

Data analysis

Signal detection of spontaneous medical device reports over time

Ty Stanford et al.

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1 Set up

1.1 Packages

```
suppressPackageStartupMessages({  
  library("readr")  
  library("dplyr")  
  library("tidyr")  
  library("forcats")  
  library("purrr")  
  library("furrr")  
  library("lubridate") # way to handle dates better than default R way  
  library("tictoc")    # measure time elapsed in calcs  
  library("ggplot2")  
  library("ggrepel")  
  library("knitr")  
  library("gsDesign")  
  library("foreach")  
  library("arrow") # read/write parquet files  
})
```

Warning: package 'future' was built under R version 4.2.3

```
# NOTE: need to run first (only once, assumes devtools installed):  
# devtools::install_github("tystan/pharmsignal")  
library("pharmsignal") # signal detection algs  
  
# here are the functions written for these analyses  
# they will be shown in the *Appendix A*  
source("r/_funcs.R")
```

1.2 Parallel computation setup

```
# options(future.globals.maxSize = 500 * 1024 ^ 2) # = 500 MiB  
options(future.globals.maxSize = 1e3 * 1024 ^ 2) # = 1 GB
```

```

# furrr parallel workers/cores setup
# change `workers = 4` based on cores available in processor being used
plan(multisession, workers = 4)

### test parallel works
# test code from https://furrr.futureverse.org/
# sequential
tic()
dev_null <- map(c(2, 2, 2), ~Sys.sleep(.x))
toc() # ~6 sec

```

6.12 sec elapsed

```

# parallel: should be (roughly, plus overheads) a third of the time of sequential
tic()
dev_null <- future_map(c(2, 2, 2), ~Sys.sleep(.x))
toc() # ~2 sec

```

2.94 sec elapsed

```

# this only applies to the non-parallel (non-"future") operations
set.seed(1234)
# this seed can be set in future_map() etc for reproducible parallel comp seeds
furrr_seed1 <- furrr_options(seed = 5678)

```

1.3 Constants

```

# arbitrarily, let's go with minimum cell count of 3 (should be discussed!)
arbitrary_cell_min <- 1

```

1.4 Functions

```
get_sig_tab <- function(nA, nB, nC, nD, alpha = 0.05, n_mcmc = 1e+05) {  
  
  out_cols_of_interest <- c("est_name", "est_scale", "est", "ci_lo", "ci_hi")  
  sig_tab <- pharmsignal::bcpnn_mcmc_signal(nA, nB, nC, nD, alpha = alpha, n_mcmc = n_mcmc)  
  sig_tab <- sig_tab[, out_cols_of_interest]  
  # sig_tab <- bind_cols(tibble(mnth = mnth), sig_tab)  
  return(sig_tab)  
  
}  
  
get_sig_tab_over_time <- function(dat, alpha = 0.05, n_mcmc = 1e+05) {  
  
  n_tp <- nrow(dat)  
  
  sig_tab_over_time <-  
    foreach(i = 1:n_tp, .combine = bind_rows, .packages = "dplyr") %do% {  
      with(  
        dat,  
        get_sig_tab(  
          # mnth[i],  
          nA[i], nB[i], nC[i], nD[i],  
          alpha = alpha, n_mcmc = n_mcmc  
        )  
      )  
    }  
  
  return(sig_tab_over_time)  
  
}
```

1.5 Load data

```
sra_dat <- read_parquet("dat/sra_dat.parquet")
```

2 Analysis

2.1 BCPNN

```
sra_cum <-  
  sra_dat %>%  
  dplyr::filter(dat_type == "cumulative")  
  
# make data for each combination of params nested for purrr like processing  
sra_cum <-  
  sra_cum %>%  
  nest(data = c(mnth, nA, nB, nC, nD))  
  
# testing/example  
sra_cum$data[[1]]
```

```
# A tibble: 44 x 5  
  mnth      nA    nB    nC    nD  
  <chr>   <dbl> <dbl> <dbl> <dbl>  
1 2013-01     3     7     1     2  
2 2013-02     3     7     1     4  
3 2013-04     3     7     1     5  
4 2013-05     4    10     1     5  
5 2013-07     4    11     1     7  
6 2013-08     5    11     1     7  
7 2013-09     5    11     2     9  
8 2013-11     8    11     2     9  
9 2013-12     9    11     2     9  
10 2014-03     9    11     2    10  
# ... with 34 more rows
```

```
get_sig_tab_over_time(sra_cum$data[[1]])
```

	est_name	est_scale	est	ci_lo	ci_hi
1	bcpnn_mcmc	log2	-0.03155473	-1.008567292	0.5011816
2	bcpnn_mcmc	log2	0.14423111	-0.881461274	0.7657139
3	bcpnn_mcmc	log2	0.22225892	-0.819459061	0.8710605
4	bcpnn_mcmc	log2	0.16982567	-0.631979264	0.6843789
5	bcpnn_mcmc	log2	0.25877437	-0.558250981	0.7901982

6	bcpnn_mcmc	log2	0.28945108	-0.359347606	0.7778787
7	bcpnn_mcmc	log2	0.24275055	-0.534420549	0.7826432
8	bcpnn_mcmc	log2	0.31486271	-0.170553281	0.7465905
9	bcpnn_mcmc	log2	0.32253264	-0.103384822	0.7293047
10	bcpnn_mcmc	log2	0.36527691	-0.067632484	0.7879723
11	bcpnn_mcmc	log2	0.40661823	-0.034175482	0.8369273
12	bcpnn_mcmc	log2	0.31431083	-0.117133215	0.7114928
13	bcpnn_mcmc	log2	0.35146856	-0.088210721	0.7597945
14	bcpnn_mcmc	log2	0.32236631	-0.145358061	0.7374164
15	bcpnn_mcmc	log2	0.41872427	0.005768877	0.8060994
16	bcpnn_mcmc	log2	0.42043314	-0.019393480	0.8193463
17	bcpnn_mcmc	log2	0.51134622	0.122251507	0.8922410
18	bcpnn_mcmc	log2	0.48997255	0.090959174	0.8764148
19	bcpnn_mcmc	log2	0.44480211	0.225394906	0.6963962
20	bcpnn_mcmc	log2	0.46323762	0.241125813	0.7192630
21	bcpnn_mcmc	log2	0.45935820	0.241366445	0.7109985
22	bcpnn_mcmc	log2	0.44698469	0.231636145	0.6971812
23	bcpnn_mcmc	log2	0.42807282	0.211373941	0.6762936
24	bcpnn_mcmc	log2	0.40986367	0.191379522	0.6565181
25	bcpnn_mcmc	log2	0.40529856	0.198145687	0.6409685
26	bcpnn_mcmc	log2	0.40021287	0.201086693	0.6289638
27	bcpnn_mcmc	log2	0.38907653	0.206723284	0.6022235
28	bcpnn_mcmc	log2	0.42060041	0.232653673	0.6397847
29	bcpnn_mcmc	log2	0.43735676	0.243979517	0.6600397
30	bcpnn_mcmc	log2	0.45236046	0.256382608	0.6771721
31	bcpnn_mcmc	log2	0.43564014	0.249050440	0.6506842
32	bcpnn_mcmc	log2	0.42578750	0.240710552	0.6406515
33	bcpnn_mcmc	log2	0.41437069	0.240937476	0.6170948
34	bcpnn_mcmc	log2	0.40562061	0.238810116	0.5993361
35	bcpnn_mcmc	log2	0.40236748	0.238325445	0.5936070
36	bcpnn_mcmc	log2	0.40588775	0.250241656	0.5864129
37	bcpnn_mcmc	log2	0.38625000	0.238767885	0.5591967
38	bcpnn_mcmc	log2	0.37617897	0.235058228	0.5426621
39	bcpnn_mcmc	log2	0.38226092	0.243192323	0.5467141
40	bcpnn_mcmc	log2	0.36736813	0.235354142	0.5233362
41	bcpnn_mcmc	log2	0.36257011	0.233573242	0.5155766
42	bcpnn_mcmc	log2	0.35553674	0.229417693	0.5041057
43	bcpnn_mcmc	log2	0.34192103	0.220789503	0.4841618
44	bcpnn_mcmc	log2	0.33766947	0.219170482	0.4782721

```

### takes ~ 90 sec
tic()
sra_cum <-
  sra_cum %>%
  mutate(
    sig_tab =
      future_map(
        .x = data,
        .f = get_sig_tab_over_time,
        .options = furrr_seed1
      )
  )
toc()

```

112.24 sec elapsed

```

# check
sra_cum$sig_tab[[1]]

```

	est_name	est_scale	est	ci_lo	ci_hi
1	bcpnn_mcmc	log2	-0.03155473	-1.01361518	0.5008338
2	bcpnn_mcmc	log2	0.14423111	-0.87754425	0.7674485
3	bcpnn_mcmc	log2	0.22225892	-0.82925568	0.8768976
4	bcpnn_mcmc	log2	0.16982567	-0.62630109	0.6841108
5	bcpnn_mcmc	log2	0.25877437	-0.56115069	0.7903408
6	bcpnn_mcmc	log2	0.28945108	-0.36195188	0.7791041
7	bcpnn_mcmc	log2	0.24275055	-0.54147551	0.7865138
8	bcpnn_mcmc	log2	0.31486271	-0.16358209	0.7464793
9	bcpnn_mcmc	log2	0.32253264	-0.09929240	0.7367417
10	bcpnn_mcmc	log2	0.36527691	-0.06997405	0.7826290
11	bcpnn_mcmc	log2	0.40661823	-0.03559667	0.8371538
12	bcpnn_mcmc	log2	0.31431083	-0.12150818	0.7158149
13	bcpnn_mcmc	log2	0.35146856	-0.08877886	0.7628745
14	bcpnn_mcmc	log2	0.32236631	-0.15065287	0.7402075
15	bcpnn_mcmc	log2	0.41872427	0.00681964	0.8113143
16	bcpnn_mcmc	log2	0.42043314	-0.01475558	0.8214752
17	bcpnn_mcmc	log2	0.51134622	0.12474650	0.8952461
18	bcpnn_mcmc	log2	0.48997255	0.09205215	0.8748550
19	bcpnn_mcmc	log2	0.44480211	0.22642533	0.6997523
20	bcpnn_mcmc	log2	0.46323762	0.24023405	0.7193275

21	bcpnn_mcmc	log2	0.45935820	0.24200791	0.7127425
22	bcpnn_mcmc	log2	0.44698469	0.23167170	0.6944536
23	bcpnn_mcmc	log2	0.42807282	0.21226736	0.6772533
24	bcpnn_mcmc	log2	0.40986367	0.19249103	0.6557647
25	bcpnn_mcmc	log2	0.40529856	0.19777767	0.6411809
26	bcpnn_mcmc	log2	0.40021287	0.20072109	0.6295022
27	bcpnn_mcmc	log2	0.38907653	0.20484737	0.6038758
28	bcpnn_mcmc	log2	0.42060041	0.23274148	0.6415478
29	bcpnn_mcmc	log2	0.43735676	0.24612622	0.6588602
30	bcpnn_mcmc	log2	0.45236046	0.25818115	0.6760420
31	bcpnn_mcmc	log2	0.43564014	0.24895062	0.6536131
32	bcpnn_mcmc	log2	0.42578750	0.23965009	0.6400406
33	bcpnn_mcmc	log2	0.41437069	0.23951271	0.6165066
34	bcpnn_mcmc	log2	0.40562061	0.23923085	0.5994463
35	bcpnn_mcmc	log2	0.40236748	0.23938575	0.5932197
36	bcpnn_mcmc	log2	0.40588775	0.24969119	0.5872826
37	bcpnn_mcmc	log2	0.38625000	0.23831821	0.5599170
38	bcpnn_mcmc	log2	0.37617897	0.23431857	0.5426523
39	bcpnn_mcmc	log2	0.38226092	0.24172534	0.5480638
40	bcpnn_mcmc	log2	0.36736813	0.23617735	0.5236761
41	bcpnn_mcmc	log2	0.36257011	0.23406734	0.5142182
42	bcpnn_mcmc	log2	0.35553674	0.22958272	0.5048607
43	bcpnn_mcmc	log2	0.34192103	0.22109856	0.4837613
44	bcpnn_mcmc	log2	0.33766947	0.22005543	0.4760598

```
sra_cum_bcpnn <-
  sra_cum %>%
  unnest(cols = c(data, sig_tab)) %>%
  mutate(dte = as_date(paste0(mnth, "-01")))

sra_cum_bcpnn
```

```
# A tibble: 4,523 x 14
  grps      dat_t~1 thresh mnth      nA      nB      nC      nD est_n~2 est_s~3      est
  <chr>      <chr>      <dbl> <chr> <dbl> <dbl> <dbl> <dbl> <chr>      <chr>      <dbl>
1 pelvic_~ cumula~  0.01 2013~  3      7      1      2 bcpnn_~ log2      -0.0316
2 pelvic_~ cumula~  0.01 2013~  3      7      1      4 bcpnn_~ log2       0.144
3 pelvic_~ cumula~  0.01 2013~  3      7      1      5 bcpnn_~ log2       0.222
4 pelvic_~ cumula~  0.01 2013~  4     10      1      5 bcpnn_~ log2       0.170
5 pelvic_~ cumula~  0.01 2013~  4     11      1      7 bcpnn_~ log2       0.259
6 pelvic_~ cumula~  0.01 2013~  5     11      1      7 bcpnn_~ log2       0.289
```



```

7 pelvic_~ cumula~ 0.01 2013~ 5 11 2 9 bcpnn_~ log2 0.243
8 pelvic_~ cumula~ 0.01 2013~ 8 11 2 9 bcpnn_~ log2 0.315
9 pelvic_~ cumula~ 0.01 2013~ 9 11 2 9 bcpnn_~ log2 0.323
10 pelvic_~ cumula~ 0.01 2014~ 9 11 2 10 bcpnn_~ log2 0.365
# ... with 4,513 more rows, 3 more variables: ci_lo <dbl>, ci_hi <dbl>,
# dte <date>, and abbreviated variable names 1: dat_type, 2: est_name,
# 3: est_scale

```

```

# first signif
bcpnn_signif <-
  sra_cum_bcpnn %>%
  group_by(grps, dat_type, thresh) %>%
  dplyr::filter(ci_lo > 0) %>%
  arrange(dte) %>%
  dplyr::filter(row_number() == 1) %>%
  ungroup() %>%
  rename(dte_reach_sig = dte)

nrow(sra_cum_bcpnn)

```

[1] 4523

```

sra_cum_bcpnn <-
  left_join(
    sra_cum_bcpnn,
    bcpnn_signif %>% select(grps, dat_type, thresh, dte_reach_sig),
    c("grps", "dat_type", "thresh")
  )
nrow(sra_cum_bcpnn)

```

[1] 4523

```
sra_cum_bcpnn
```

```

# A tibble: 4,523 x 15
  grps    dat_t~1 thresh mnth    nA    nB    nC    nD est_n~2 est_s~3    est

```

	<chr>	<chr>	<dbl>	<chr>	<dbl>	<dbl>	<dbl>	<dbl>	<chr>	<chr>	<dbl>
1	pelvic_~	cumula~	0.01	2013~	3	7	1	2	bcpnn_~	log2	-0.0316
2	pelvic_~	cumula~	0.01	2013~	3	7	1	4	bcpnn_~	log2	0.144
3	pelvic_~	cumula~	0.01	2013~	3	7	1	5	bcpnn_~	log2	0.222
4	pelvic_~	cumula~	0.01	2013~	4	10	1	5	bcpnn_~	log2	0.170
5	pelvic_~	cumula~	0.01	2013~	4	11	1	7	bcpnn_~	log2	0.259
6	pelvic_~	cumula~	0.01	2013~	5	11	1	7	bcpnn_~	log2	0.289
7	pelvic_~	cumula~	0.01	2013~	5	11	2	9	bcpnn_~	log2	0.243
8	pelvic_~	cumula~	0.01	2013~	8	11	2	9	bcpnn_~	log2	0.315
9	pelvic_~	cumula~	0.01	2013~	9	11	2	9	bcpnn_~	log2	0.323
10	pelvic_~	cumula~	0.01	2014~	9	11	2	10	bcpnn_~	log2	0.365

```
# ... with 4,513 more rows, 4 more variables: ci_lo <dbl>, ci_hi <dbl>,
#   dte <date>, dte_reach_sig <date>, and abbreviated variable names
#   1: dat_type, 2: est_name, 3: est_scale
```

```
sra_cum_bcpnn <-
  sra_cum_bcpnn %>%
  mutate(
    dte_reach_sig = if_else(is.na(dte_reach_sig), as_date(today()), dte_reach_sig),
    reach_sig = dte >= dte_reach_sig
  )
```

2.2 BCPNN with mult comp adjust

2.3 MaxSPRT

3 Session information

```
sra_cum_bcpnn %>%  
  write_parquet(., sink = "out/sra_cum_bcpnn.parquet")
```

```
## close multisession workers by switching plan  
plan(sequential)
```

```
format(Sys.time(), '%d %b %Y')
```

```
[1] "19 Jun 2023"
```

```
Sys.info() %>% as.data.frame(.)
```

```
      .  
sysname      Windows  
release      10 x64  
version      build 19044  
nodename     DESKTOP-R5P5N23  
machine      x86-64  
login        ty  
user         ty  
effective_user ty
```

```
sessionInfo()
```

```
R version 4.2.2 (2022-10-31 ucrt)  
Platform: x86_64-w64-mingw32/x64 (64-bit)  
Running under: Windows 10 x64 (build 19044)
```

```
Matrix products: default
```

```
locale:
```

```
[1] LC_COLLATE=English_Australia.utf8  LC_CTYPE=English_Australia.utf8  
[3] LC_MONETARY=English_Australia.utf8 LC_NUMERIC=C  
[5] LC_TIME=English_Australia.utf8
```

attached base packages:

```
[1] stats      graphics  grDevices  utils      datasets  methods    base
```

other attached packages:

```
[1] pharmsignal_0.1.0 arrow_11.0.0.2   foreach_1.5.2    gsDesign_3.4.0
[5] knitr_1.42         ggrepel_0.9.3    ggplot2_3.4.1     tictoc_1.1
[9] lubridate_1.9.2    furr_0.3.1       future_1.32.0     purrr_1.0.1
[13] forcats_1.0.0      tidyr_1.3.0      dplyr_1.1.0       readr_2.1.4
```

loaded via a namespace (and not attached):

```
[1] tidyselect_1.2.0  xfun_0.37        listenv_0.9.0     splines_4.2.2
[5] lattice_0.20-45   colorspace_2.1-0 vctrs_0.5.2       generics_0.1.3
[9] htmltools_0.5.4   yaml_2.3.7        survival_3.4-0     MCMCpack_1.6-3
[13] utf8_1.2.3        rlang_1.0.6       pillar_1.8.1      glue_1.6.2
[17] withr_2.5.0        bit64_4.0.5       lifecycle_1.0.3   MatrixModels_0.5-1
[21] munsell_0.5.0      gtable_0.3.1      coda_0.19-4        codetools_0.2-18
[25] evaluate_0.20     SparseM_1.81      tzdb_0.3.0         fastmap_1.1.0
[29] quantreg_5.94     parallel_4.2.2    fansi_1.0.4        Rcpp_1.0.10
[33] xtable_1.8-4       scales_1.2.1      jsonlite_1.8.4     parallelly_1.34.0
[37] bit_4.0.5          mcmc_0.9-7        hms_1.1.2          digest_0.6.31
[41] grid_4.2.2         cli_3.6.0         tools_4.2.2        magrittr_2.0.3
[45] tibble_3.1.8       pkgconfig_2.0.3   Matrix_1.5-3       MASS_7.3-58.1
[49] ellipsis_0.3.2     timechange_0.2.0  assertthat_0.2.1   rmarkdown_2.20
[53] rstudioapi_0.14    iterators_1.0.14  R6_2.5.1           globals_0.16.2
[57] compiler_4.2.2
```