Patients assignments to specTra and SWT groups through minimization

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Author: Pauline Gut

Definitions [1–4]

- Minimization is a method of ensuring excellent balance between groups for several prognostic factors, even in small samples.
- Minimization is a widely acceptable approach.
- Minimization is a form of restricted randomization.
- Minimization is an adaptive randomization type: the treatment allocated to the next participant enrolled in the trial depends (wholly or partly) on the characteristics of those participants already enrolled.
- Minimization is a dynamic method since the randomization list is not produced before the trial starts, but during participant recruitment.
- Minimization balances treatment assignments across multiple prognostic factors simultaneously. The aim is that each allocation should minimize the imbalance across multiple factors.
- Minimization has been first proposed by *Taves* [1] and independently generalized by *Pocock and Simon* [2].

Notations

The following is described in [2].

- \bullet N different treatments to assign participants to.
- M prognostic factors that we want to balance.
- Each factor has n_i levels (n_1, \dots, n_M) .
- At any point during the trial, x_{ijk} is the number of patients who have been assigned to treatment k with level j of factor i for $j=1,2,\cdots,n_i;\ i=1,2,\cdots,M$ and $k=1,\cdots,N$.

Consider a new patient entering the trial. Let $r_1, ..., r_M$ be the levels of factors $1, \dots, M$ for this patient.

For each treatment k one considers the new x_{ijl} , denoted by x_{ijl}^k , that would arise if that treatment were assigned to the patient, where the superscript k refers to the particular treatment under consideration:

$$x_{ir_ik}^k = x_{ijk} + 1$$

$$x_{ijl}^k = x_{ijl} \text{ for } l \neq k, j \neq r_i$$

i.e. we add 1 to all the counts x_{ijk} that match the treatment k and the factor levels the participant has, and all the other counts are unchanged.

1. Imbalance within each factor

We start by calculating the imbalance within each factor.

Let $D(\{z\}_{l=1}^N)$ be some function which measures the **amount of variation** in any set of non-negative integers $\{z\}_{l=1}^N$.

Then, if treatment k were assigned to the new patient, we can get the resulting amount of variation that would be produced by assigning the patient to treatment k, and calculate

$$d_{ik} = D(\{x_{ir,i}^k\}_{l=1}^N)$$

the resultant *lack of balance* or *imbalance* in treatment numbers across all treatments, for all patients with level r_i of factor i.

Note that only patients who match the new patient's factor level r_i matter for the calculation of imbalance for factor i, participants with other factor levels don't affect the imbalance calculation.

We can choose different functions for D, see below.

2. Total imbalance

Once we have the imbalance within each factor, we combine them to produce a total imbalance score.

Let G be some function from $\mathbf{R}^M \to \mathbf{R}$ that combines the d_{ik} for all the M factors:

$$G_k = G(d_{1k}, \cdots, d_{Mk}).$$

A choice for G can just the sum.

Then, G_k represents the **total amount of imbalance** in treatment numbers which would exist at all the factor levels of the new patient if treatment k were assigned to that patient. In other words, G_k is the total imbalance we would have if the new patient is assigned to treatment k. We calculate G_k for all the N treatments.

3. Assigning probabilities

We can then rank the treatments according to their values G_k (ascending ranks with increasing G_k values corresponding to larger amounts of imbalance), treatment (1) having minimal G_k , treatment (2) having the next smallest G_k , etc., so that (s) \dagger (t) iff $G_{(s)} < G_{(t)}$. In the case of ties, a random ordering can be determined.

The assigned treatment T can be determined from the following set of probabilities:

$$\operatorname{prob}(T=(k))=p_k$$
 where $p_1\geq p_2\geq \cdots \geq p_N$ and $\sum p_k=1.$

The values of p_k can be fixed constants or functions of G_k . This ordering of probabilities means that treatments with small values of G_k have a higher probability of being chosen.

The entire procedure is repeated when the next patient enters the trial.

4. The choice of D

Four possible formulae for D are considered in [4]. Here we present the one we are using:

(i) The Range distance measure of $\{z\}_l$ is defined to be the absolute value of the difference between each $\{z\}_{l=1}^N$.

If we are interested in comparing pairs of treatments in analysis it may be preferable since D would then be measuring the most imbalance in any pair. Also, with only two treatments, the range $(=|z_1-z_2|)$ is equivalent to the standard deviation.

5. The choice of G

One reasonable way of combining $\{d_{ik}\}_{i=1}^{M}$ is to sum then. That is,

$$G_k = G(d_{ik}, \cdots, d_{Mk}) = \sum_{i=1}^M d_{ik}.$$

In the case where some prognostic factors are considered more important than others. One can then make G_k a weighted sum of $\{d_{ik}\}$ so that

$$G_k = G(d_{ik}, \cdots, d_{Mk}) = \sum_{i=1}^{M} w_i \ d_{ik}$$

where $\{w_i\}$ are constants.

6. The choice $\{p_k\}$

The formula for $\{p_k\}$ determines the extent to which one wishes to bias treatment assignment in favor of those treatments with small G_k . Different types of formula are suggested in [4].

Summary:

Minimization is a dynamic algorithm involving three process steps [2]:

- 1. First, an **imbalance score** is computed for each available treatment based on all previous allocations, as well as the hypothetical allocation of the current patient to each treatment.
- 2. In the second step, the preferred treatment is selected by choosing the treatment allocation associated with the smallest imbalance score after the hypothetical allocation the remaining treatments are considered non-preferred. If several treatments are tied with respect to imbalance score, the tie can be broken by selecting the preferred treatment at random.
- 3. Finally, in the third step, **allocation probabilities** are computed for the treatments and the patient is randomized accordingly.

References

- D. R. Taves. Minimization: A new method of assigning patients to treatment and control groups. CLINICAL PHARMACOLOGY and THERAPEUTICS, 15(5), 1974. ISSN 01406736. doi: 10.1016/S0140-6736(64)90114-X.
- [2] Stuart J. Pocock and Richard Simon. Sequential Treatment Assignment with Balancing for Prognostic Factors in the Controlled Clinical Trial. *Biometrics*, 31(1):103, 1975. ISSN 0006341X. doi: 10.2307/2529712.
- [3] Douglas G. Altman and J. Martin Bland. Treatment allocation by minimisation. *BMJ*, 330(7495): 843, 2005. ISSN 14685833. doi: 10.1136/bmj.330.7495.843.
- [4] Neil W. Scott, Gladys C. McPherson, Craig R. Ramsay, and Marion K. Campbell. The method of minimization for allocation to clinical trials: A review. Controlled Clinical Trials, 23(6):662–674, 2002. ISSN 01972456. doi: 10.1016/S0197-2456(02)00242-8.

Minimization algorithm

```
clear; close all; clc;
% Choose the excel file containing the previous allocated patients to the trial
% Each row is a patient
% The first column corresponds to the patient ID
% Each other column corresponds to a variate to be balanced over the treatment groups
disp('Choose the excel file containing the previous allocated patients to the trial.')
disp('Each row must be a patient. The first column corresponds to the patient ID.')
disp('Each other column corresponds to a variate to be balanced over the treatment groups.')
[Filename,Pathname]=uigetfile('*.xlsx','Allocated patients to the trial');
cd(Pathname)
Allocated_Patients = readtable(Filename);
% Enter the number of patients you want to randomize with simple
% randomization
n = size(Allocated_Patients,1);
clc;
New_Patient.ID = input('New patient ID: ', 's');
New_Patient.SixMWT = input('6MWT of the new patient (in meters): ');
New_Patient.Age = input('Age of the new patient: ');
if n == 0
p = size(Allocated Patients,1) + 1;
PatientID{1,1} = New Patient.ID;
Patient6MWT(1,1) = New_Patient.SixMWT;
PatientAge(1,1) = New_Patient.Age;
Allocated_Patients(p,1) = table(PatientID(1,1));
Allocated_Patients(p,2) = table(Patient6MWT(1,1));
Allocated_Patients(p,3) = table(PatientAge(1,1));
% Treatment allocation to the patient through simple randomization
Treat = randi(2,1,1);
SWT{1,1} = 'SWT';
DOT{1,1} = 'DOT';
  if Treat == 1
    Allocated_Patients(p,4) = DOT;
    disp('Patient allocated to DOT')
  else
    Allocated_Patients(p,4) = SWT;
    disp('Patient allocated to SWT')
  end
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Allocated Patients.Properties.VariableNames = {'PatientID', 'x6MWT', 'Age', 'Treatment'};
writetable(Allocated_Patients, Filename, 'WriteVariableNames', true)
disp('Finished!')
else
% Treatment allocation to the patient through minimization
% Calculate the mean 6MWT and the mean age of the patients previsouly
% allocated to the trial
Mean_SixMWT = mean(cell2mat(table2cell(Allocated_Patients(:,2))));
Mean_Age = mean(cell2mat(table2cell(Allocated_Patients(:,3))));
% Computation of the amount of variation among treatments
Smaller_Mean_SixMWT = 0;
Equal_Mean_SixMWT = 0;
Larger Mean SixMWT = 0;
Smaller Mean Age = 0;
Equal_Mean_Age = 0;
Larger_Mean_Age = 0;
for i = 1:size(Allocated_Patients,1)
if strcmp(table2cell(Allocated_Patients(i,4)), 'DOT')
  Smaller Mean SixMWT = Smaller Mean SixMWT + sum(cell2mat(table2cell(Allocated Patients(i,2)))
< Mean_SixMWT);
  Equal_Mean_SixMWT = Equal_Mean_SixMWT + sum(cell2mat(table2cell(Allocated_Patients(i,2))) ==
Mean_SixMWT);
  Larger Mean SixMWT = Larger Mean SixMWT + sum(cell2mat(table2cell(Allocated Patients(i,2))) >
Mean SixMWT);
  Smaller Mean Age = Smaller Mean Age + sum(cell2mat(table2cell(Allocated Patients(i,3))) <
Mean Age);
  Equal_Mean_Age = Equal_Mean_Age + sum(cell2mat(table2cell(Allocated_Patients(i,3))) ==
Mean_Age);
  Larger Mean Age = Larger Mean Age + sum(cell2mat(table2cell(Allocated Patients(i,3))) >
Mean_Age);
  T(1,:) = table(Smaller_Mean_SixMWT, Equal_Mean_SixMWT, Larger_Mean_SixMWT,
Smaller_Mean_Age, Equal_Mean_Age, Larger_Mean_Age);
end
end
clear Smaller_Mean_SixMWT Equal_Mean_SixMWT Larger_Mean_SixMWT Smaller_Mean_Age
Equal_Mean_Age Larger_Mean_Age
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Smaller_Mean_SixMWT = 0;
Equal Mean SixMWT = 0;
Larger_Mean_SixMWT = 0;
Smaller_Mean_Age = 0;
Equal_Mean_Age = 0;
Larger_Mean_Age = 0;
for i = 1:size(Allocated_Patients,1)
if strcmp(table2cell(Allocated_Patients(i,4)), 'SWT')
  Smaller_Mean_SixMWT = Smaller_Mean_SixMWT + sum(cell2mat(table2cell(Allocated_Patients(i,2)))
< Mean_SixMWT);
  Equal_Mean_SixMWT = Equal_Mean_SixMWT + sum(cell2mat(table2cell(Allocated_Patients(i,2))) ==
Mean_SixMWT);
  Larger_Mean_SixMWT = Larger_Mean_SixMWT + sum(cell2mat(table2cell(Allocated_Patients(i,2))) >
Mean_SixMWT);
  Smaller_Mean_Age = Smaller_Mean_Age + sum(cell2mat(table2cell(Allocated_Patients(i,3))) <
Mean Age);
  Equal_Mean_Age = Equal_Mean_Age + sum(cell2mat(table2cell(Allocated_Patients(i,3))) ==
Mean_Age);
  Larger_Mean_Age = Larger_Mean_Age + sum(cell2mat(table2cell(Allocated_Patients(i,3))) >
Mean_Age);
  T(2,:) = table(Smaller Mean SixMWT, Equal Mean SixMWT, Larger Mean SixMWT,
Smaller_Mean_Age, Equal_Mean_Age, Larger_Mean_Age);
end
end
clear Smaller_Mean_SixMWT Equal_Mean_SixMWT Larger_Mean_SixMWT Smaller_Mean_Age
Equal_Mean_Age Larger_Mean_Age
if New Patient.SixMWT < Mean SixMWT
  T(3,1) = cell2table(num2cell(1));
  T(4,1) = cell2table(num2cell(sum(T{[1 3],1}, 1)));
  T(5,1) = T(2,1);
  T(6,1) = cell2table(num2cell(abs(diff(T{[4 5],1}, 1))));
  T(7,1) = T(1,1);
  T(8,1) = cell2table(num2cell(sum(T{[2 3],1}, 1)));
  T(9,1) = cell2table(num2cell(abs(diff(T{[7 8], 1}, 1))));
elseif New_Patient.SixMWT == Mean_SixMWT
  T(3,2) = cell2table(num2cell(1));
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T(4,2) = cell2table(num2cell(sum(T{[1 3],2}, 1)));
  T(5,2) = T(2,2);
  T(6,2) = cell2table(num2cell(abs(diff(T{[4 5],2}, 1))));
  T(7,2) = T(1,2);
  T(8,2) = cell2table(num2cell(sum(T{[2 3],2}, 1)));
  T(9,2) = cell2table(num2cell(abs(diff(T{[7 8], 2}, 1))));
elseif New_Patient.SixMWT > Mean_SixMWT
  T(3,3) = cell2table(num2cell(1));
  T(4,3) = cell2table(num2cell(sum(T{[1 3],3}, 1)));
  T(5,3) = T(2,3);
  T(6,3) = cell2table(num2cell(abs(diff(T{[4 5],3}, 1))));
  T(7,3) = T(1,3);
  T(8,3) = cell2table(num2cell(sum(T{[2 3],3}, 1)));
  T(9,3) = cell2table(num2cell(abs(diff(T{[7 8], 3}, 1))));
end
if New_Patient.Age < Mean_Age
  T(3,4) = cell2table(num2cell(1));
  T(4,4) = cell2table(num2cell(sum(T{[1 3],4}, 1)));
  T(5,4) = T(2,4);
  T(6,4) = cell2table(num2cell(abs(diff(T{[4 5],4}, 1))));
  T(7,4) = T(1,4);
  T(8,4) = cell2table(num2cell(sum(T{[2 3],4}, 1)));
  T(9,4) = cell2table(num2cell(abs(diff(T{[7 8], 4}, 1))));
elseif New Patient.Age == Mean Age
  T(3,5) = cell2table(num2cell(1));
  T(4,5) = cell2table(num2cell(sum(T{[1 3],5}, 1)));
  T(5,5) = T(2,5);
  T(6,5) = cell2table(num2cell(abs(diff(T{[4 5],5}, 1))));
  T(7,5) = T(1,5);
  T(8,5) = cell2table(num2cell(sum(T{[2 3],5}, 1)));
  T(9,5) = cell2table(num2cell(abs(diff(T{[7 8], 5}, 1))));
elseif New_Patient.Age > Mean_Age
  T(3,6) = cell2table(num2cell(1));
  T(4,6) = cell2table(num2cell(sum(T{[1 3],6}, 1)));
  T(5,6) = T(2,6);
  T(6,5) = cell2table(num2cell(abs(diff(T{[4 5],6}, 1))));
  T(7,6) = T(1,6);
  T(8,6) = cell2table(num2cell(sum(T{[2 3],6}, 1)));
  T(9,6) = cell2table(num2cell(abs(diff(T{[7 8], 6}, 1))));
end
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```
T(6,7) = cell2table(num2cell(sum(T(6,:), 2)));
T(9,7) = cell2table(num2cell(sum(T{9,:}, 2)));
T.Properties.RowNames = {'DOT group', 'SWT group', 'New patient', 'DOT + new patient', 'SWT',
'Absolute difference 1', 'DOT', 'SWT + new patient', 'Absolute difference 2'};
% Treatment allocation to the patient
p = size(Allocated_Patients,1) + 1;
PatientID{1,1} = New_Patient.ID;
Patient6MWT(1,1) = New_Patient.SixMWT;
PatientAge(1,1) = New_Patient.Age;
Allocated_Patients(p,1) = table(PatientID(1,1));
Allocated_Patients(p,2) = table(Patient6MWT(1,1));
Allocated_Patients(p,3) = table(PatientAge(1,1));
SWT{1,1} = 'SWT';
DOT{1,1} = 'DOT';
if cell2mat(table2cell(T(6,7))) > cell2mat(table2cell(T(9,7)))
  Allocated_Patients(p,4) = SWT;
  disp('Patient allocated to SWT')
elseif cell2mat(table2cell(T(6,7))) == cell2mat(table2cell(T(9,7)))
  disp('Simple randomization')
  Treat = randi(2,1,1);
  if Treat == 1
    Allocated_Patients(p,4) = DOT;
    disp('Patient allocated to DOT')
  else
    Allocated_Patients(p,4) = SWT;
    disp('Patient allocated to SWT')
  end
elseif cell2mat(table2cell(T(6,7))) < cell2mat(table2cell(T(9,7)))
  Allocated Patients(p,4) = DOT;
  disp('Patient allocated to DOT')
end
S{1,1} = 'Number of patients in DOT: ';
S{2,1} = 'Number of patients in SWT: ';
S{3,1} = 'Mean 6MWT in DOT: ';
```

```
S{4,1} = 'Mean 6MWT in SWT: ';
S{5,1} = 'Mean age in DOT: ';
S{6,1} = 'Mean age in SWT: ';
S{1,2} = sum(strcmp(table2cell(Allocated_Patients(:,4)), 'DOT'),1);
S{2,2} = sum(strcmp(table2cell(Allocated_Patients(:,4)), 'SWT'),1);
for i = 1:size(Allocated_Patients,1)
  if strcmp(table2cell(Allocated_Patients(i,4)), 'DOT')
    Mean_DOT1(i,1) = cell2mat(table2cell(Allocated_Patients(i,2)));
    Mean_DOT2(i,1) = cell2mat(table2cell(Allocated_Patients(i,3)));
  elseif strcmp(table2cell(Allocated_Patients(i,4)), 'SWT')
    Mean_SWT1(i,1) = cell2mat(table2cell(Allocated_Patients(i,2)));
    Mean_SWT2(i,1) = cell2mat(table2cell(Allocated_Patients(i,3)));
  end
end
for i = 1:size(Allocated_Patients,1)
  if ~strcmp(table2cell(Allocated_Patients(i,4)), 'DOT')
    Mean_DOT1(i,1) = 0;
    Mean_DOT2(i,1) = 0;
  elseif ~strcmp(table2cell(Allocated_Patients(i,4)), 'SWT')
    Mean SWT1(i,1) = 0;
    Mean_SWT2(i,1) = 0;
  end
end
Mean SixMWT DOT = mean(Mean DOT1);
Mean_Age_DOT = mean(Mean_DOT2);
Mean SixMWT SWT = mean(Mean SWT1);
Mean_Age_SWT = mean(Mean_SWT2);
S{3,2} = round(Mean SixMWT DOT);
S{4,2} = round(Mean SixMWT SWT);
S{5,2} = round(Mean\_Age\_DOT);
S{6,2} = round(Mean_Age_SWT);
Population Stat = table(S);
T.Properties.VariableNames([7]) = {'Sum'};
writetable(Allocated_Patients, Filename, 'WriteVariableNames', true)
writetable(Population_Stat, 'TrialPopStat.xlsx', 'WriteVariableNames', false)
disp('Finished!')
```

%clearvars -except Allocated_Patients Population_Stat T

end

system(['set PATH=' Pathname ' && ' Filename]);