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% a04 p4 model recovery
% This script sets ExpInfo.nTrials = 20 and counts = 100 to do model
% recovery and generates a confusion matrix
clear all; close all; clc; rng(1); %parpool(20);
% initiate confusion matrix
nModel = 6;
CM = zeros(nModel);
counts = 100;
% experimental parameters
ExpInfo.ifi
                        = 1000/60;
ExpInfo.SOA
                        = [-0.5000, -0.3000, -0.2500, -0.2000, -0.1500, -0.1000, ...
    -0.0500, 0, 0.0500, 0.1000, 0.1500, 0.2000, 0.2500, 0.3000, 0.5000] *1000; % in ms
ExpInfo.lenS
                        = length(ExpInfo.SOA);
ExpInfo.nTrials
                        = 20; % num of trials per SOA
%% ground truth parameters
% used the ground truth parameters from sub7 sess 7 SOA = -300 ms
                        = \{ [40, -20, 70, 50, 110, 80, 0.02, 0.04], \dots \%M1 \}
    [40, -20, 70, 110, 80, 0.02, 0.04], ...%M2
    [40, -20, 70, 50, 110, 0.02, 0.04], ...%M3
    [40, -20, 70, 110, 0.02, 0.04], ...%M4
    [40, 70, 110, 80, 0.02, 0.04], ...%M5
    [40, 70, 110, 0.02, 0.04]};%M6
ParaLabel = {{'\mu_{pre}','\mu_{post}','\sigma_{pre}','\sigma_{post}','criterion_{pre}}
   {'\mu','\sigma','criterion','\lambda_{pre}','\lambda_{post}'}};
%% functions
% define PMF
P Afirst
                        = @(SOA, mu, sig, c, lambda) lambda/3 + (1-lambda).*normcdf(-
P Vfirst
                        = @(SOA, mu, sig, c, lambda) lambda/3 + (1-lambda).*(1 - norm)
                        = @(SOA, mu, sig, c, lambda) ...
P simultaneous
    1 - (lambda/3 + (1-lambda).*normcdf(-c, SOA - mu, sig)) ...
    - (lambda/3 + (1-lambda).*(1 - normcdf(c, SOA-mu, sig)));
M1 = \{
   @(p) P_Vfirst(ExpInfo.SOA, p(1), p(3), p(5), p(7));...
   @(p) P_simultaneous(ExpInfo.SOA, p(1), p(3), p(5), p(7));...
   @(p) P_Afirst(ExpInfo.SOA, p(1), p(3), p(5), p(7));...
   @(p) P_Vfirst(ExpInfo.SOA, p(2), p(4), p(6), p(8));...
   Q(p) P_simultaneous(ExpInfo.SOA, p(2), p(4), p(6), p(8));
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\mathfrak{Q}(\mathfrak{p}) P Afirst(ExpInfo.SOA, \mathfrak{p}(2), \mathfrak{p}(4), \mathfrak{p}(6), \mathfrak{p}(8));
M2 = {
    @(p) P Vfirst(ExpInfo.SOA, p(1), p(3), p(4), p(6));...
    @(p) P_simultaneous(ExpInfo.SOA, p(1), p(3), p(4), p(6));...
    @(p) P_Afirst(ExpInfo.SOA, p(1), p(3), p(4), p(6));...
    @(p) P_Vfirst(ExpInfo.SOA, p(2), p(3), p(5), p(7));...
    @(p) P_simultaneous(ExpInfo.SOA, p(2), p(3), p(5), p(7));...
    @(p) P_Afirst(ExpInfo.SOA, p(2), p(3), p(5), p(7))};
M3 = \{
    @(p) P_Vfirst(ExpInfo.SOA, p(1), p(3), p(5), p(6));...
    @(p) P_simultaneous(ExpInfo.SOA, p(1), p(3), p(5), p(6));...
    @(p) P_Afirst(ExpInfo.SOA, p(1), p(3), p(5), p(6));...
    @(p) P_Vfirst(ExpInfo.SOA, p(2), p(4), p(5), p(7));...
    @(p) P_simultaneous(ExpInfo.SOA, p(2), p(4), p(5), p(7));...
    @(p) P_Afirst(ExpInfo.SOA, p(2), p(4), p(5), p(7))};
M4 = \{
    @(p) P_Vfirst(ExpInfo.SOA, p(1), p(3), p(4), p(5));...
    @(p) P_simultaneous(ExpInfo.SOA, p(1), p(3), p(4), p(5));...
    @(p) P_Afirst(ExpInfo.SOA, p(1), p(3), p(4), p(5));...
    @(p) P_Vfirst(ExpInfo.SOA, p(2), p(3), p(4), p(6));...
    @(p) P_simultaneous(ExpInfo.SOA, p(2), p(3), p(4), p(6));...
    @(p) P_Afirst(ExpInfo.SOA, p(2), p(3), p(4), p(6))};
M5 = {
    @(p) P_Vfirst(ExpInfo.SOA, p(1), p(2), p(3), p(5));...
    @(p) P_simultaneous(ExpInfo.SOA, p(1), p(2), p(3), p(5));...
    @(p) P_Afirst(ExpInfo.SOA, p(1), p(2), p(3), p(5));...
    @(p) P_Vfirst(ExpInfo.SOA, p(1), p(2), p(4), p(6));...
    @(p) P_simultaneous(ExpInfo.SOA, p(1), p(2), p(4), p(6));...
    @(p) P_Afirst(ExpInfo.SOA, p(1), p(2), p(4), p(6))};
M6 = {
    @(p) P_Vfirst(ExpInfo.SOA, p(1), p(2), p(3), p(4));...
    @(p) P_simultaneous(ExpInfo.SOA, p(1), p(2), p(3), p(4));...
    @(p) P_Afirst(ExpInfo.SOA, p(1), p(2), p(3), p(4));...
    @(p) P_Vfirst(ExpInfo.SOA, p(1), p(2), p(3), p(5));...
    @(p) P_simultaneous(ExpInfo.SOA, p(1), p(2), p(3), p(5));...
    @(p) P_Afirst(ExpInfo.SOA, p(1), p(2), p(3), p(5))};
models = {M1; M2; M3; M4; M5; M6};
% run model fitting for multiple times
% initialization
[AIC, min_NLL, estP] = deal(cell(counts, nModel));
for count = 1:counts
    count
    %% simulate fake datasets for 6 models
    parfor iModel
                                      = 1:nModel
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% for the specific model and its corresponding ground-truth
        % parameter sets, generate the probability of reporting: Vfirst,
        % simultaneous, Afirst in pretest; Vfirst, simultaneous, Afirst in
        % posttest
        p_all_cond
                                   = NaN(6, ExpInfo.lenS); % len of conditions x len of
        Para
                                   = TruePara{iModel}; % true parameters for this model
        for iCondition
            p_all_cond(iCondition,:) = models{iModel}{iCondition}(Para);
        end
        % simulate data set
                                   = cell(1,2);
        sim_r_org
                                   = 1:2 % loop through pre and post session
        for s
                                       = p_all_cond(((s-1)*3+1):s*3, :); % pre and post
            sim_r_org{s}
                                       = sampleMatrix(p, ExpInfo.nTrials);
        end
        %% fit fake data to 6 models
        % obtain best-fitting parameters and AIC for each model
        [AIC{count, iModel}, min_NLL{count, iModel}, estP{count, iModel}] = model_rec
        % iBest is the index of model that best fits this specific fake data set
        [M iBEST] = min(AIC{count, iModel});
        BEST = AIC{count, iModel} == M;
        BEST = BEST / sum(BEST);
        CM(iModel,:) = CM(iModel,:) + BEST;
    end
end
count = 1
Starting parallel pool (parpool) using the 'local' profile ...
Connected to the parallel pool (number of workers: 8).
count = 2
count = 3
count = 4
count = 5
count = 6
count = 7
count = 8
count = 9
count = 10
count = 11
count = 12
count = 13
count = 14
count = 15
count = 16
count = 17
count = 18
count = 19
count = 20
count = 21
count = 22
count = 23
count = 24
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count = 25count = 26count = 27count = 28count = 29count = 30count = 31count = 32count = 33count = 34count = 35count = 36count = 37count = 38count = 39count = 40count = 41count = 42count = 43count = 44count = 45count = 46count = 47count = 48count = 49count = 50count = 51count = 52count = 53count = 54count = 55count = 56count = 57count = 58count = 59count = 60count = 61count = 62count = 63count = 64count = 65count = 66count = 67count = 68count = 69count = 70count = 71count = 72count = 73count = 74count = 75count = 76count = 77count = 78count = 79count = 80count = 81count = 82count = 83count = 84count = 85count = 86count = 87count = 88

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count = 89
count = 90
count = 91
count = 93
count = 94
count = 95
count = 96
count = 97
count = 98
count = 99
count = 100
```

## **count = 100**

## fit model 1 2 3 4 5 6 0.25 0.04 0.01 simulated model 0.7 0 0 0.06 0.18 0.75 0.01 0 0 0.07 0.04 0.22 0.67 0 0 0.02 0.18 0.09 0.71 0 0 0.07 0.01 0.71 0.07 0.14 0 0.01 0.05 0.08 0.13 0.1 0.63