 2
```

Predicted choice: Robot 2
Actual choice: Robot 3

X Prediction differs from actual choice

=== Trial Analysis ===

Trial: 5

1

Participant: 125802

Participant: Actual Choice				
	ternatives × attributes):			
	C1 - Easy Nav, Low Exposure	C2 - Hard Nav, Low Exposure	C3 - Easy Nav, High Exposure	C4 - Hard Nav, High Exposure
Robot1 Robot2	1 1	0.55138 0.48487	0.44862 0.51513	0.01 0.01
Robot2	1	0.50079	0.49921	0.01
DFT Results: E_P: 0.02 -0	0.00 -0.01			
_	oilities: 0.386 0.325 0.289			
	pice: Robot 1			
Actual choice	E: KODOT 3			
	differs from actual choice			
=== Trial Ana	alysis ===			
Trial: 6 Participant:	141831			
Actual Choice				
M matrix (al	ternatives × attributes):	62 Hand New Levy Symposius	62 Farry New Hitch Francisco	CA Hand New High Europe
	C1 - Easy Nav, Low Exposure	C2 - Hard Nav, Low Exposure	C3 - Easy Nav, High Exposure	C4 - Hard Nav, High Exposure
Robot1	1	0.50662	0.49338	0.01
Robot2 Robot3	1	0.50436 0.53586	0.49564 0.46414	0.01 0.01
Noocs	-	0.33300	0.40424	0.01
DFT Results:				
_	0.00 -0.00 Dilities: 0.342 0.341 0.317			
	pice: Robot 1			
Actual choice	e: Robot 3			
Y Prediction	differs from actual choice			
=== Trial Ana				
Trial: 7				
Participant: Actual Choice				
	ternatives × attributes):			
	C1 - Easy Nav, Low Exposure	C2 - Hard Nav, Low Exposure	C3 - Easy Nav, High Exposure	C4 - Hard Nav, High Exposure
Robot1	1	0.51739	0.48261	0.01
Robot2	1	0.48096	0.51904	0.01
Robot3	1	0.47196	0.52804	0.01
DFT Results:				
E_P: 0.01				
	oilities: 0.374 0.341 0.285 Dice: Robot 1			
Actual choice				
✓ Prediction === Trial Ana	matches actual choice			
Trial: 8	arysis ===			
Participant:	125802			
Actual Choice				
m matrix (all	ternatives × attributes): C1 - Easy Nav, Low Exposure	C2 - Hard Nav, Low Exposure	C3 - Easy Nav, High Exposure	C4 - Hard Nav, High Exposure
Robot1	1	0.52664	0.47336	0.01
Robot1	1	0.54328	0.45672	0.01
Robot3	1	0.50397	0.49603	0.01
DFT Results:				
E_P: 0.01 (0.01 -0.01			
Choice probab	oilities: 0.350 0.368 0.283			
Predicted cho Actual choice	pice: Robot 2			
Actual choice	e: Kobot 1			
X Prediction	differs from actual choice			
=== Trial Ana	alysis ===			
Trial: 9 Participant:	125802			
Actual Choice				
M matrix (al	ternatives × attributes):			
	C1 - Easy Nav, Low Exposure	C2 - Hard Nav, Low Exposure	C3 - Easy Nav, High Exposure	C4 - Hard Nav, High Exposure
Robot1	1	0.50479	0.49521	0.01
Robot2	1	0.55569 0.48468	0.44431 0.51532	0.01 0.01

0.51532

DFT Results:

E_P: 0.00 0.02 -0.02

Choice probabilities: 0.337 0.388 0.274

Predicted choice: Robot 2 Actual choice: Robot 3

 ${\sf X}$ Prediction differs from actual choice

=== Trial Analysis ===

Trial: 10

Participant: 125802 Actual Choice: Robot 1

M matrix (alternatives \times attributes):

C1 - Easy Nav, Low Exposure

C2 - Hard Nav, Low Exposure C3 - Easy Nav, High Exposure C4 - Hard Nav, High Exposure

Robot1	1	0.40645	0.59355	0.01
Robot2	1	0.48832	0.51168	0.01
Robot3	1	0.47183	0.52817	0.01

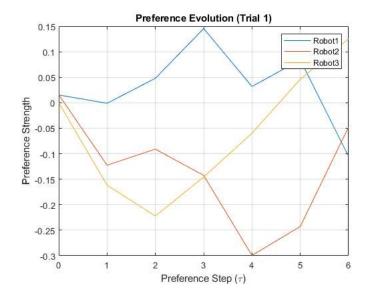
DFT Results:

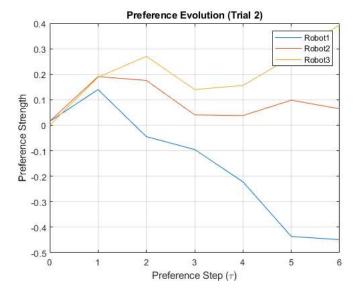
E_P: -0.01 0.02 -0.00

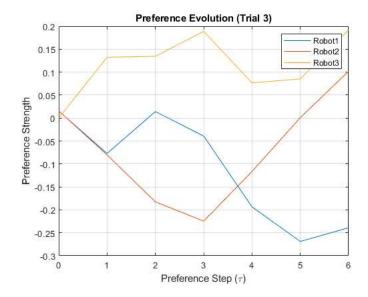
Choice probabilities: 0.305 0.382 0.312

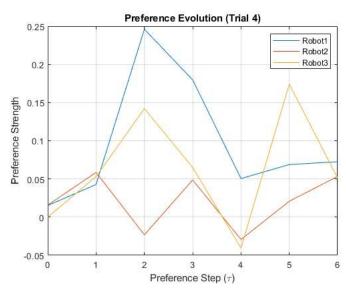
Predicted choice: Robot 2 Actual choice: Robot 1

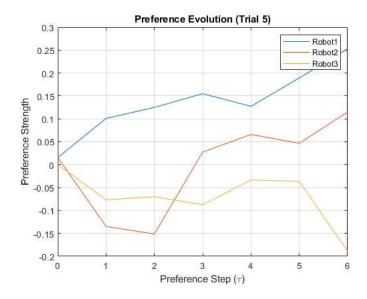
X Prediction differs from actual choice

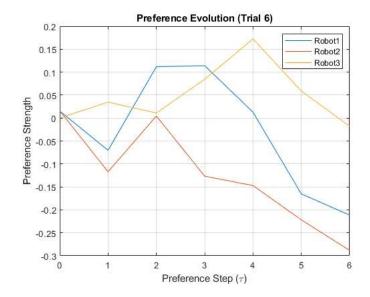


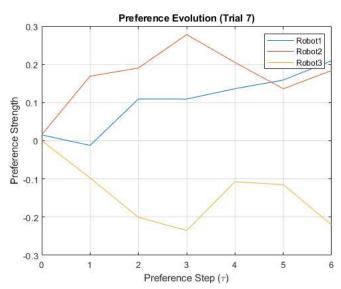


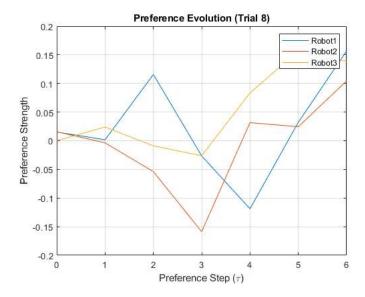


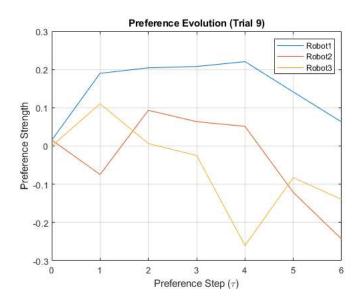


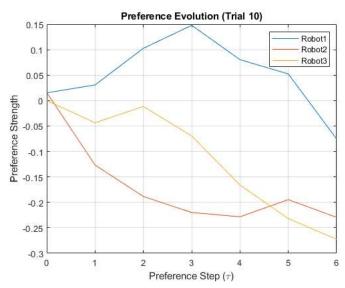












Step 3b: MDFT Formulation with State Continuity

```
% Initialize preference state tracking
if ~exist('P_final_prev', 'var')
         P_final_prev = initial_P; % Use estimated initial preferences for first trial
end
 for current_trial = 1:height(robotChoice_Data)
         % Create M matrix for current trial
         M = [
                   robotChoice_Data.c11(current_trial), robotChoice_Data.c12(current_trial), robotChoice_Data.c13(current_trial), robotChoice_Data.c14(current_trial);
                   robotChoice\_Data.c21(current\_trial), \ robotChoice\_Data.c22(current\_trial), \ robotChoice\_Data.c23(current\_trial); \\ robotChoice\_Data.c24(current\_trial); \\ robotChoice\_Data.c24(current\_trial); \\ robotChoice\_Data.c23(current\_trial); \\ robotChoice\_Data.c24(current\_trial); \\ robotChoice\_Data
                    robotChoice_Data.c31(current_trial), robotChoice_Data.c32(current_trial), robotChoice_Data.c33(current_trial), robotChoice_Data.c34(current_trial)
         1:
         % Normalize beta weights
         beta = beta_weights ./ sum(abs(beta_weights));
          % Calculate DFT dynamics using previous trial's final state
         [E_P, V_P, choice_probs, P_tau] = calculateDFTdynamics(...
                   phi1, phi2, tau, error_sd, beta, M, P_final_prev);
         \ensuremath{\text{\%}} Store final preference state for next trial
          P_final_prev = P_tau(:, end);
         % Display results
         disp('=== Trial Analysis ===');
          disp(['Trial: ', num2str(current_trial)]);
          disp(['Participant: ', num2str(participant_ids(current_trial))]);
          disp(['Actual Choice: Robot ', num2str(choices(current_trial))]);
          disp('Initial Preferences (from previous trial):');
```

```
disp('Final Preferences:');
                 \label{lem:disp(array2table(P_tau(:,end)', 'VariableNames', {'Robot1', 'Robot2', 'Robot3'})); \% \  \  \, Fixed \  \, this \  \, line \  \,
                 % Enhanced plotting with initial/final state markers
                 figure;
                 plot(0:tau, P_tau, 'LineWidth', 2);
                  hold on;
                 % Mark initial state
                 scatter(zeros(3,1), P_tau(:,1), 100, 'filled');
                 % Mark final state
                 scatter(tau*ones(3,1), P_tau(:,end), 100, 'x', 'LineWidth', 2);
                 xlabel('Preference Step (\tau)');
                 ylabel('Preference Strength');
                 legend({'Robot1','Robot2','Robot3','Initial State','Final State'});
                  title(sprintf('Preference Evolution (Trial %d)', current_trial));
                 grid on;
%}
```

Step 4: Output Results

```
Saving results to CSV...

Error using table
All table variables must have the same number of rows.

Error in main_BoundingOverwatch (line 302)
output_table = table(E_P, V_P, P_tau(end,:)', ...
```

Helper Functions

```
function param = validateParam(params, name, default)
    if isfield(params, name) && isnumeric(params.(name))
        param = params.(name);
    else
        warning('Using default for %s', name);
        param = default;
    end
end

function [phi1, phi2, tau, error_sd] = getFallbackParams()
    phi1 = 0.5 + 0.1*randn();
    phi2 = 0.8 + 0.1*randn();
    tau = 10 + randi(5);
    error_sd = 0.1 + 0.05*rand();
    warning('Using randomized default parameters');
end
```

Published with MATLAB® R2022b