

Supplementary material for “A study of two types of split-plot designs” by Tsai (2016)

This supplementary file describes the use of the R functions for the study of split-plot designs with few whole-plot factors and blocked split-plot designs discussed in Tsai (2016). Two functions `find.SSP()` and `find.BSP()` are available for generating optimal designs with respect to our criterion for a given design configuration. Four functions `SSP()`, `SSP.wd()`, `BSP()`, `BSP.wd()` are given for computing the m_i -values, wordlength pattern, and the criterion values of Section 3 for a given design. All the functions are included in a program file, and we can source the entire script by typing the following `source()` command in the R console.

```
source("http://math.ntnu.edu.tw/~pwtsai/doe/sspsp_prg.R")
```

- `find.SSP(nfat, nUnit)` — this is used to generate optimal split-plot designs with few whole-plot factors for `nfat = c(n1, n2)` and `nUnit = c(w, s)`, where `n1`, `n2`, `w`, `s` are the numbers of whole-plot factors, subplot factors, whole-plots, and subplots per whole-plot, respectively.
- `SSP.wd(nfat, nUnit, wd1, wd2)`: — this is used to compute the m_i -values, wordlength pattern, and the criterion values in (b) and (c) of Section 3 for a split-plot design with few whole-plot factors, where `wd1` is the subplot words and `wd2` is the splitting words. For example, `SSP.wd(c(3,4), c(16,2), c("ABPR", "ABCQS"), c("PQ"))` is used to evaluate a 32-run design with `n1=3`, `n2=4`, `w=16`, `s=2`, and the design has $R = ABP$ and $S = ABCQ$ as subplot words and PQ as the splitting word.
- `SSP(nfat, nUnit, vec1, vec2)` — this function is similar to that of `SSP.wd()` and is useful for the cases when, instead of subplot and splitting words, subplot columns (`vec1`) and splitting columns (`vec2`) are given, as in Tables 3 and 4 of Tsai (2016).
- `find.BSP(nfat, nUnit)` — this is used to generate optimal blocked split-plot designs for `nfat = c(n1, n2)` and `nUnit = c(b, w, s)`, where `n1`, `n2`, `w`, `s` are defined as before and `b` is the number of blocks.
- `BSP.wd(nfat, nUnit, wd1, wd2)` — this is used to compute the m_i -values, word-length pattern, and the criterion values in (a), (b) and (c) of Section 3 for a given blocked split-plot design, where `wd1` are the whole-plot and subplot defining words, and `wd2` is the blocking and splitting words. We note that in all the cases discussed in McLeod and Brewster (2004), and therefore in Tsai (2016), splitting words (if they are needed) are a subset of blocking words and we do not need to separate these two words in this function.
- `BSP(nfat, nUnit, vec1, vec2)` — this function is similar to that of `BSP.wd()` with `vec1` as the whole-plot and subplot columns and `vec2` the blocking and splitting columns.

Split-plot designs with few whole-plot factors

- (1). Consider the design configuration with $(n_1 \ n_2; w \ s) = (3 \ 4; 16 \ 2)$. We use the following function to generate the split-plot design with few whole-plot factors with respect to our criterion.

```
find.SSP(c(3,4),c(16,2))

defining words:  ABCPR ABCQS ( 15 23 )
splitting words:  PQ ( 24 )
mi in whole-plot stratum:  2 2 2 1 1 1 0 0 0 0 0 0
mi in subplot stratum:  1 1 1 1 1 1 1 1 1 1 1 1
(b) and (c) : ( 21 27 ); ( 12 12 )
word length pattern:  0 1 2 0 0
    user  system elapsed
    0.051   0.000   0.051
```

The best design we found has *ABCPR* and *ABCQS* as subplot words (columns 15 and 23) and *PQ* (column 24) as the splitting word. The m_i^W 's and m_i^S 's are given. The sums of m_i and m_i^2 in conditions (b) and (c) are (21, 27) and (12, 12), respectively. The word length pattern is $A_3 = 0$, $A_4 = 1$, and $A_5 = 2$. We note that this design is denoted d_2 in Tsai (2016).

- (2). For the case $(n_1 \ n_2; w \ s) = (3 \ 4; 16 \ 2)$, consider a design which has *ABPR* and *ABCQS* as the independent words and *PQ* as the splitting word. This design is denoted d_1 in Tsai (2016). To evaluate this design, we use the following function.

```
SSP.wd(c(3,4), c(16,2), c("ABPR", "ABCQS"), c("PQ"))

mi in whole-plot stratum:  2 1 1 1 1 1 1 1 0 0 0 0
mi in subplot stratum:  2 2 1 1 1 1 1 1 1 1 0 0
(b) and (c) : ( 21 27 ); ( 12 16 )
word length pattern:  0 1 2 0 0
```

Comparing (1) and (2), we note that d_1 and d_2 are equally good with respect to the usual minimum aberration criterion but d_2 is a better design with respect to our criterion.

- (3). Function `SSP()` is used to evaluate a design when, instead of subplot and splitting words, subplot and splitting columns are given.

For example, for the case $(n_1 \ n_2; w \ s) = (1 \ 13; 8 \ 4)$, the design in Table 2 of BSS has 3, 5, 9, 14, 15, 17, 22, 26, 28 as subplot columns and 6, 24 as splitting columns. To evaluate this design, we use the following function.

```
SSP(c(1, 13), c(8, 4), c(3,5,9,14,15,17,22,26,28), c(6, 24))
```

```

mi in whole-plot stratum: 6 5 5 5 5 5
mi in subplot stratum: 6 6 5 5 5 5 5 1 1 1
(b) and (c) : ( 76 386 ); ( 45 225 )
word length pattern: 5 55 45 96 106 87 82 16 17 1 1 0

```

For the case (1 13; 8 4), we consider the design given in Table 3 of Tsai (2016).

```

> SSP(c(1,13), c(8,4), c(3,7,13,14,19,22,25,26,28), c(10,20))
mi in whole-plot stratum: 6 6 6 1 1 1
mi in subplot stratum: 5 5 5 5 5 5 5 5 5 5
(b) and (c) : ( 76 386 ); ( 55 275 )
word length pattern: 5 55 45 96 106 87 82 16 17 1 1 0

```

We note that these designs are equally good with respect to the usual minimum aberration criterion but our design has more two-factor interactions in the subplot stratum. Clearly, our design is a better design, as discussed on Page 10 of Tsai (2016).

(4). Three cases where there are more than one admissible design.

```

> find.SSP(c(1,8),c(8,4))
defining words: APQT PQRU PQSV AQRSW ( 7 14 22 29 ) splitting words: PR QS ( 10 20 )
(b) and (c) : ( 36 72 ); ( 27 57 ) #BSS
defining words: APQT PQRU AQRV APRSW ( 7 14 13 27 ) splitting words: PR QS ( 10 20 )
(b) and (c) : ( 36 78 ); ( 28 64 ) #New
defining words: APQT PQRU APSV QRSW ( 7 14 19 28 ) splitting words: PR QS ( 10 20 )
(b) and (c) : ( 36 90 ); ( 29 73 ) #New
user system elapsed
10.058 0.052 10.110

```

```

> find.SSP(c(1,9),c(8,4))
defining words: APQT PQRU PQSV AQRSW APRSX ( 7 14 22 29 27 ) splitting words:PR QS (10 20)
(b) and (c) : ( 45 105 ); ( 33 81 ) #BSS
defining words: APQT PQRU APSV QRSW ARSX ( 7 14 19 28 25 ) splitting words:PR QS (10 20)
(b) and (c) : ( 45 135 ); ( 36 108 ) #New
user system elapsed
40.453 0.124 40.574

```

```

> find.SSP(c(2,8),c(8,4))
defining words: ABPS ABQT APQRU BPQRV ABRW ( 7 11 29 30 19 ) splitting words: PR ( 20 )
(b) and (c) : ( 45 105 ); ( 32 64 ) #BSS
defining words: APQS ABQT PQRU BQRV ABPQRW ( 13 11 28 26 31 ) splitting words: PR ( 20 )
(b) and (c) : ( 45 135 ); ( 36 108 ) #New
defining words: ABPS ABQT PQRU AQRV BQRW ( 7 11 28 25 26 ) splitting words: PR ( 20 )
(b) and (c) : ( 45 141 ); ( 37 117 ) #New
user system elapsed
51.097 0.204 51.297

```

- (5). Eight cases where better 32-run split-plot designs are listed in Table 3 of Tasi (2016). Clearly, for each class, our design is better when compared with BSS's designs.

```

(n1, n2; w s) = ( 1 5 ; 8 4 )
subplot columns: 15  splitting column: 10 22  #New
(b) and (c) : ( 15 15 ); ( 13 13 )
word length pattern: 0 0 1 0
subplot columns: 31 ; splitting column: 10 18  #BSS
(b) and (c) : ( 15 15 ); ( 12 12 )
word length pattern: 0 0 0 1
=====
(n1, n2; w s) = ( 1 6 ; 8 4 )
subplot columns: 14 27  splitting column: 10 22  #New
(b) and (c) : ( 21 27 ); ( 18 22 )
word length pattern: 0 1 2 0 0
subplot columns: 15 30  splitting column: 10 18  #BSS
(b) and (c) : ( 21 27 ); ( 17 23 )
word length pattern: 0 1 2 0 0
=====
(n1, n2; w s) = ( 1 13 ; 8 4 ) ( discussed in (3))
(b) and (c) : ( 76 386 ); ( 55 275 )  #New
(b) and (c) : ( 76 386 ); ( 45 225 )  #BSS
=====
(n1, n2; w s) = ( 1 14 ; 8 4 )
subplot columns: 3 5 9 14 15 17 22 23 26 28  splitting column: 10 20  #New
(b) and (c) : ( 87 519 ); ( 51 303 )
word length pattern: 6 77 62 168 188 203 188 56 62 7 6 0 0
subplot columns: 3 5 9 14 15 17 22 23 26 28  splitting column: 6 24  #BSS
(b) and (c) : ( 87 519 ); ( 50 290 )
word length pattern: 6 77 62 168 188 203 188 56 62 7 6 0 0
=====
(n1, n2; w s) = ( 2 4 ; 8 4 )
subplot columns: 15  splitting column: 20  #New
(b) and (c) : ( 15 15 ); ( 13 13 )
word length pattern: 0 0 1 0
subplot columns: 31  splitting column: 20  #BSS
(b) and (c) : ( 15 15 ); ( 12 12 )
word length pattern: 0 0 0 1
=====
(n1, n2; w s) = ( 2 5 ; 8 4 )
subplot columns: 15 28  splitting column: 20  #New
(b) and (c) : ( 21 27 ); ( 18 22 )
word length pattern: 0 1 2 0 0
subplot columns: 25 30  splitting column: 20  #BSS
(b) and (c) : ( 21 27 ); ( 18 24 )
word length pattern: 0 1 2 0 0
=====
(n1, n2; w s) = ( 2 6 ; 8 4 )
subplot columns: 15 27 28  splitting column: 20  #New
(b) and (c) : ( 28 46 ); ( 24 36 )
word length pattern: 0 3 4 0 0 0

```

```

subplot columns: 15 21 27 splitting column: 29      #BSS
(b) and (c) : ( 28 46 ); ( 24 40 )
word length pattern: 0 3 4 0 0 0
=====
(n1, n2; w s) = ( 3 4 ; 16 2 )
subplot columns: 15 23 splitting column: 24      #New
(b) and (c) : ( 21 27 ); ( 12 12 )
word length pattern: 0 1 2 0 0
subplot columns: 11 23 splitting column: 24      #BSS
(b) and (c) : ( 21 27 ); ( 12 16 )
word length pattern: 0 1 2 0 0

```

Blocked split-plot designs

- (1). Consider the design configuration $(n_1 \ n_2; b \ w \ s) = (3 \ 4; 2 \ 4 \ 4)$. We use the following function to generate optimal blocked split-plot design with respect to our criterion.

```
find.BSP(c(3,4),c(2,4,4))
```

```

defining words:  ABPQR ACPQS ( 27 29 )
splitting/blocking words:  ABC ( 7 )
mi in block stratum:  0
mi in whole-plot stratum:  2 1 1
mi in subplot stratum:  2 2 1 1 1 1 1 1 1 1 1 1 1 1 0 0 0 0 0
(a), (b) and (c) : ( 21 27 ); ( 21 27 ); ( 17 21 )
word length pattern:  0 1 2 0 0

```

The best design we found has $ABPQR$ and $ACPQS$ as defining words (columns 27 and 29) and ABC (column 7) as the blocking word. The sums of m_i and m_i^2 in conditions (a), (b) and (c) are (21, 27) (21, 27) and (17, 21), respectively. The word length pattern for this design is $A_3 = 0$, $A_4 = 1$, and $A_5 = 2$. This design is denoted d_4 in Tsai (2016).

- (2). For the case $(n_1 \ n_2; b \ w \ s) = (3 \ 4; 2 \ 4 \ 4)$, consider another design with $ABPR$ and $ABCQS$ as the defining words and ABC as the blocking word. This design is denoted d_3 in Tsai (2016). We evaluate this design by running the following function.

```

BSP.wd(c(3,4), c(2,4,4), c("ABPR", "ABCQS"), c("ABC"))

mi in block stratum:  1
mi in whole-plot stratum:  2 1 1
mi in subplot stratum:  2 2 1 1 1 1 1 1 1 1 1 1 1 1 0 0 0 0 0
(a), (b) and (c) : ( 21 27 ); ( 20 26 ); ( 16 20 )
word length pattern:  0 1 2 0 0
MB word length pattern:  0 1 1 2 2 0 0 0 0 1

```

Comparing (1) and (2), we note that d_3 and d_4 are equally good with respect to the usual minimum aberration criterion, but clearly d_4 is a better design as discussed in Section 3 of Tsai (2016).

- (3). Four cases where more than one design are listed in MB's Table 1, only the first design is admissible and an additional admissible design up to equivalence is found.

```
> find.BSP(c(2,4), c(4,4,2))
defining words: ABPQRS ( 31 )   splitting/blocking words: PQ APR ( 12 21 ) #New
(a), (b) and (c) : ( 15 15 ); ( 14 14 ); ( 8 8 )
word length pattern: 0 0 0 1
defining words: PQRS ( 28 )   splitting/blocking words: APQ BPR ( 13 22 ) #MB1
(a), (b) and (c) : ( 15 21 ); ( 15 21 ); ( 8 8 )
word length pattern: 0 1 0 0
```

```
> find.BSP(c(3,3), c(4,4,2))
defining words: ABCPR ( 15 )   splitting/blocking words: AC APQ ( 5 25 ) #New
(a), (b) and (c) : ( 15 15 ); ( 14 14 ); ( 9 9 )
word length pattern: 0 0 1 0
defining words: ABPR ( 11 )   splitting/blocking words: ABC APQ ( 7 25 ) #MB1
(a), (b) and (c) : ( 15 21 ); ( 15 21 ); ( 9 13 )
word length pattern: 0 1 0 0
```

```
> find.BSP(c(3,4), c(4,4,2))
defining words: ABCPR ABCQS ( 15 23 )   splitting/blocking words: AC APQ ( 5 25 ) #New
(a), (b) and (c) : ( 21 27 ); ( 20 26 ); ( 12 12 )
word length pattern: 0 1 2 0 0
defining words: ABPR ABQS ( 11 19 )   splitting/blocking words: ABC APQ ( 7 25 ) #MB1
(a), (b) and (c) : ( 21 39 ); ( 21 39 ); ( 12 20 )
word length pattern: 0 3 0 0 0
```

```
> find.BSP(c(5,3), c(4,4,2))
number of admissible design(s): 2
defining words: ABCE ABPQ BCDPR ( 7 19 30 ) splitting/blocking words: AC ABD ( 5 11 ) #New
(a), (b) and (c) : ( 28 46 ); ( 25 41 ); ( 15 23 )
word length pattern: 0 3 4 0 0 0
defining words: ABCE ABPQ BDPR ( 7 19 26 ) splitting/blocking words: AC ABD ( 5 11 ) #MB1
(a), (b) and (c) : ( 28 58 ); ( 26 54 ); ( 15 31 )
word length pattern: 0 5 0 2 0 0
```

- (4). Seven cases where only one design is listed in MB's Table 1 and an additional admissible design up to equivalence is found.

```
> find.BSP(c(1,5), c(4,2,4))
defining words: APQRT ( 15 )   splitting/blocking words: PR PQS ( 10 22 ) #New
(a), (b) and (c) : ( 15 15 ); ( 14 14 ); ( 13 13 )
word length pattern: 0 0 1 0
defining words: PQRT ( 14 )   splitting/blocking words: APR PQS ( 11 22 ) #MB
(a), (b) and (c) : ( 15 21 ); ( 15 21 ); ( 13 17 )
```

```

word length pattern: 0 1 0 0

> find.BSP(c(1,6), c(4,2,4))
defining words: APQRT APRSU ( 15 27 )  splitting/blocking words: PR PQS ( 10 22 )  #New
(a), (b) and (c) : ( 21 27 ); ( 20 26 ); ( 18 22 )
word length pattern: 0 1 2 0 0
defining words: PQRT PRSU ( 14 26 )  splitting/blocking words: APR PQS ( 11 22 )  #MB
(a), (b) and (c) : ( 21 39 ); ( 21 39 ); ( 18 30 )
word length pattern: 0 3 0 0 0

> find.BSP(c(3,4), c(4,2,4))
defining words: ABPR ACPQS ( 11 29 )  splitting/blocking words: AB AC ( 3 5 )  #New
(a), (b) and (c) : ( 21 27 ); ( 17 21 ); ( 17 21 )
word length pattern: 0 1 2 0 0
defining words: APQR BCPQS ( 25 30 )  splitting/blocking words: AB AC ( 3 5 )  #MB
(a), (b) and (c) : ( 21 27 ); ( 18 24 ); ( 17 23 )
word length pattern: 0 1 2 0 0

> find.BSP(c(3,5), c(4,2,4))
defining words: ABPR ABQS ACPQT ( 11 19 29 )  splitting/blocking words: AB AC ( 3 5 )
(a), (b) and (c) : ( 28 46 ); ( 23 35 ); ( 23 35 )  #New
word length pattern: 0 3 4 0 0 0
defining words: APQR BCPQS ABCPT ( 25 30 15 )  splitting/blocking words: AB AC ( 3 5 )
(a), (b) and (c) : ( 28 46 ); ( 25 43 ); ( 23 39 )  #MB
word length pattern: 0 3 4 0 0 0

> find.BSP(c(6,2), c(4,4,2))
defining words: ABCE ACDF ABDPQ ( 7 13 27 )  splitting/blocking words: AC ABD ( 5 11 )
(a), (b) and (c) : ( 28 46 ); ( 24 36 ); ( 12 12 )  #New
word length pattern: 0 3 4 0 0 0
defining words: ABCE ACDF ABPQ ( 7 13 19 )  splitting/blocking words: AC ABD ( 5 11 )
(a), (b) and (c) : ( 28 58 ); ( 25 49 ); ( 12 20 )  #MB
word length pattern: 0 5 0 2 0 0

> find.BSP(c(6,3), c(2,8,2))
defining words: ABCE ACDF ACPQ ABDPQ ( 7 13 21 27 )  splitting/blocking words: ABD ( 11 )
(a), (b) and (c) : ( 36 72 ); ( 35 71 ); ( 18 30 )  #New
word length pattern: 0 6 8 0 0 1 0
defining words: ABCE ACDF ABPQ BDPR ( 7 13 19 26 )  splitting/blocking words: ABD ( 11 )
(a), (b) and (c) : ( 36 90 ); ( 36 90 ); ( 18 42 )  #MB
word length pattern: 0 9 0 6 0 0 0

> find.BSP(c(7,2), c(2,8,2))
defining words: ABCE ACDF BCDG ABDPQ ( 7 13 14 27 )  splitting/blocking words: ABD ( 11 )
(a), (b) and (c) : ( 36 78 ); ( 35 77 ); ( 14 14 )  #New
word length pattern: 0 7 7 0 0 0 1
defining words: ABCE ACDF BCDG ABPQ ( 7 13 14 19 )  splitting/blocking words: ABD ( 11 )
(a), (b) and (c) : ( 36 96 ); ( 36 96 ); ( 14 26 )  #MB
word length pattern: 0 10 0 4 0 1 0

```

(5). Function `BSP()` is used to evaluate a design when, instead of defining and splitting/blocking words, defining and splitting/blocking columns are given.

(a) Consider the case $(n_1 n_2; b w s) = (1 7; 2 2 8)$. The design on Page 15 of Tsai (2016) has 15,26,29 as the defining columns and 22 as the splitting/blocking column. We use the following function to evaluate the design.

```
BSP(c(1,7),c(2,2,8), c(15,26,29), c(22))

mi in block stratum: 0
mi in whole-plot stratum: 0
mi in subplot stratum: 3 2 2 2 2 2 2 1 1 1 1 1 1 1 1 1 1 1 1 0
(a), (b) and (c) : ( 28 46 ); ( 28 46 ); ( 28 46 )
word length pattern: 0 3 4 0 0 0
MB word length pattern: 0 0 3 3 4 4 0 0 0 0 0 1
```

For $(n_1 n_2; b w s) = (1 7; 2 2 8)$, the design in MB's Table 1 has 15,23,30 as the defining columns and 27 as the splitting/blocking column. We have the following output.

```
> BSP(c(1,7),c(2,2,8), c(15,23,30), c(27)) #MB
mi in block stratum: 0
mi in whole-plot stratum: 1
mi in subplot stratum: 3 2 2 2 2 2 2 1 1 1 1 1 1 1 1 1 1 1 1 0 0
(a), (b) and (c) : ( 28 46 ); ( 28 46 ); ( 27 45 )
word length pattern: 0 3 4 0 0 0
MB word length pattern: 0 0 3 3 4 4 0 0 0 0 0 1
```

(b) Consider the case $(n_1 n_2; b w s) = (1 8; 2 2 8)$

```
> BSP(c(1,8), c(2,2,8), c(15,21,25,30), c(19)) #New
mi in block stratum: 0
mi in whole-plot stratum: 1
mi in subplot stratum: 4 2 2 2 2 2 2 2 2 2 2 2 1 1 1 1 1 1 1
(a), (b) and (c) : ( 36 72 ); ( 36 72 ); ( 35 71 )
word length pattern: 0 6 8 0 0 1 0
MB word length pattern: 0 0 6 4 8 8 0 0 0 0 1 4 0 0
```

```
> BSP(c(1,8), c(2,2,8), c(15,23,27,29), c(30)) #MB's design
mi in block stratum: 0
mi in whole-plot stratum: 4
mi in subplot stratum: 2 2 2 2 2 2 2 2 2 2 2 2 1 1 1 1 1 1 1
(a), (b) and (c) : ( 36 72 ); ( 36 72 ); ( 32 56 )
word length pattern: 0 6 8 0 0 1 0
MB word length pattern: 0 0 6 4 8 8 0 0 0 0 1 4 0 0
```

In each case, the new design is a better design, as discussed in Tsai (2016).

References

Tsai, P. W. (2016), "A study of two types of split-plot designs", *Journal of Quality Technology* 48, 44-53.