Contents

- Step 1: Import CSV Data
- Step 2: R Bridge Implementation
- Step 3: MDFT Formulation to Calculate Preference Dynamics
- 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.');

```
\label{local_potential} \textbf{robotChoice\_Data} = \textbf{readtable('G:\My Drive\myResearch\Research Experimentation\Apollo\apollo\ata\Bounding\_Overwatch\_Data\HumanData\_Bounding\_Overwatch - 80Split.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
% 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\cample\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(rscript)
    error('Input CSV not found at: %s', csvFile);
elseif ~isfolder(outputDir)
    warning('Output folder does not exist, creating: %s', outputDir);
    mkdir(outputDir);
```

```
end
% 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');
        else
            jsonFile = outputDir;
        end
        % Parse JSON output
        if exist(jsonFile, 'file')
           jsonText = fileread(jsonFile);
            params = jsondecode(jsonText);
            \% Extract parameters with validation
            %Boundedphi1, phi2 parameters
            %phi1 = max(0, validateParam(params, 'phi1', 0.5)); % Ensure non-negative
            \%phi2 = min(max(0, validateParam(params, 'phi2', 0.8)), 1); \% Constrain 0-1
            %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);
            1;
            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');
        end
    else
        error('R execution failed: %s', result);
    end
catch ME
   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
end
```

Initializing R bridge...

Step 3: MDFT Formulation to Calculate Preference Dynamics

(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.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]
    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 trial
    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(array2table(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');
    else
       disp('X Prediction differs from actual choice');
    % Plot evolution
    figure;
    plot(0:tau, P_tau);
    xlabel('Preference Step (\tau)');
    ylabel('Preference Strength');
    legend({'Robot1','Robot2','Robot3'});
    title(sprintf('Preference Evolution (Trial %d)', current_trial));
    grid on;
%% Step 4: Output Results
disp('Saving results to CSV...');
output_table = table(E_P, V_P, P_tau(end,:)', ...
                     'VariableNames', {'ExpectedPreference', 'VariancePreference', 'FinalPreferences'});
writetable(output_table, 'results.csv');
disp('Results saved successfully!');
%}
=== Trial Analysis ===
Trial: 1
Participant: 175044
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
                       0.53131
                                                      0.27442
                                                                                     0.31001
                                                                                                                      0.053131
                       0.56469
                                                       0.34829
                                                                                      0.27287
                                                                                                                      0.056469
    Robot2
                       0.53834
                                                       0.28146
                                                                                      0.31072
                                                                                                                      0.053834
    Robot3
DFT Results:
E P: NaN NaN NaN
Choice probabilities: NaN NaN NaN
Predicted choice: Robot 1
Actual choice: Robot 1

√ Prediction matches actual choice

=== Trial Analysis ===
```

Trial: 2

Participant: 175044 Actual Choice: Robot 2

0.39007

0.48055

Robot3

•		C2 - Hard Nav, Low Exposure	C3 - Easy Nav, High Exposure	C4 - Hard Nav, High Exposure
Robot1 Robot2 Robot3	0.4991 0.49456 0.48951	0.26554 0.2957 0.26006	0.28347 0.24832 0.2784	0.04991 0.049456 0.048951
DFT Results: E_P: NaN NaN NaI Choice probabilit: Predicted choice: Actual choice: Rol	ies: NaN NaN NaN Robot 1			
=== Trial Analysi: Trial: 3 Participant: 1418: Actual Choice: Rol M matrix (alternat	31	C2 - Hard Nav, Low Exposure	C3 - Easy Nav, High Exposure	C4 - Hard Nav, High Exposure
Robot1 Robot2 Robot3	0.64723 0.55374 0.6102	0.39436 0.32678 0.38052	0.31759 0.28233 0.29071	0.064723 0.055374 0.06102
DFT Results: E_P: NaN NaN NaI Choice probabilit: Predicted choice: Actual choice: Rol	ies: NaN NaN NaN Robot 1			
=== Trial Analysi: Trial: 4 Participant: 1247: Actual Choice: Rol M matrix (alterna	37 bot 3 tives × attributes):			
C1 ·	- Easy Nav, Low Exposure	C2 - Hard Nav, Low Exposure	C3 - Easy Nav, High Exposure	C4 - Hard Nav, High Exposure
Robot1 Robot2 Robot3	0.57548 0.57667 0.59855	0.32031 0.32959 0.33528	0.31272 0.30474 0.32312	0.057548 0.057667 0.059855
DFT Results: E_P: NaN NaN Nal Choice probabilit: Predicted choice: Actual choice: Rol	ies: NaN NaN NaN Robot 1			
<pre>X Prediction diff === Trial Analysi: Trial: 5 Participant: 18170 Actual Choice: Rol</pre>	90			
M matrix (alterna	tives × attributes): - Easy Nav, Low Exposure	C2 - Hard Nav, Low Exposure	C3 - Easy Nav, High Exposure	C4 - Hard Nav, High Exposure
Robot1 Robot2 Robot3	1 0.94014 0.8242	0.57644 0.55706 0.4816	0.52356 0.47709 0.42501	0.1 0.094014 0.08242
DFT Results: E_P: NaN NaN Nal Choice probabilit: Predicted choice: Actual choice: Rol	ies: NaN NaN NaN Robot 1			
=== Trial Analysis Trial: 6 Participant: 1233	10			
	tives × attributes): - Easy Nav, Low Exposure	C2 - Hard Nav, Low Exposure	C3 - Easy Nav, High Exposure	C4 - Hard Nav, High Exposure
Robot1	0.41557 0.39007	0.25914 0.23465	0.19799 0.19442	0.041557 a a390a7

0.23465

0.31041

0.19442

0.21819

0.039007

0.048055

DFT Results: E_P: NaN NaN NaN

Choice probabilities: NaN NaN NaN

Predicted choice: Robot 1 Actual choice: Robot 2

 $\ensuremath{\mathsf{X}}$ Prediction differs from actual choice

=== Trial Analysis === Trial: 7

Participant: 214504

Actual Choice: Robot 1

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
 0.48079
 0.30381
 0.22505
 0.048079

 Robot2
 0.47977
 0.2866
 0.24114
 0.047977

 Robot3
 0.050229
 0.03518
 0.020072
 0.01

DFT Results:

E_P: NaN NaN NaN

Choice probabilities: NaN NaN NaN

Predicted choice: Robot 1 Actual choice: Robot 1

 \checkmark Prediction matches actual choice

=== Trial Analysis ===

Trial: 8

Participant: 123310 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
 0.457
 0.29531
 0.20739
 0.0457

 Robot2
 0.44162
 0.3
 0.18579
 0.044162

 Robot3
 0.42026
 0.25048
 0.21181
 0.042026

DFT Results:

E_P: NaN NaN NaN

Choice probabilities: NaN NaN NaN

Predicted choice: Robot 1 Actual choice: Robot 3

X Prediction differs from actual choice

=== Trial Analysis ===

Trial: 9

Participant: 124737 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 0.40595 0.22564 0.2209 0.040595 Robot2 0.38863 0.19213 0.23536 0.038863 Robot3 0.44683 0.22874 0.26277 0.044683

DFT Results:

E_P: NaN NaN NaN

Choice probabilities: NaN NaN NaN

Predicted choice: Robot 1 Actual choice: Robot 1

 \checkmark Prediction matches actual choice

=== Trial Analysis ===

Trial: 10

Participant: 214504 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

 Robot1
 0.26733
 0.16662
 0.12744
 0.026733

 Robot2
 0.10843
 0.064393
 0.054875
 0.010843

 Robot3
 0.10486
 0.068926
 0.046422
 0.010486

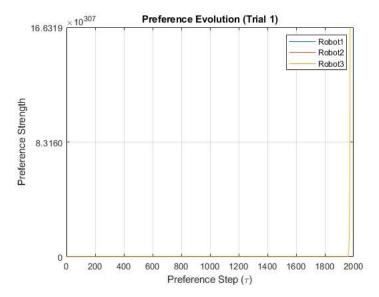
DFT Results:

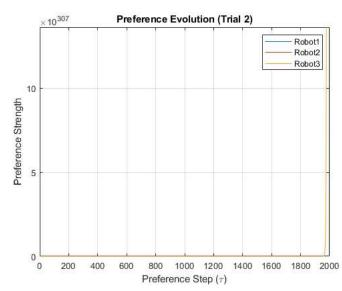
E_P: NaN NaN NaN

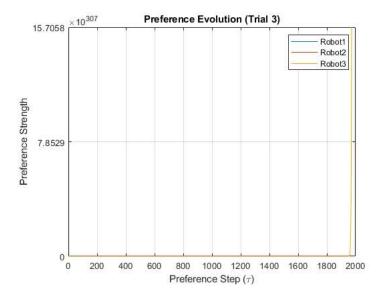
Choice probabilities: NaN NaN NaN

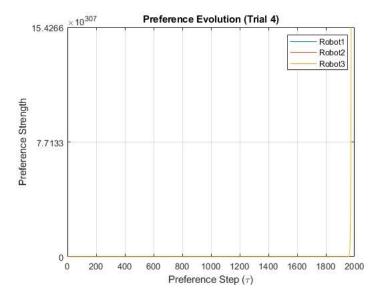
Predicted choice: Robot 1 Actual choice: Robot 2

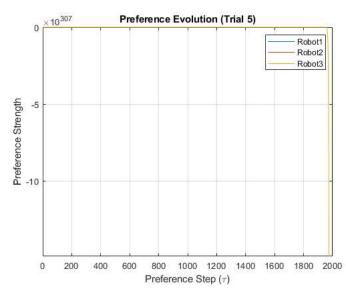
X Prediction differs from actual choice

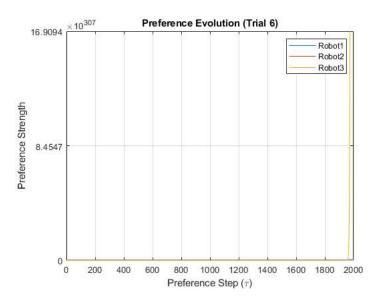


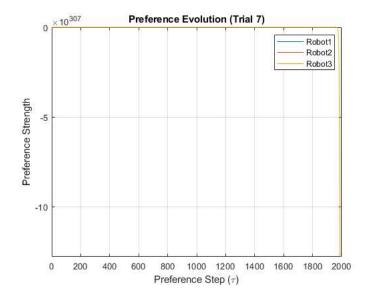


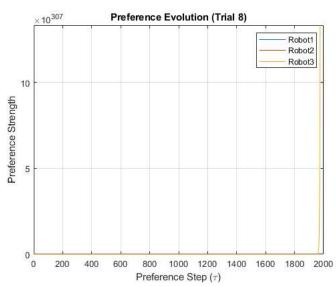


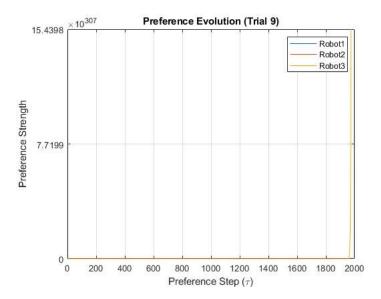


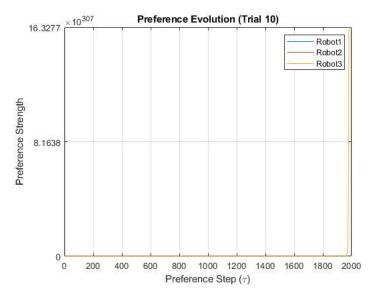












Helper Functions

```
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
```