

# 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("ggthemes")  
  library("knitr")  
  library("gsDesign")  
  library("foreach")  
  library("arrow") # read/write parquet files  
})
```

Warning: package 'ggthemes' was built under R version 4.3.2

```
# 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")  
  
### NB: packages required that are used in above sourced file  
# Sequential  
# EmpiricalCalibration  
  
### Note setting `plan(sequential)` for Quarto doc generation,  
# errors occur otherwise  
plan(sequential)
```

```

# 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)
furrr_seed2 <- furrr_options(seed = 9012)
furrr_seed3 <- furrr_options(seed = 3456)
furrr_seed4 <- furrr_options(seed = 7890)

```

## 1.2 Constants

```

# arbitrarily, let's go with minimum cell count of 1 (will change based on context/application)
arbitrary_cell_min <- 1

```

## 1.3 Functions

```

# do 90% CI only with lower == one sided 0.05
get_sig_tab <- function(nA, nB, nC, nD, alpha = 0.10, method = "bcpnn", n_mcmc = 1e+05) {

  out_cols_of_interest <- c("est_name", "est_scale", "est", "alpha", "ci_lo") # "ci_hi" (d
  sig_tab <- NULL # initialise in scope
  if (method == "bcpnn") {
    sig_tab <- pharmsignal::bcpnn_mcmc_signal(nA, nB, nC, nD, alpha = alpha, n_mcmc = n_mcmc)
  } else if (method == "prp") {
    sig_tab <- pharmsignal::prp_signal(nA, nB, nC, nD, alpha = alpha)
  } else {
    stop("method for calculations unknown")
  }
  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 = 0.10, method = "bcpnn", 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, method = method, n_mcmc = n_mcmc
    )
  )
}

return(sig_tab_over_time)
}

# if it's multiple comparisons central need to sparing use alpha
get_mult_compare_adj_alpha <- function(dat, alpha = 0.1) {

  n_reports <- nrow(dat)

  information_frcs <- (1:n_reports) / n_reports

  ### alternatives:
  # spend_obj <- sfLDPocock(alpha = alpha, t = information_frcs, param = NULL)
  # spend_obj <- sfLDOF(alpha = alpha, t = information_frcs, param = NULL)
  spend_obj <- sfExponential(alpha = alpha, t = information_frcs, param = 0.5)

  # plot(1:n_reports, spend_obj$spend, main = "alpha spending func", xlab = "look")

  return(bind_cols(dat, adj_alpha = spend_obj$spend))
}

# same as get_sig_tab_over_time(), however, alpha assumed included as column in data
get_sig_tab_over_time_2 <- function(dat, method = "bcpnn", 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 = adj_alpha[i],
        method = method,
        n_mcmc = n_mcmc
      )
    )
  }

  return(sig_tab_over_time)
}

# test
data.frame(nA = 30, nB = 5512, nC = 41, nD = 17445, adj_alpha = 0.1) %>%
  get_sig_tab_over_time_2(.)

```

```

      est_name est_scale      est alpha    ci_lo
1 bcpnn_mcmc      log2 0.7942907    0.1 0.4317622

```

```

data.frame(nA = 30, nB = 5512, nC = 41, nD = 17445, adj_alpha = 0.1) %>%
  get_sig_tab_over_time_2(., method = "prrr")

```

```

      est_name est_scale      est alpha    ci_lo
1      prrr orig scale 2.308667    0.1 1.556277

```

```

2 ^ c(0.432304, 0.7942907) # similar to prr on ratio scale

```

```

[1] 1.349387 1.734225

```

```

log2(c(1.556277, 2.308667)) # similar to bcpnn on log2 scale

```

```

[1] 0.6380989 1.2070601

```

## 1.4 Load data

```
### monthly for testing
sra_dat <- read_parquet("dat/sra_dat.parquet")

### want this
cumul_qtrly_dat <- read_parquet("dat/cumul_qtrly_dat.parquet")

(thresholds <- sort(unique(sra_dat$thresh)))
```

```
[1] "0.010" "0.015" "0.020" "0.025" "0.030" "0.035" "0.040" "0.045" "0.050"
[10] "0.055" "0.060" "0.065" "0.070" "0.075" "0.080" "0.085" "0.090" "0.095"
[19] "0.100"
```

```
cumul_qtrly_dat
```

```
# A tibble: 1,707 x 8
```

	grps		dat_type	thresh	mnth	nA	nB	nC	nD
	<chr>		<chr>	<chr>	<chr>	<dbl>	<dbl>	<dbl>	<dbl>
1	(a)	pelvic_mesh v hernia_mesh	cumulative	0.010	2013~	3	7	1	4
2	(a)	pelvic_mesh v hernia_mesh	cumulative	0.010	2013~	4	10	1	5
3	(a)	pelvic_mesh v hernia_mesh	cumulative	0.010	2013~	5	11	2	9
4	(a)	pelvic_mesh v hernia_mesh	cumulative	0.010	2013~	9	11	2	9
5	(a)	pelvic_mesh v hernia_mesh	cumulative	0.010	2014~	9	11	2	10
6	(a)	pelvic_mesh v hernia_mesh	cumulative	0.010	2014~	10	12	3	12
7	(a)	pelvic_mesh v hernia_mesh	cumulative	0.010	2014~	12	14	5	19
8	(a)	pelvic_mesh v hernia_mesh	cumulative	0.010	2014~	30	15	7	24
9	(a)	pelvic_mesh v hernia_mesh	cumulative	0.010	2015~	31	15	7	25
10	(a)	pelvic_mesh v hernia_mesh	cumulative	0.010	2015~	31	15	7	25

```
# i 1,697 more rows
```

```
# continuity_chk <-
#   cumul_qtrly_dat %>%
#   mutate(
#     yr = as.integer(substr(mnth, 1, 4)),
#     qtr = as.integer(substr(mnth, 7, 7))
#   )
#
# with(
```

```
# continuity_chk,  
# table(  
#   yr,  
#   qtr,  
#   grps,  
#   thresh,  
#   useNA = "ifany"  
# )  
# )  
#  
# cumul_qtrly_dat %>%  
#   dplyr::filter(substr(grps, 1, 3) == "(b)", thresh == "0.040")
```

## 2 Analysis

### 2.1 BCPNN

```
sra_cum <-  
  cumul_qtrly_dat  
  
# make data for each combination of params nested for purrr like processing  
sra_cum <-  
  sra_cum %>%  
  nest(data = c(mnth, nA, nB, nC, nD))  
  
sra_cum2 <-  
  sra_dat %>%  
  dplyr::filter(dat_type == "cumulative") %>%  
  nest(data = c(mnth, nA, nB, nC, nD))  
  
# testing/example  
sra_cum$data[[9]] %>% print(., n = nrow(.))
```

```
# A tibble: 18 x 5  
  mnth      nA      nB      nC      nD  
  <chr>   <dbl> <dbl> <dbl> <dbl>  
1 2013-Q3     4    12     1    10  
2 2013-Q4     6    14     1    10  
3 2014-Q1     6    14     1    11  
4 2014-Q2     7    15     1    14  
5 2014-Q3     9    17     3    21  
6 2014-Q4    26    19     4    27  
7 2015-Q1    27    19     4    28  
8 2015-Q2    27    19     4    28  
9 2015-Q3    27    20     4    28  
10 2015-Q4    27    20     6    28  
11 2016-Q1    30    21     6    28  
12 2016-Q2    34    21     6    28  
13 2016-Q3    34    21     7    33  
14 2016-Q4    36    23     7    33  
15 2017-Q1    45    23     8    34  
16 2017-Q2    58    24     8    37  
17 2017-Q3    68    24     8    38
```



18 2017-Q4      77      25      8      38

```
sra_cum2$data[[9]] %>% print(., n = nrow(.))
```

# A tibble: 38 x 5

	mnth <chr>	nA <dbl>	nB <dbl>	nC <dbl>	nD <dbl>
1	2013-09	4	12	1	10
2	2013-11	6	13	1	10
3	2013-12	6	14	1	10
4	2014-03	6	14	1	11
5	2014-04	6	14	1	12
6	2014-05	7	15	1	13
7	2014-06	7	15	1	14
8	2014-07	7	15	2	15
9	2014-08	9	17	2	19
10	2014-09	9	17	3	21
11	2014-10	10	18	3	24
12	2014-11	10	18	3	26
13	2014-12	26	19	4	27
14	2015-01	26	19	4	28
15	2015-03	27	19	4	28
16	2015-09	27	20	4	28
17	2015-10	27	20	5	28
18	2015-11	27	20	6	28
19	2016-01	29	20	6	28
20	2016-03	30	21	6	28
21	2016-04	34	21	6	28
22	2016-07	34	21	6	30
23	2016-08	34	21	7	32
24	2016-09	34	21	7	33
25	2016-11	36	22	7	33
26	2016-12	36	23	7	33
27	2017-01	40	23	7	33
28	2017-02	43	23	7	33
29	2017-03	45	23	8	34
30	2017-04	50	24	8	36
31	2017-05	54	24	8	37
32	2017-06	58	24	8	37
33	2017-07	60	24	8	38
34	2017-08	66	24	8	38

```

35 2017-09      68      24      8      38
36 2017-10      71      24      8      38
37 2017-11      75      25      8      38
38 2017-12      77      25      8      38

```

```
get_sig_tab_over_time(sra_cum$data[[9]])
```

	est_name	est_scale	est	alpha	ci_lo
1	bcpnn_mcmc	log2	0.3778604	0.1	-0.27151962
2	bcpnn_mcmc	log2	0.3738316	0.1	-0.06911637
3	bcpnn_mcmc	log2	0.4150070	0.1	-0.03108974
4	bcpnn_mcmc	log2	0.5127001	0.1	0.11104715
5	bcpnn_mcmc	log2	0.4951774	0.1	0.06535270
6	bcpnn_mcmc	log2	0.5370558	0.1	0.33465530
7	bcpnn_mcmc	log2	0.5500767	0.1	0.35025304
8	bcpnn_mcmc	log2	0.5500767	0.1	0.35086730
9	bcpnn_mcmc	log2	0.5377636	0.1	0.33836792
10	bcpnn_mcmc	log2	0.4850381	0.1	0.28365055
11	bcpnn_mcmc	log2	0.4647627	0.1	0.27878002
12	bcpnn_mcmc	log2	0.4520794	0.1	0.27929577
13	bcpnn_mcmc	log2	0.5093946	0.1	0.32733841
14	bcpnn_mcmc	log2	0.4823956	0.1	0.30910135
15	bcpnn_mcmc	log2	0.4519206	0.1	0.30245415
16	bcpnn_mcmc	log2	0.4402744	0.1	0.30985519
17	bcpnn_mcmc	log2	0.4208998	0.1	0.30189089
18	bcpnn_mcmc	log2	0.3915001	0.1	0.28170266

```

### for i5-8400/48GB 2133mhz memory
# takes ~ 90 sec for monthly
# takes ~ 40 sec for quarterly
### divide by a fair bit for r9-5900X
tic()
sra_cum <-
  sra_cum %>%
  mutate(
    sig_tab =
      future_map(
        .x = data,
        .f = get_sig_tab_over_time,
        .options = furrr_seed1
      )
  )

```

```
)
toc()
```

82.47 sec elapsed

```
# check
sra_cum$sig_tab[[9]]
```

	est_name	est_scale	est	alpha	ci_lo
1	bcpnn_mcmc	log2	0.3778604	0.1	-0.27857272
2	bcpnn_mcmc	log2	0.3738316	0.1	-0.06530536
3	bcpnn_mcmc	log2	0.4150070	0.1	-0.03003815
4	bcpnn_mcmc	log2	0.5127001	0.1	0.11613920
5	bcpnn_mcmc	log2	0.4951774	0.1	0.06855644
6	bcpnn_mcmc	log2	0.5370558	0.1	0.33396400
7	bcpnn_mcmc	log2	0.5500767	0.1	0.34967589
8	bcpnn_mcmc	log2	0.5500767	0.1	0.35032689
9	bcpnn_mcmc	log2	0.5377636	0.1	0.33806521
10	bcpnn_mcmc	log2	0.4850381	0.1	0.28283536
11	bcpnn_mcmc	log2	0.4647627	0.1	0.27663489
12	bcpnn_mcmc	log2	0.4520794	0.1	0.28009528
13	bcpnn_mcmc	log2	0.5093946	0.1	0.33020338
14	bcpnn_mcmc	log2	0.4823956	0.1	0.31108265
15	bcpnn_mcmc	log2	0.4519206	0.1	0.30188642
16	bcpnn_mcmc	log2	0.4402744	0.1	0.30898281
17	bcpnn_mcmc	log2	0.4208998	0.1	0.30167104
18	bcpnn_mcmc	log2	0.3915001	0.1	0.28215524

```
sra_cum_bcpnn <-
sra_cum %>%
unnest(cols = c(data, sig_tab)) %>%
mutate(
  # dte = as_date(paste0(mnth, "-01"))
  dte =
    as_date(paste0(
      substr(mnth, 1, 5),
      sprintf("%02.0f", (as.integer(substr(mnth, 7, 7)) - 1) * 3 + 1),
      "-01"
    ))
))
```

```
)
```

```
sra_cum_bcpnn
```

```
# A tibble: 1,707 x 14
```

	grps	dat_type	thresh	mnth	nA	nB	nC	nD	est_name	est_scale	est	
	<chr>	<chr>	<chr>	<chr>	<dbl>	<dbl>	<dbl>	<dbl>	<chr>	<chr>	<dbl>	
1	(a)	p~	cumulat~	0.010	2013~	3	7	1	4	bcpnn_m~	log2	0.144
2	(a)	p~	cumulat~	0.010	2013~	4	10	1	5	bcpnn_m~	log2	0.170
3	(a)	p~	cumulat~	0.010	2013~	5	11	2	9	bcpnn_m~	log2	0.243
4	(a)	p~	cumulat~	0.010	2013~	9	11	2	9	bcpnn_m~	log2	0.323
5	(a)	p~	cumulat~	0.010	2014~	9	11	2	10	bcpnn_m~	log2	0.365
6	(a)	p~	cumulat~	0.010	2014~	10	12	3	12	bcpnn_m~	log2	0.351
7	(a)	p~	cumulat~	0.010	2014~	12	14	5	19	bcpnn_m~	log2	0.420
8	(a)	p~	cumulat~	0.010	2014~	30	15	7	24	bcpnn_m~	log2	0.445
9	(a)	p~	cumulat~	0.010	2015~	31	15	7	25	bcpnn_m~	log2	0.459
10	(a)	p~	cumulat~	0.010	2015~	31	15	7	25	bcpnn_m~	log2	0.459

```
# i 1,697 more rows
```

```
# i 3 more variables: alpha <dbl>, ci_lo <dbl>, dte <date>
```

```
# 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] 1707
```

```
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] 1707
```

```
sra_cum_bcpnn
```

```
# A tibble: 1,707 x 15
```

	grps	dat_type	thresh	mnth	nA	nB	nC	nD	est_name	est_scale	est	
	<chr>	<chr>	<chr>	<chr>	<dbl>	<dbl>	<dbl>	<dbl>	<chr>	<chr>	<dbl>	
1	(a)	p~	cumulat~	0.010	2013~	3	7	1	4	bcpnn_m~	log2	0.144
2	(a)	p~	cumulat~	0.010	2013~	4	10	1	5	bcpnn_m~	log2	0.170
3	(a)	p~	cumulat~	0.010	2013~	5	11	2	9	bcpnn_m~	log2	0.243
4	(a)	p~	cumulat~	0.010	2013~	9	11	2	9	bcpnn_m~	log2	0.323
5	(a)	p~	cumulat~	0.010	2014~	9	11	2	10	bcpnn_m~	log2	0.365
6	(a)	p~	cumulat~	0.010	2014~	10	12	3	12	bcpnn_m~	log2	0.351
7	(a)	p~	cumulat~	0.010	2014~	12	14	5	19	bcpnn_m~	log2	0.420
8	(a)	p~	cumulat~	0.010	2014~	30	15	7	24	bcpnn_m~	log2	0.445
9	(a)	p~	cumulat~	0.010	2015~	31	15	7	25	bcpnn_m~	log2	0.459
10	(a)	p~	cumulat~	0.010	2015~	31	15	7	25	bcpnn_m~	log2	0.459

```
# i 1,697 more rows
```

```
# i 4 more variables: alpha <dbl>, ci_lo <dbl>, dte <date>,
```

```
#   dte_reach_sig <date>
```

```

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
  )

```

```

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

```

## 2.2 BCPNN with mult comp adjust

```
# sra_cum <-  
#   sra_dat %>%  
#   dplyr::filter(dat_type == "cumulative")  
sra_cum <-  
  cumul_qtrly_dat  
  
sra_cum <-  
  sra_cum %>%  
  nest(data = c(mnth, nA, nB, nC, nD))  
  
# test get_mult_compare_adj_alpha()  
get_mult_compare_adj_alpha(sra_cum$data[[1]])
```

```
# A tibble: 18 x 6  
  mnth      nA      nB      nC      nD adj_alpha  
  <chr>   <dbl> <dbl> <dbl> <dbl>   <dbl>  
1 2013-Q3     4    12     1    10 0.0000572  
2 2013-Q4     5    15     1    10 0.001  
3 2014-Q1     5    15     1    11 0.00355  
4 2014-Q2     6    16     1    14 0.00756  
5 2014-Q3     8    18     3    21 0.0127  
6 2014-Q4    25    20     4    27 0.0185  
7 2015-Q1    26    20     4    28 0.0249  
8 2015-Q2    26    20     4    28 0.0316  
9 2015-Q3    26    21     4    28 0.0385  
10 2015-Q4    26    21     4    30 0.0455  
11 2016-Q1    29    22     4    30 0.0526  
12 2016-Q2    33    22     4    30 0.0596  
13 2016-Q3    33    22     5    35 0.0666  
14 2016-Q4    35    24     5    35 0.0735  
15 2017-Q1    44    24     6    36 0.0803  
16 2017-Q2    57    25     6    39 0.0870  
17 2017-Q3    67    25     6    40 0.0935  
18 2017-Q4    76    26     6    40 0.1
```

```
get_sig_tab_over_time_2(get_mult_compare_adj_alpha(sra_cum$data[[11]]))
```

	est_name	est_scale	est	alpha	ci_lo
1	bcpnn_mcmc	log2	0.3778604	5.719516e-05	-2.7775539
2	bcpnn_mcmc	log2	0.3314299	1.000000e-03	-1.3130928
3	bcpnn_mcmc	log2	0.3719247	3.552305e-03	-0.9735553
4	bcpnn_mcmc	log2	0.4794164	7.562748e-03	-0.4878578
5	bcpnn_mcmc	log2	0.4505463	1.266582e-02	-0.3737733
6	bcpnn_mcmc	log2	0.5291089	1.853315e-02	0.2334080
7	bcpnn_mcmc	log2	0.5426635	2.491337e-02	0.2676446
8	bcpnn_mcmc	log2	0.5426635	3.162278e-02	0.2787698
9	bcpnn_mcmc	log2	0.5303613	3.852888e-02	0.2786349
10	bcpnn_mcmc	log2	0.5654356	4.553645e-02	0.3168146
11	bcpnn_mcmc	log2	0.5392350	5.257699e-02	0.3160579
12	bcpnn_mcmc	log2	0.5198015	5.960122e-02	0.3204447
13	bcpnn_mcmc	log2	0.5742360	6.657378e-02	0.3693870
14	bcpnn_mcmc	log2	0.5445855	7.346941e-02	0.3548644
15	bcpnn_mcmc	log2	0.5026231	8.027030e-02	0.3392577
16	bcpnn_mcmc	log2	0.4817174	8.696406e-02	0.3429426
17	bcpnn_mcmc	log2	0.4572070	9.354240e-02	0.3327737
18	bcpnn_mcmc	log2	0.4241559	1.000000e-01	0.3113718

```
get_sig_tab_over_time(sra_cum$data[[11]])
```

	est_name	est_scale	est	alpha	ci_lo
1	bcpnn_mcmc	log2	0.3778604	0.1	-0.27276457
2	bcpnn_mcmc	log2	0.3314299	0.1	-0.18935042
3	bcpnn_mcmc	log2	0.3719247	0.1	-0.15817042
4	bcpnn_mcmc	log2	0.4794164	0.1	0.01891085
5	bcpnn_mcmc	log2	0.4505463	0.1	-0.02705216
6	bcpnn_mcmc	log2	0.5291089	0.1	0.32291402
7	bcpnn_mcmc	log2	0.5426635	0.1	0.33973052
8	bcpnn_mcmc	log2	0.5426635	0.1	0.33969117
9	bcpnn_mcmc	log2	0.5303613	0.1	0.32927472
10	bcpnn_mcmc	log2	0.5654356	0.1	0.36001100
11	bcpnn_mcmc	log2	0.5392350	0.1	0.34945147
12	bcpnn_mcmc	log2	0.5198015	0.1	0.34340898
13	bcpnn_mcmc	log2	0.5742360	0.1	0.38955969
14	bcpnn_mcmc	log2	0.5445855	0.1	0.36988520
15	bcpnn_mcmc	log2	0.5026231	0.1	0.34910679

```

16 bcpnn_mcmc      log2 0.4817174    0.1  0.34742783
17 bcpnn_mcmc      log2 0.4572070    0.1  0.33516223
18 bcpnn_mcmc      log2 0.4241559    0.1  0.31176932

```

```

tic()
sra_cum <-
  sra_cum %>%
  mutate(
    data =
      map(
        .x = data,
        .f = get_mult_compare_adj_alpha
      )
  )
toc()

```

0.03 sec elapsed

```

# test
sra_cum$data[[11]] # check adj_alpha added as column in data

```

```

# A tibble: 18 x 6
  mnth      nA    nB    nC    nD adj_alpha
<chr>   <dbl> <dbl> <dbl> <dbl>   <dbl>
1 2013-Q3     4    12     1    10 0.0000572
2 2013-Q4     5    15     1    10 0.001
3 2014-Q1     5    15     1    11 0.00355
4 2014-Q2     6    16     1    14 0.00756
5 2014-Q3     8    18     3    21 0.0127
6 2014-Q4    25    20     4    27 0.0185
7 2015-Q1    26    20     4    28 0.0249
8 2015-Q2    26    20     4    28 0.0316
9 2015-Q3    26    21     4    28 0.0385
10 2015-Q4    26    21     4    30 0.0455
11 2016-Q1    29    22     4    30 0.0526
12 2016-Q2    33    22     4    30 0.0596
13 2016-Q3    33    22     5    35 0.0666
14 2016-Q4    35    24     5    35 0.0735
15 2017-Q1    44    24     6    36 0.0803
16 2017-Q2    57    25     6    39 0.0870

```



```
17 2017-Q3      67      25      6      40 0.0935
18 2017-Q4      76      26      6      40 0.1
```

```
### takes ~ 40 sec (i5-8400 6c/6t)
### takes ~ 55 sec on laptop (i5 8th gen 4c/8t)
### takes ~ 10 sec (R9-5900X 12c/24t)
tic()
sra_cum <-
  sra_cum %>%
  mutate(
    sig_tab =
      future_map(
        .x = data,
        .f = get_sig_tab_over_time_2, # the alpha in data version
        .options = furrr_seed1
      )
  )
toc()
```

82.72 sec elapsed

```
# check
sra_cum$sig_tab[[11]]
```

	est_name	est_scale	est	alpha	ci_lo
1	bcpnn_mcmc	log2	0.3778604	5.719516e-05	-2.4252741
2	bcpnn_mcmc	log2	0.3314299	1.000000e-03	-1.2692260
3	bcpnn_mcmc	log2	0.3719247	3.552305e-03	-0.9503514
4	bcpnn_mcmc	log2	0.4794164	7.562748e-03	-0.4816096
5	bcpnn_mcmc	log2	0.4505463	1.266582e-02	-0.3786932
6	bcpnn_mcmc	log2	0.5291089	1.853315e-02	0.2341781
7	bcpnn_mcmc	log2	0.5426635	2.491337e-02	0.2692897
8	bcpnn_mcmc	log2	0.5426635	3.162278e-02	0.2796775
9	bcpnn_mcmc	log2	0.5303613	3.852888e-02	0.2788266
10	bcpnn_mcmc	log2	0.5654356	4.553645e-02	0.3196894
11	bcpnn_mcmc	log2	0.5392350	5.257699e-02	0.3167301
12	bcpnn_mcmc	log2	0.5198015	5.960122e-02	0.3204761
13	bcpnn_mcmc	log2	0.5742360	6.657378e-02	0.3686319
14	bcpnn_mcmc	log2	0.5445855	7.346941e-02	0.3541094
15	bcpnn_mcmc	log2	0.5026231	8.027030e-02	0.3392261

```

16 bcpnn_mcmc      log2 0.4817174 8.696406e-02 0.3425696
17 bcpnn_mcmc      log2 0.4572070 9.354240e-02 0.3331946
18 bcpnn_mcmc      log2 0.4241559 1.000000e-01 0.3126421

```

```

sra_cum_bcpnn_mc_adj <-
  sra_cum %>%
  unnest(cols = c(data, sig_tab)) %>%
  mutate(
    # dte = as_date(paste0(mnth, "-01"))
    dte =
      as_date(paste0(
        substr(mnth, 1, 5),
        sprintf("%02.0f", (as.integer(substr(mnth, 7, 7)) - 1) * 3 + 1),
        "-01"
      ))
  )

sra_cum_bcpnn_mc_adj

```

```
# A tibble: 1,707 x 15
```

	grps	dat_type	thresh	mnth	nA	nB	nC	nD	adj_alpha	est_name
	<chr>	<chr>	<chr>	<chr>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<chr>
1	(a) pelvic_~	cumulat~	0.010	2013~	3	7	1	4	0.0000337	bcpnn_m~
2	(a) pelvic_~	cumulat~	0.010	2013~	4	10	1	5	0.000688	bcpnn_m~
3	(a) pelvic_~	cumulat~	0.010	2013~	5	11	2	9	0.00262	bcpnn_m~
4	(a) pelvic_~	cumulat~	0.010	2013~	9	11	2	9	0.00581	bcpnn_m~
5	(a) pelvic_~	cumulat~	0.010	2014~	9	11	2	10	0.01	bcpnn_m~
6	(a) pelvic_~	cumulat~	0.010	2014~	10	12	3	12	0.0149	bcpnn_m~
7	(a) pelvic_~	cumulat~	0.010	2014~	12	14	5	19	0.0204	bcpnn_m~
8	(a) pelvic_~	cumulat~	0.010	2014~	30	15	7	24	0.0262	bcpnn_m~
9	(a) pelvic_~	cumulat~	0.010	2015~	31	15	7	25	0.0323	bcpnn_m~
10	(a) pelvic_~	cumulat~	0.010	2015~	31	15	7	25	0.0385	bcpnn_m~

```
# i 1,697 more rows
```

```
# i 5 more variables: est_scale <chr>, est <dbl>, alpha <dbl>, ci_lo <dbl>,
```

```
# dte <date>
```

```

with(sra_cum_bcpnn_mc_adj, table(dte, mnth, useNA = "ifany")) %>%
  as.data.frame() %>%
  dplyr::filter(Freq > 0) %>%
  arrange(mnth, dte)

```

	dte	mnth	Freq
1	2012-04-01	2012-Q2	16
2	2012-07-01	2012-Q3	16
3	2012-10-01	2012-Q4	37
4	2013-01-01	2013-Q1	41
5	2013-04-01	2013-Q2	51
6	2013-07-01	2013-Q3	71
7	2013-10-01	2013-Q4	71
8	2014-01-01	2014-Q1	71
9	2014-04-01	2014-Q2	71
10	2014-07-01	2014-Q3	81
11	2014-10-01	2014-Q4	89
12	2015-01-01	2015-Q1	89
13	2015-04-01	2015-Q2	89
14	2015-07-01	2015-Q3	89
15	2015-10-01	2015-Q4	89
16	2016-01-01	2016-Q1	89
17	2016-04-01	2016-Q2	89
18	2016-07-01	2016-Q3	89
19	2016-10-01	2016-Q4	89
20	2017-01-01	2017-Q1	95
21	2017-04-01	2017-Q2	95
22	2017-07-01	2017-Q3	95
23	2017-10-01	2017-Q4	95

```
# first signif
bcpnn_mc_adj_signif <-
  sra_cum_bcpnn_mc_adj %>%
  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_mc_adj)
```

```
[1] 1707
```

```
sra_cum_bcpnn_mc_adj <-
  left_join(
    sra_cum_bcpnn_mc_adj,
    bcpnn_mc_adj_signif %>% select(grps, dat_type, thresh, dte_reach_sig),
    c("grps", "dat_type", "thresh")
  )
nrow(sra_cum_bcpnn_mc_adj)
```

[1] 1707

```
sra_cum_bcpnn_mc_adj
```

# A tibble: 1,707 x 16

	grps	dat_type	thresh	mnth	nA	nB	nC	nD	adj_alpha	est_name
	<chr>	<chr>	<chr>	<chr>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<chr>
1	(a) pelvic_~	cumulat~	0.010	2013~	3	7	1	4	0.0000337	bcpnn_m~
2	(a) pelvic_~	cumulat~	0.010	2013~	4	10	1	5	0.000688	bcpnn_m~
3	(a) pelvic_~	cumulat~	0.010	2013~	5	11	2	9	0.00262	bcpnn_m~
4	(a) pelvic_~	cumulat~	0.010	2013~	9	11	2	9	0.00581	bcpnn_m~
5	(a) pelvic_~	cumulat~	0.010	2014~	9	11	2	10	0.01	bcpnn_m~
6	(a) pelvic_~	cumulat~	0.010	2014~	10	12	3	12	0.0149	bcpnn_m~
7	(a) pelvic_~	cumulat~	0.010	2014~	12	14	5	19	0.0204	bcpnn_m~
8	(a) pelvic_~	cumulat~	0.010	2014~	30	15	7	24	0.0262	bcpnn_m~
9	(a) pelvic_~	cumulat~	0.010	2015~	31	15	7	25	0.0323	bcpnn_m~
10	(a) pelvic_~	cumulat~	0.010	2015~	31	15	7	25	0.0385	bcpnn_m~

# i 1,697 more rows

# i 6 more variables: est\_scale <chr>, est <dbl>, alpha <dbl>, ci\_lo <dbl>,

# dte <date>, dte\_reach\_sig <date>

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

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

## 2.3 PRR with mult comp adjust

```
# sra_cum <-  
#   sra_dat %>%  
#   dplyr::filter(dat_type == "cumulative")  
sra_cum <-  
  cumul_qtrly_dat  
  
sra_cum <-  
  sra_cum %>%  
  nest(data = c(mnth, nA, nB, nC, nD))  
  
# test  
get_mult_compare_adj_alpha(sra_cum$data[[1]])
```

```
# A tibble: 18 x 6  
  mnth      nA      nB      nC      nD adj_alpha  
  <chr>   <dbl> <dbl> <dbl> <dbl>   <dbl>  
1 2013-Q3     4    12     1    10 0.0000572  
2 2013-Q4     5    15     1    10 0.001  
3 2014-Q1     5    15     1    11 0.00355  
4 2014-Q2     6    16     1    14 0.00756  
5 2014-Q3     8    18     3    21 0.0127  
6 2014-Q4    25    20     4    27 0.0185  
7 2015-Q1    26    20     4    28 0.0249  
8 2015-Q2    26    20     4    28 0.0316  
9 2015-Q3    26    21     4    28 0.0385  
10 2015-Q4    26    21     4    30 0.0455  
11 2016-Q1    29    22     4    30 0.0526  
12 2016-Q2    33    22     4    30 0.0596  
13 2016-Q3    33    22     5    35 0.0666  
14 2016-Q4    35    24     5    35 0.0735  
15 2017-Q1    44    24     6    36 0.0803  
16 2017-Q2    57    25     6    39 0.0870  
17 2017-Q3    67    25     6    40 0.0935  
18 2017-Q4    76    26     6    40 0.1
```

```
get_sig_tab_over_time_2(get_mult_compare_adj_alpha(sra_cum$data[[1]]))
```

	est_name	est_scale	est	alpha	ci_lo
1	bcpnn_mcmc	log2	0.3778604	5.719516e-05	-2.3823226
2	bcpnn_mcmc	log2	0.3314299	1.000000e-03	-1.3454188
3	bcpnn_mcmc	log2	0.3719247	3.552305e-03	-0.9331261
4	bcpnn_mcmc	log2	0.4794164	7.562748e-03	-0.4892387
5	bcpnn_mcmc	log2	0.4505463	1.266582e-02	-0.3800232
6	bcpnn_mcmc	log2	0.5291089	1.853315e-02	0.2342780
7	bcpnn_mcmc	log2	0.5426635	2.491337e-02	0.2666768
8	bcpnn_mcmc	log2	0.5426635	3.162278e-02	0.2782165
9	bcpnn_mcmc	log2	0.5303613	3.852888e-02	0.2785088
10	bcpnn_mcmc	log2	0.5654356	4.553645e-02	0.3178754
11	bcpnn_mcmc	log2	0.5392350	5.257699e-02	0.3159453
12	bcpnn_mcmc	log2	0.5198015	5.960122e-02	0.3207910
13	bcpnn_mcmc	log2	0.5742360	6.657378e-02	0.3682926
14	bcpnn_mcmc	log2	0.5445855	7.346941e-02	0.3535223
15	bcpnn_mcmc	log2	0.5026231	8.027030e-02	0.3395410
16	bcpnn_mcmc	log2	0.4817174	8.696406e-02	0.3430154
17	bcpnn_mcmc	log2	0.4572070	9.354240e-02	0.3327655
18	bcpnn_mcmc	log2	0.4241559	1.000000e-01	0.3124256

```
get_sig_tab_over_time_2(get_mult_compare_adj_alpha(sra_cum$data[[1]]), method = "prrr")
```

	est_name	est_scale	est	alpha	ci_lo
1	prrr orig	scale	2.750000	5.719516e-05	0.04066616
2	prrr orig	scale	2.750000	1.000000e-03	0.09303769
3	prrr orig	scale	3.000000	3.552305e-03	0.14772511
4	prrr orig	scale	4.090909	7.562748e-03	0.26339441
5	prrr orig	scale	2.461538	1.266582e-02	0.53132229
6	prrr orig	scale	4.305556	1.853315e-02	1.37316395
7	prrr orig	scale	4.521739	2.491337e-02	1.52286634
8	prrr orig	scale	4.521739	3.162278e-02	1.59362698
9	prrr orig	scale	4.425532	3.852888e-02	1.61983333
10	prrr orig	scale	4.702128	4.553645e-02	1.77347101
11	prrr orig	scale	4.833333	5.257699e-02	1.88692159
12	prrr orig	scale	5.100000	5.960122e-02	2.05551981
13	prrr orig	scale	4.800000	6.657378e-02	2.17067746
14	prrr orig	scale	4.745763	7.346941e-02	2.19025371
15	prrr orig	scale	4.529412	8.027030e-02	2.29602631

```

16      prr orig scale 5.213415 8.696406e-02 2.68795062
17      prr orig scale 5.583333 9.354240e-02 2.92265162
18      prr orig scale 5.712418 1.000000e-01 3.03209290

```

```
get_sig_tab_over_time(sra_cum$data[[11]], method = "prr")
```

	est_name	est_scale	est	alpha	ci_lo
1	prr orig scale	2.750000	0.1	0.4912212	
2	prr orig scale	2.750000	0.1	0.5060331	
3	prr orig scale	3.000000	0.1	0.5487100	
4	prr orig scale	4.090909	0.1	0.7555189	
5	prr orig scale	2.461538	0.1	0.8951429	
6	prr orig scale	4.305556	0.1	1.9379924	
7	prr orig scale	4.521739	0.1	2.0354629	
8	prr orig scale	4.521739	0.1	2.0354629	
9	prr orig scale	4.425532	0.1	1.9905948	
10	prr orig scale	4.702128	0.1	2.1084471	
11	prr orig scale	4.833333	0.1	2.1757613	
12	prr orig scale	5.100000	0.1	2.3065749	
13	prr orig scale	4.800000	0.1	2.3563014	
14	prr orig scale	4.745763	0.1	2.3318890	
15	prr orig scale	4.529412	0.1	2.3909246	
16	prr orig scale	5.213415	0.1	2.7583297	
17	prr orig scale	5.583333	0.1	2.9591422	
18	prr orig scale	5.712418	0.1	3.0320929	

```

tic()
sra_cum <-
  sra_cum %>%
  mutate(
    data =
      map(
        .x = data,
        .f = get_mult_compare_adj_alpha
      )
  )
toc()

```

0.03 sec elapsed

```
# test
sra_cum$data[[11]] # check adj_alpha added as column in data
```

```
# A tibble: 18 x 6
  mnth      nA      nB      nC      nD adj_alpha
  <chr>   <dbl> <dbl> <dbl> <dbl>   <dbl>
1 2013-Q3     4    12     1    10 0.0000572
2 2013-Q4     5    15     1    10 0.001
3 2014-Q1     5    15     1    11 0.00355
4 2014-Q2     6    16     1    14 0.00756
5 2014-Q3     8    18     3    21 0.0127
6 2014-Q4    25    20     4    27 0.0185
7 2015-Q1    26    20     4    28 0.0249
8 2015-Q2    26    20     4    28 0.0316
9 2015-Q3    26    21     4    28 0.0385
10 2015-Q4    26    21     4    30 0.0455
11 2016-Q1    29    22     4    30 0.0526
12 2016-Q2    33    22     4    30 0.0596
13 2016-Q3    33    22     5    35 0.0666
14 2016-Q4    35    24     5    35 0.0735
15 2017-Q1    44    24     6    36 0.0803
16 2017-Q2    57    25     6    39 0.0870
17 2017-Q3    67    25     6    40 0.0935
18 2017-Q4    76    26     6    40 0.1
```

```
get_sig_tab_over_time_2_prr <- function(dat) {
  get_sig_tab_over_time_2(dat, method = "prrr")
}
```

```
### takes ~2 sec on laptop (i5 8th gen 4c/8t)
tic()
sra_cum <-
  sra_cum %>%
  mutate(
    sig_tab =
      future_map(
        .x = data,
        .f = get_sig_tab_over_time_2_prr, # the alpha in data version
        .options = furrr_seed1
      )
  )
```



```
)
toc()
```

2.16 sec elapsed

```
# check
sra_cum$sig_tab[[11]]
```

	est_name	est_scale	est	alpha	ci_lo	
1	pr	orig	scale	2.750000	5.719516e-05	0.04066616
2	pr	orig	scale	2.750000	1.000000e-03	0.09303769
3	pr	orig	scale	3.000000	3.552305e-03	0.14772511
4	pr	orig	scale	4.090909	7.562748e-03	0.26339441
5	pr	orig	scale	2.461538	1.266582e-02	0.53132229
6	pr	orig	scale	4.305556	1.853315e-02	1.37316395
7	pr	orig	scale	4.521739	2.491337e-02	1.52286634
8	pr	orig	scale	4.521739	3.162278e-02	1.59362698
9	pr	orig	scale	4.425532	3.852888e-02	1.61983333
10	pr	orig	scale	4.702128	4.553645e-02	1.77347101
11	pr	orig	scale	4.833333	5.257699e-02	1.88692159
12	pr	orig	scale	5.100000	5.960122e-02	2.05551981
13	pr	orig	scale	4.800000	6.657378e-02	2.17067746
14	pr	orig	scale	4.745763	7.346941e-02	2.19025371
15	pr	orig	scale	4.529412	8.027030e-02	2.29602631
16	pr	orig	scale	5.213415	8.696406e-02	2.68795062
17	pr	orig	scale	5.583333	9.354240e-02	2.92265162
18	pr	orig	scale	5.712418	1.000000e-01	3.03209290

```
sra_cum_prr_mc_adj <-
  sra_cum %>%
  unnest(cols = c(data, sig_tab)) %>%
  mutate(
    # dte = as_date(paste0(mnth, "-01"))
    dte =
      as_date(paste0(
        substr(mnth, 1, 5),
        sprintf("%02.0f", (as.integer(substr(mnth, 7, 7)) - 1) * 3 + 1),
        "-01"
      ))
  )
```

)

sra\_cum\_prr\_mc\_adj

# A tibble: 1,707 x 15

	grps	dat_type	thresh	mnth	nA	nB	nC	nD	adj_alpha	est_name
	<chr>	<chr>	<chr>	<chr>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<chr>
1	(a) pelvic_~	cumulat~	0.010	2013~	3	7	1	4	0.0000337	prr
2	(a) pelvic_~	cumulat~	0.010	2013~	4	10	1	5	0.000688	prr
3	(a) pelvic_~	cumulat~	0.010	2013~	5	11	2	9	0.00262	prr
4	(a) pelvic_~	cumulat~	0.010	2013~	9	11	2	9	0.00581	prr
5	(a) pelvic_~	cumulat~	0.010	2014~	9	11	2	10	0.01	prr
6	(a) pelvic_~	cumulat~	0.010	2014~	10	12	3	12	0.0149	prr
7	(a) pelvic_~	cumulat~	0.010	2014~	12	14	5	19	0.0204	prr
8	(a) pelvic_~	cumulat~	0.010	2014~	30	15	7	24	0.0262	prr
9	(a) pelvic_~	cumulat~	0.010	2015~	31	15	7	25	0.0323	prr
10	(a) pelvic_~	cumulat~	0.010	2015~	31	15	7	25	0.0385	prr

# i 1,697 more rows  
# i 5 more variables: est\_scale <chr>, est <dbl>, alpha <dbl>, ci\_lo <dbl>,  
# dte <date>

```
with(sra_cum_prr_mc_adj, table(dte, mnth, useNA = "ifany")) %>%  
  as.data.frame() %>%  
  dplyr::filter(Freq > 0) %>%  
  arrange(mnth, dte)
```

	dte	mnth	Freq
1	2012-04-01	2012-Q2	16
2	2012-07-01	2012-Q3	16
3	2012-10-01	2012-Q4	37
4	2013-01-01	2013-Q1	41
5	2013-04-01	2013-Q2	51
6	2013-07-01	2013-Q3	71
7	2013-10-01	2013-Q4	71
8	2014-01-01	2014-Q1	71
9	2014-04-01	2014-Q2	71
10	2014-07-01	2014-Q3	81
11	2014-10-01	2014-Q4	89
12	2015-01-01	2015-Q1	89
13	2015-04-01	2015-Q2	89

14	2015-07-01	2015-Q3	89
15	2015-10-01	2015-Q4	89
16	2016-01-01	2016-Q1	89
17	2016-04-01	2016-Q2	89
18	2016-07-01	2016-Q3	89
19	2016-10-01	2016-Q4	89
20	2017-01-01	2017-Q1	95
21	2017-04-01	2017-Q2	95
22	2017-07-01	2017-Q3	95
23	2017-10-01	2017-Q4	95

```
# first signif
pr_rmc_adj_signif <-
  sra_cum_prr_mc_adj %>%
  group_by(grps, dat_type, thresh) %>%
  dplyr::filter(ci_lo > 1) %>% # 1 is the critical value on ratio scale
  arrange(dte) %>%
  dplyr::filter(row_number() == 1) %>%
  ungroup() %>%
  rename(dte_reach_sig = dte)

nrow(sra_cum_prr_mc_adj)
```

[1] 1707

```
sra_cum_prr_mc_adj <-
  left_join(
    sra_cum_prr_mc_adj,
    pr_rmc_adj_signif %>% select(grps, dat_type, thresh, dte_reach_sig),
    c("grps", "dat_type", "thresh")
  )
nrow(sra_cum_prr_mc_adj)
```

[1] 1707

```
sra_cum_prr_mc_adj
```

```
# A tibble: 1,707 x 16
```

```
  grps      dat_type thresh mnth    nA    nB    nC    nD adj_alpha est_name
  <chr>      <chr>    <chr> <chr> <dbl> <dbl> <dbl> <dbl>    <dbl> <chr>
1 (a) pelvic_~ cumulat~ 0.010 2013~    3     7     1     4 0.0000337 prr
2 (a) pelvic_~ cumulat~ 0.010 2013~    4    10     1     5 0.000688 prr
3 (a) pelvic_~ cumulat~ 0.010 2013~    5    11     2     9 0.00262 prr
4 (a) pelvic_~ cumulat~ 0.010 2013~    9    11     2     9 0.00581 prr
5 (a) pelvic_~ cumulat~ 0.010 2014~    9    11     2    10 0.01 prr
6 (a) pelvic_~ cumulat~ 0.010 2014~   10    12     3    12 0.0149 prr
7 (a) pelvic_~ cumulat~ 0.010 2014~   12    14     5    19 0.0204 prr
8 (a) pelvic_~ cumulat~ 0.010 2014~   30    15     7    24 0.0262 prr
9 (a) pelvic_~ cumulat~ 0.010 2015~   31    15     7    25 0.0323 prr
10 (a) pelvic_~ cumulat~ 0.010 2015~   31    15     7    25 0.0385 prr
# i 1,697 more rows
# i 6 more variables: est_scale <chr>, est <dbl>, alpha <dbl>, ci_lo <dbl>,
#   dte <date>, dte_reach_sig <date>
```

```
sra_cum_prr_mc_adj %>%
  arrange(grps, thresh, dte, mnth) %>%
  group_by(grps, thresh, dte, mnth) %>%
  summarise(n = n()) %>%
  ungroup() %>%
  dplyr::filter(n > 1)
```

`summarise()` has grouped output by 'grps', 'thresh', 'dte'. You can override using the `.groups` argument.

```
# A tibble: 0 x 5
```

```
# i 5 variables: grps <chr>, thresh <chr>, dte <date>, mnth <chr>, n <int>
```

```
sra_cum_prr_mc_adj %>%
  dplyr::filter(thresh == "0.070", grepl("(a)", grps, fixed = TRUE))
```

```
# A tibble: 14 x 16
```

```
  grps      dat_type thresh mnth    nA    nB    nC    nD adj_alpha est_name
  <chr>      <chr>    <chr> <chr> <dbl> <dbl> <dbl> <dbl>    <dbl> <chr>
1 (a) pelvic_~ cumulat~ 0.070 2014~    7    19     1    23 0.000181 prr
2 (a) pelvic_~ cumulat~ 0.070 2014~   24    21     2    29 0.00226 prr
3 (a) pelvic_~ cumulat~ 0.070 2015~   25    21     2    30 0.00691 prr
```

```

4 (a) pelvic_~ cumulat~ 0.070 2015~ 25 21 2 30 0.0135 prr
5 (a) pelvic_~ cumulat~ 0.070 2015~ 25 22 2 30 0.0212 prr
6 (a) pelvic_~ cumulat~ 0.070 2015~ 25 22 2 32 0.0297 prr
7 (a) pelvic_~ cumulat~ 0.070 2016~ 28 23 2 32 0.0385 prr
8 (a) pelvic_~ cumulat~ 0.070 2016~ 32 23 2 32 0.0475 prr
9 (a) pelvic_~ cumulat~ 0.070 2016~ 32 23 2 38 0.0566 prr
10 (a) pelvic_~ cumulat~ 0.070 2016~ 34 25 2 38 0.0656 prr
11 (a) pelvic_~ cumulat~ 0.070 2017~ 42 26 3 39 0.0744 prr
12 (a) pelvic_~ cumulat~ 0.070 2017~ 54 28 3 42 0.0832 prr
13 (a) pelvic_~ cumulat~ 0.070 2017~ 64 28 3 43 0.0917 prr
14 (a) pelvic_~ cumulat~ 0.070 2017~ 73 29 3 43 0.1 prr
# i 6 more variables: est_scale <chr>, est <dbl>, alpha <dbl>, ci_lo <dbl>,
# dte <date>, dte_reach_sig <date>

```

```

sra_cum_prr_mc_adj %>%
  dplyr::filter(thresh == "0.050", grepl("(c)", grps, fixed = TRUE)) %>%
  print(., n = nrow())

```

# A tibble: 23 x 16

	grps	dat_type	thresh	mnth	nA	nB	nC	nD	adj_alpha	est_name
	<chr>	<chr>	<chr>	<chr>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<chr>
1	(c) pelvic_~	cumulat~	0.050	2012~	1	3	21	155	0.0000160	prr
2	(c) pelvic_~	cumulat~	0.050	2012~	1	4	71	551	0.000406	prr
3	(c) pelvic_~	cumulat~	0.050	2012~	3	4	121	925	0.00170	prr
4	(c) pelvic_~	cumulat~	0.050	2013~	3	7	167	1366	0.00400	prr
5	(c) pelvic_~	cumulat~	0.050	2013~	3	11	234	1955	0.00717	prr
6	(c) pelvic_~	cumulat~	0.050	2013~	4	12	293	2501	0.0110	prr
7	(c) pelvic_~	cumulat~	0.050	2013~	6	14	356	3129	0.0154	prr
8	(c) pelvic_~	cumulat~	0.050	2014~	6	14	456	3740	0.0202	prr
9	(c) pelvic_~	cumulat~	0.050	2014~	7	15	534	4318	0.0252	prr
10	(c) pelvic_~	cumulat~	0.050	2014~	9	17	649	5019	0.0304	prr
11	(c) pelvic_~	cumulat~	0.050	2014~	26	19	791	5753	0.0358	prr
12	(c) pelvic_~	cumulat~	0.050	2015~	27	19	861	6231	0.0413	prr
13	(c) pelvic_~	cumulat~	0.050	2015~	27	19	922	6815	0.0468	prr
14	(c) pelvic_~	cumulat~	0.050	2015~	27	20	967	7315	0.0523	prr
15	(c) pelvic_~	cumulat~	0.050	2015~	27	20	1041	7803	0.0578	prr
16	(c) pelvic_~	cumulat~	0.050	2016~	30	21	1099	8283	0.0632	prr
17	(c) pelvic_~	cumulat~	0.050	2016~	34	21	1160	8787	0.0687	prr
18	(c) pelvic_~	cumulat~	0.050	2016~	34	21	1219	9363	0.0741	prr
19	(c) pelvic_~	cumulat~	0.050	2016~	36	23	1281	9933	0.0794	prr
20	(c) pelvic_~	cumulat~	0.050	2017~	45	23	1338	10515	0.0846	prr

```

21 (c) pelvic_~ cumulat~ 0.050 2017~ 58 24 1417 11218 0.0898 prr
22 (c) pelvic_~ cumulat~ 0.050 2017~ 68 24 1481 11832 0.0950 prr
23 (c) pelvic_~ cumulat~ 0.050 2017~ 77 25 1584 12471 0.1 prr
# i 6 more variables: est_scale <chr>, est <dbl>, alpha <dbl>, ci_lo <dbl>,
# dte <date>, dte_reach_sig <date>

```

```

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

```

```

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

```

## 2.4 MaxSPRT

```
# sra_cum <-
#   sra_dat %>%
#   dplyr::filter(dat_type == "cumulative")
sra_cum <-
  cumul_qtrly_dat

cv_tab <-
  sra_cum %>%
  # dplyr::filter(thresh < 0.070) %>%
  group_by(grps, thresh) %>%
  summarise(
    min_dte = min(mnth),
    max_dte = max(mnth),
    rows = n(),
    sum_nA = max(nA),
    sum_nC = max(nC),
    tot_n = sum_nA + sum_nC,
    .groups = "drop"
  ) %>%
  mutate(
    # qtrs = interval(paste0(min_dte, "-01"), paste0(max_dte, "-01")) / months(1) / 4,
    qtrs = rows,
    n_per_qtr = tot_n / qtrs,
    z = sum_nC / sum_nA
  )

cv_tab %>%
  kable(., digits = 1)
```

grps	thresh	min_dte	max_dte	rows	sum_nA	sum_nC	tot_n	qtrs	n_per_qtr	z
(a) pelvic_mesh v hernia_mesh	0.010	2013- Q1	2017- Q4	20	82	12	94	20	4.7	0.1
(a) pelvic_mesh v hernia_mesh	0.015	2013- Q1	2017- Q4	20	82	12	94	20	4.7	0.1
(a) pelvic_mesh v hernia_mesh	0.020	2013- Q3	2017- Q4	18	82	10	92	18	5.1	0.1
(a) pelvic_mesh v hernia_mesh	0.025	2013- Q3	2017- Q4	18	82	10	92	18	5.1	0.1

grps	thresh	min_dt	max_dt	rows	sum	rs	Am	it	6t	nqtrs	n_per	qtz
(a) pelvic_mesh v hernia_mesh	0.030	2013- Q3	2017- Q4	18	82	10	92	18	5.1	0.1		
(a) pelvic_mesh v hernia_mesh	0.035	2013- Q3	2017- Q4	18	82	10	92	18	5.1	0.1		
(a) pelvic_mesh v hernia_mesh	0.040	2013- Q3	2017- Q4	18	81	9	90	18	5.0	0.1		
(a) pelvic_mesh v hernia_mesh	0.045	2013- Q3	2017- Q4	18	79	8	87	18	4.8	0.1		
(a) pelvic_mesh v hernia_mesh	0.050	2013- Q3	2017- Q4	18	77	8	85	18	4.7	0.1		
(a) pelvic_mesh v hernia_mesh	0.055	2013- Q3	2017- Q4	18	77	8	85	18	4.7	0.1		
(a) pelvic_mesh v hernia_mesh	0.060	2013- Q3	2017- Q4	18	76	6	82	18	4.6	0.1		
(a) pelvic_mesh v hernia_mesh	0.065	2013- Q3	2017- Q4	18	75	5	80	18	4.4	0.1		
(a) pelvic_mesh v hernia_mesh	0.070	2014- Q3	2017- Q4	14	73	3	76	14	5.4	0.0		
(a) pelvic_mesh v hernia_mesh	0.075	2014- Q3	2017- Q4	14	72	3	75	14	5.4	0.0		
(a) pelvic_mesh v hernia_mesh	0.080	2014- Q3	2017- Q4	14	72	3	75	14	5.4	0.0		
(a) pelvic_mesh v hernia_mesh	0.085	2014- Q3	2017- Q4	14	71	2	73	14	5.2	0.0		
(a) pelvic_mesh v hernia_mesh	0.090	2017- Q1	2017- Q4	4	70	1	71	4	17.8	0.0		
(a) pelvic_mesh v hernia_mesh	0.095	2017- Q1	2017- Q4	4	69	1	70	4	17.5	0.0		
(a) pelvic_mesh v hernia_mesh	0.100	2017- Q1	2017- Q4	4	69	1	70	4	17.5	0.0		
(b) pelvic_mesh v hernia_mesh/other_mesh	0.010	2012- Q4	2017- Q4	21	82	59	141	21	6.7	0.7		
(b) pelvic_mesh v hernia_mesh/other_mesh	0.015	2012- Q4	2017- Q4	21	82	59	141	21	6.7	0.7		
(b) pelvic_mesh v hernia_mesh/other_mesh	0.020	2012- Q4	2017- Q4	21	82	56	138	21	6.6	0.7		
(b) pelvic_mesh v hernia_mesh/other_mesh	0.025	2012- Q4	2017- Q4	21	82	56	138	21	6.6	0.7		
(b) pelvic_mesh v hernia_mesh/other_mesh	0.030	2012- Q4	2017- Q4	21	82	55	137	21	6.5	0.7		



grps	thresh	min_dt	max_dt	rows	sum	rs	Am	rt	6t_nqtrs	n_per	qtz
(b) pelvic_mesh v	0.035	2012-	2017-	21	82	54	136	21	6.5	0.7	
hernia_mesh/other_mesh		Q4	Q4								
(b) pelvic_mesh v	0.040	2012-	2017-	21	81	51	132	21	6.3	0.6	
hernia_mesh/other_mesh		Q4	Q4								
(b) pelvic_mesh v	0.045	2012-	2017-	21	79	47	126	21	6.0	0.6	
hernia_mesh/other_mesh		Q4	Q4								
(b) pelvic_mesh v	0.050	2012-	2017-	21	77	45	122	21	5.8	0.6	
hernia_mesh/other_mesh		Q4	Q4								
(b) pelvic_mesh v	0.055	2013-	2017-	19	77	43	120	19	6.3	0.6	
hernia_mesh/other_mesh		Q2	Q4								
(b) pelvic_mesh v	0.060	2013-	2017-	19	76	41	117	19	6.2	0.5	
hernia_mesh/other_mesh		Q2	Q4								
(b) pelvic_mesh v	0.065	2013-	2017-	19	75	38	113	19	5.9	0.5	
hernia_mesh/other_mesh		Q2	Q4								
(b) pelvic_mesh v	0.070	2013-	2017-	19	73	36	109	19	5.7	0.5	
hernia_mesh/other_mesh		Q2	Q4								
(b) pelvic_mesh v	0.075	2013-	2017-	19	72	35	107	19	5.6	0.5	
hernia_mesh/other_mesh		Q2	Q4								
(b) pelvic_mesh v	0.080	2014-	2017-	14	72	33	105	14	7.5	0.5	
hernia_mesh/other_mesh		Q3	Q4								
(b) pelvic_mesh v	0.085	2014-	2017-	14	71	31	102	14	7.3	0.4	
hernia_mesh/other_mesh		Q3	Q4								
(b) pelvic_mesh v	0.090	2014-	2017-	13	70	30	100	13	7.7	0.4	
hernia_mesh/other_mesh		Q4	Q4								
(b) pelvic_mesh v	0.095	2014-	2017-	13	69	30	99	13	7.6	0.4	
hernia_mesh/other_mesh		Q4	Q4								
(b) pelvic_mesh v	0.100	2014-	2017-	13	69	30	99	13	7.6	0.4	
hernia_mesh/other_mesh		Q4	Q4								
(c) pelvic_mesh v her-	0.010	2012-	2017-	23	82	2017	2099	23	91.3	24.6	
nia_mesh/other_mesh/other_device		Q2	Q4								
(c) pelvic_mesh v her-	0.015	2012-	2017-	23	82	1994	2076	23	90.3	24.3	
nia_mesh/other_mesh/other_device		Q2	Q4								
(c) pelvic_mesh v her-	0.020	2012-	2017-	23	82	1951	2033	23	88.4	23.8	
nia_mesh/other_mesh/other_device		Q2	Q4								
(c) pelvic_mesh v her-	0.025	2012-	2017-	23	82	1910	1992	23	86.6	23.3	
nia_mesh/other_mesh/other_device		Q2	Q4								
(c) pelvic_mesh v her-	0.030	2012-	2017-	23	82	1852	1934	23	84.1	22.6	
nia_mesh/other_mesh/other_device		Q2	Q4								
(c) pelvic_mesh v her-	0.035	2012-	2017-	23	82	1783	1865	23	81.1	21.7	
nia_mesh/other_mesh/other_device		Q2	Q4								

grps	thresh	min_dt	max_dt	rows	sum	rs	Am	16t	n_qtrs	n_per_qtr	z
(c) pelvic_mesh v her-	0.040	2012-	2017-	23	81	1715	1796	23	78.1	21.2	
nia_mesh/other_mesh/other_device		Q2	Q4								
(c) pelvic_mesh v her-	0.045	2012-	2017-	23	79	1656	1735	23	75.4	21.0	
nia_mesh/other_mesh/other_device		Q2	Q4								
(c) pelvic_mesh v her-	0.050	2012-	2017-	23	77	1584	1661	23	72.2	20.6	
nia_mesh/other_mesh/other_device		Q2	Q4								
(c) pelvic_mesh v her-	0.055	2012-	2017-	23	77	1510	1587	23	69.0	19.6	
nia_mesh/other_mesh/other_device		Q2	Q4								
(c) pelvic_mesh v her-	0.060	2012-	2017-	23	76	1406	1482	23	64.4	18.5	
nia_mesh/other_mesh/other_device		Q2	Q4								
(c) pelvic_mesh v her-	0.065	2012-	2017-	23	75	1331	1406	23	61.1	17.7	
nia_mesh/other_mesh/other_device		Q2	Q4								
(c) pelvic_mesh v her-	0.070	2012-	2017-	23	73	1258	1331	23	57.9	17.2	
nia_mesh/other_mesh/other_device		Q2	Q4								
(c) pelvic_mesh v her-	0.075	2012-	2017-	23	72	1186	1258	23	54.7	16.5	
nia_mesh/other_mesh/other_device		Q2	Q4								
(c) pelvic_mesh v her-	0.080	2012-	2017-	23	72	1104	1176	23	51.1	15.3	
nia_mesh/other_mesh/other_device		Q2	Q4								
(c) pelvic_mesh v her-	0.085	2012-	2017-	23	71	1024	1095	23	47.6	14.4	
nia_mesh/other_mesh/other_device		Q2	Q4								
(c) pelvic_mesh v her-	0.090	2012-	2017-	21	70	1013	1083	21	51.6	14.5	
nia_mesh/other_mesh/other_device		Q4	Q4								
(c) pelvic_mesh v her-	0.095	2012-	2017-	21	69	940	1009	21	48.0	13.6	
nia_mesh/other_mesh/other_device		Q4	Q4								
(c) pelvic_mesh v her-	0.100	2012-	2017-	21	69	933	1002	21	47.7	13.5	
nia_mesh/other_mesh/other_device		Q4	Q4								
(d) hernia_mesh v	0.010	2013-	2017-	20	12	47	59	20	3.0	3.9	
other_mesh		Q1	Q4								
(d) hernia_mesh v	0.015	2013-	2017-	20	12	47	59	20	3.0	3.9	
other_mesh		Q1	Q4								
(d) hernia_mesh v	0.020	2013-	2017-	18	10	46	56	18	3.1	4.6	
other_mesh		Q3	Q4								
(d) hernia_mesh v	0.025	2013-	2017-	18	10	46	56	18	3.1	4.6	
other_mesh		Q3	Q4								
(d) hernia_mesh v	0.030	2013-	2017-	18	10	45	55	18	3.1	4.5	
other_mesh		Q3	Q4								
(d) hernia_mesh v	0.035	2013-	2017-	18	10	44	54	18	3.0	4.4	
other_mesh		Q3	Q4								
(d) hernia_mesh v	0.040	2013-	2017-	18	9	42	51	18	2.8	4.7	
other_mesh		Q3	Q4								

grps	thresh	min_dt	max_dt	n_rows	sum_n	sum_m	sum_t	sum_q	n_per_qtr	tz
(d) hernia_mesh v other_mesh	0.045	2013- Q3	2017- Q4	18	8	39	47	18	2.6	4.9
(d) hernia_mesh v other_mesh	0.050	2013- Q3	2017- Q4	18	8	37	45	18	2.5	4.6
(d) hernia_mesh v other_mesh	0.055	2013- Q3	2017- Q4	18	8	35	43	18	2.4	4.4
(d) hernia_mesh v other_mesh	0.060	2013- Q3	2017- Q4	18	6	35	41	18	2.3	5.8
(d) hernia_mesh v other_mesh	0.065	2013- Q3	2017- Q4	18	5	33	38	18	2.1	6.6
(d) hernia_mesh v other_mesh	0.070	2014- Q3	2017- Q4	14	3	33	36	14	2.6	11.0
(d) hernia_mesh v other_mesh	0.075	2014- Q3	2017- Q4	14	3	32	35	14	2.5	10.7
(d) hernia_mesh v other_mesh	0.080	2014- Q4	2017- Q4	13	3	30	33	13	2.5	10.0
(d) hernia_mesh v other_mesh	0.085	2014- Q4	2017- Q4	13	2	29	31	13	2.4	14.5
(d) hernia_mesh v other_mesh	0.090	2017- Q1	2017- Q4	4	1	29	30	4	7.5	29.0
(d) hernia_mesh v other_mesh	0.095	2017- Q1	2017- Q4	4	1	29	30	4	7.5	29.0
(d) hernia_mesh v other_mesh	0.100	2017- Q1	2017- Q4	4	1	29	30	4	7.5	29.0
(e) hernia_mesh/other_mesh v other_device	0.010	2012- Q4	2017- Q4	21	59	1958	2017	21	96.0	33.2
(e) hernia_mesh/other_mesh v other_device	0.015	2012- Q4	2017- Q4	21	59	1935	1994	21	95.0	32.8
(e) hernia_mesh/other_mesh v other_device	0.020	2012- Q4	2017- Q4	21	56	1895	1951	21	92.9	33.8
(e) hernia_mesh/other_mesh v other_device	0.025	2012- Q4	2017- Q4	21	56	1854	1910	21	91.0	33.1
(e) hernia_mesh/other_mesh v other_device	0.030	2012- Q4	2017- Q4	21	55	1797	1852	21	88.2	32.7
(e) hernia_mesh/other_mesh v other_device	0.035	2012- Q4	2017- Q4	21	54	1729	1783	21	84.9	32.0
(e) hernia_mesh/other_mesh v other_device	0.040	2012- Q4	2017- Q4	21	51	1664	1715	21	81.7	32.6
(e) hernia_mesh/other_mesh v other_device	0.045	2012- Q4	2017- Q4	21	47	1609	1656	21	78.9	34.2

grps	thresh	min_dt	max_dt	rows	sum	sd	Am	16	64	n_qtrs	n_per_qtr	z
(e) hernia_mesh/other_mesh v other_device	0.050	2012- Q4	2017- Q4	21	45	1539	1584	21	75.4	34.2		
(e) hernia_mesh/other_mesh v other_device	0.055	2013- Q2	2017- Q4	19	43	1467	1510	19	79.5	34.1		
(e) hernia_mesh/other_mesh v other_device	0.060	2013- Q2	2017- Q4	19	41	1365	1406	19	74.0	33.3		
(e) hernia_mesh/other_mesh v other_device	0.065	2013- Q2	2017- Q4	19	38	1293	1331	19	70.1	34.0		
(e) hernia_mesh/other_mesh v other_device	0.070	2013- Q2	2017- Q4	19	36	1222	1258	19	66.2	33.9		
(e) hernia_mesh/other_mesh v other_device	0.075	2013- Q2	2017- Q4	19	35	1151	1186	19	62.4	32.9		
(e) hernia_mesh/other_mesh v other_device	0.080	2014- Q3	2017- Q4	14	33	1071	1104	14	78.9	32.5		
(e) hernia_mesh/other_mesh v other_device	0.085	2014- Q3	2017- Q4	14	31	993	1024	14	73.1	32.0		
(e) hernia_mesh/other_mesh v other_device	0.090	2014- Q4	2017- Q4	13	30	983	1013	13	77.9	32.8		
(e) hernia_mesh/other_mesh v other_device	0.095	2014- Q4	2017- Q4	13	30	910	940	13	72.3	30.3		
(e) hernia_mesh/other_mesh v other_device	0.100	2014- Q4	2017- Q4	13	30	903	933	13	71.8	30.1		

```

# maxsprt: create alternative CV tab

### create CV tab for alternative n_per_qtr and z ratios
cv_tab_alts <-
  tribble(
    ~alt_str, ~modifier, ~mult,
    "half_n", "n_per_qtr", 0.5,
    "doub_n", "n_per_qtr", 2,
    "half_z", "z", 0.5,
    "doub_z", "z", 2
  )

cv_tab_alts <-

```

```

cross_join(
  cv_tab_alts,
  cv_tab
) %>%
arrange(
  grps, thresh, modifier, alt_str
)

if (4 * nrow(cv_tab) != nrow(cv_tab_alts)) {
  stop("cross_join() has gone wrong")
}

cv_tab_alts <-
  cv_tab_alts %>%
  mutate(
    n_per_qtr = if_else(modifier == "n_per_qtr", mult * n_per_qtr, n_per_qtr),
    z         = if_else(modifier == "z", mult * z, z),
  )

cv_tab_alts

```

# A tibble: 380 x 14

	alt_str	modifier	mult	grps	thresh	min_dte	max_dte	rows	sum_nA	sum_nC	tot_n
	<chr>	<chr>	<dbl>	<chr>	<chr>	<chr>	<chr>	<int>	<dbl>	<dbl>	<dbl>
1	doub_n	n_per_q~	2	(a)	~ 0.010	2013-Q1	2017-Q4	20	82	12	94
2	half_n	n_per_q~	0.5	(a)	~ 0.010	2013-Q1	2017-Q4	20	82	12	94
3	doub_z	z	2	(a)	~ 0.010	2013-Q1	2017-Q4	20	82	12	94
4	half_z	z	0.5	(a)	~ 0.010	2013-Q1	2017-Q4	20	82	12	94
5	doub_n	n_per_q~	2	(a)	~ 0.015	2013-Q1	2017-Q4	20	82	12	94
6	half_n	n_per_q~	0.5	(a)	~ 0.015	2013-Q1	2017-Q4	20	82	12	94
7	doub_z	z	2	(a)	~ 0.015	2013-Q1	2017-Q4	20	82	12	94
8	half_z	z	0.5	(a)	~ 0.015	2013-Q1	2017-Q4	20	82	12	94
9	doub_n	n_per_q~	2	(a)	~ 0.020	2013-Q3	2017-Q4	18	82	10	92
10	half_n	n_per_q~	0.5	(a)	~ 0.020	2013-Q3	2017-Q4	18	82	10	92

# i 370 more rows

# i 3 more variables: qtrs <int>, n\_per\_qtr <dbl>, z <dbl>

```
# maxsprt: create CVs
```

```
# testing/example
```

```
row_i <- 1
```

```
cv_tab[row_i, ]
```

```
# A tibble: 1 x 11
```

	grps	thresh	min_dte	max_dte	rows	sum_nA	sum_nC	tot_n	qtrs	n_per_qtr	z
	<chr>	<chr>	<chr>	<chr>	<int>	<dbl>	<dbl>	<dbl>	<int>	<dbl>	<dbl>
1	(a) pe~	0.010	2013-Q1	2017-Q4	20	82	12	94	20	4.7	0.146

```
get_maxsprt_cv(cv_tab$tot_n[row_i], floor(cv_tab$n_per_qtr[row_i]), cv_tab$z[row_i])
```

```
[1] 3.27782
```

```
row_i <- 50  
cv_tab[row_i, ]
```

```
# A tibble: 1 x 11
```

	grps	thresh	min_dte	max_dte	rows	sum_nA	sum_nC	tot_n	qtrs	n_per_qtr	z
	<chr>	<chr>	<chr>	<chr>	<int>	<dbl>	<dbl>	<dbl>	<int>	<dbl>	<dbl>
1	(c) pe~	0.065	2012-Q2	2017-Q4	23	75	1331	1406	23	61.1	17.7

```
get_maxsprt_cv(cv_tab$tot_n[row_i], floor(cv_tab$n_per_qtr[row_i]), cv_tab$z[row_i])
```

Selected alpha: 0.048 (least conservative value below 0.05)

```
[1] 2.740269  
attr(,"alpha")  
[1] 0.048193
```

```
### takes ~ 70 sec (i5-8400)  
# note purrr::possibly() will just catch when model fails and return as.numeric(NA)  
get_maxsprt_cv_poss <-  
  possibly(get_maxsprt_cv, otherwise = NA_real_, quiet = FALSE)  
  
tic()  
cv_tab <-  
  cv_tab %>%  
    # dplyr::filter(row_number() < 7) %>% ### testing
```

```

mutate(
  cv =
    future_pmap_dbl(
      .l = list(tot_n, floor(n_per_qtr), z),
      .f = ~get_maxsprt_cv_poss(..1, ..2, ..3),
      .options = furrr_seed3
    )
)

```

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```
toc()
```

197.59 sec elapsed

```
cv_tab
```

```
# A tibble: 95 x 12
```

	grps	thresh	min_dte	max_dte	rows	sum_nA	sum_nC	tot_n	qtrs	n_per_qtr	z
	<chr>	<chr>	<chr>	<chr>	<int>	<dbl>	<dbl>	<dbl>	<int>	<dbl>	<dbl>
1	(a)	p~	0.010	2013-Q1	2017-Q4	20	82	12	94	20	4.7 0.146
2	(a)	p~	0.015	2013-Q1	2017-Q4	20	82	12	94	20	4.7 0.146
3	(a)	p~	0.020	2013-Q3	2017-Q4	18	82	10	92	18	5.11 0.122
4	(a)	p~	0.025	2013-Q3	2017-Q4	18	82	10	92	18	5.11 0.122
5	(a)	p~	0.030	2013-Q3	2017-Q4	18	82	10	92	18	5.11 0.122
6	(a)	p~	0.035	2013-Q3	2017-Q4	18	82	10	92	18	5.11 0.122
7	(a)	p~	0.040	2013-Q3	2017-Q4	18	81	9	90	18	5 0.111
8	(a)	p~	0.045	2013-Q3	2017-Q4	18	79	8	87	18	4.83 0.101
9	(a)	p~	0.050	2013-Q3	2017-Q4	18	77	8	85	18	4.72 0.104
10	(a)	p~	0.055	2013-Q3	2017-Q4	18	77	8	85	18	4.72 0.104

```
# i 85 more rows
```

```
# i 1 more variable: cv <dbl>
```

```
cv_tab %>% dplyr::filter(is.na(cv))
```

```
# A tibble: 4 x 12
```

	grps	thresh	min_dte	max_dte	rows	sum_nA	sum_nC	tot_n	qtrs	n_per_qtr	z
	<chr>	<chr>	<chr>	<chr>	<int>	<dbl>	<dbl>	<dbl>	<int>	<dbl>	<dbl>
1	(a)	p~	0.085	2014-Q3	2017-Q4	14	71	2	73	14	5.21 0.0282
2	(a)	p~	0.090	2017-Q1	2017-Q4	4	70	1	71	4	17.8 0.0143
3	(a)	p~	0.095	2017-Q1	2017-Q4	4	69	1	70	4	17.5 0.0145
4	(a)	p~	0.100	2017-Q1	2017-Q4	4	69	1	70	4	17.5 0.0145

```
# i 1 more variable: cv <dbl>
```

```
# remove analyses where thresholds don't allow enough events (extreme threshold values)
# cv_tab <- cv_tab %>% dplyr::filter(!is.na(cv))
```

```

# maxsprt: create alt CVs

### takes ~ 120 sec (R9-5900X)
tic()
cv_tab_alts <-
  cv_tab_alts %>%
    # dplyr::filter(row_number() < 7) %>% ### testing
    mutate(
      cv =
        future_pmap_dbl(
          .l = list(tot_n, floor(n_per_qtr), z),
          .f = ~get_maxsprt_cv_poss(..1, ..2, ..3),
          .options = furrr_seed4
        )
    )

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Selected alpha: 0.048 (least conservative value below 0.05)

Selected alpha: 0.050 (least conservative value below 0.05)

`toc()`

861.36 sec elapsed

`cv_tab_alts`

```
# A tibble: 380 x 15
```

	alt_str	modifier	mult	grps	thresh	min_dte	max_dte	rows	sum_nA	sum_nC	tot_n
	<chr>	<chr>	<dbl>	<chr>	<chr>	<chr>	<chr>	<int>	<dbl>	<dbl>	<dbl>
1	doub_n	n_per_q~	2	(a)	~ 0.010	2013-Q1	2017-Q4	20	82	12	94
2	half_n	n_per_q~	0.5	(a)	~ 0.010	2013-Q1	2017-Q4	20	82	12	94
3	doub_z	z	2	(a)	~ 0.010	2013-Q1	2017-Q4	20	82	12	94
4	half_z	z	0.5	(a)	~ 0.010	2013-Q1	2017-Q4	20	82	12	94
5	doub_n	n_per_q~	2	(a)	~ 0.015	2013-Q1	2017-Q4	20	82	12	94
6	half_n	n_per_q~	0.5	(a)	~ 0.015	2013-Q1	2017-Q4	20	82	12	94
7	doub_z	z	2	(a)	~ 0.015	2013-Q1	2017-Q4	20	82	12	94
8	half_z	z	0.5	(a)	~ 0.015	2013-Q1	2017-Q4	20	82	12	94
9	doub_n	n_per_q~	2	(a)	~ 0.020	2013-Q3	2017-Q4	18	82	10	92
10	half_n	n_per_q~	0.5	(a)	~ 0.020	2013-Q3	2017-Q4	18	82	10	92

```
# i 370 more rows
```

```
# i 4 more variables: qtrs <int>, n_per_qtr <dbl>, z <dbl>, cv <dbl>
```

```
cv_tab_alts %>% dplyr::filter(is.na(cv))
```

```
# A tibble: 19 x 15
```

	alt_str	modifier	mult	grps	thresh	min_dte	max_dte	rows	sum_nA	sum_nC	tot_n
	<chr>	<chr>	<dbl>	<chr>	<chr>	<chr>	<chr>	<int>	<dbl>	<dbl>	<dbl>
1	half_z	z	0.5	(a)	~ 0.065	2013-Q3	2017-Q4	18	75	5	80
2	half_z	z	0.5	(a)	~ 0.070	2014-Q3	2017-Q4	14	73	3	76
3	half_z	z	0.5	(a)	~ 0.075	2014-Q3	2017-Q4	14	72	3	75
4	half_z	z	0.5	(a)	~ 0.080	2014-Q3	2017-Q4	14	72	3	75
5	doub_n	n_per_q~	2	(a)	~ 0.085	2014-Q3	2017-Q4	14	71	2	73
6	half_n	n_per_q~	0.5	(a)	~ 0.085	2014-Q3	2017-Q4	14	71	2	73
7	half_z	z	0.5	(a)	~ 0.085	2014-Q3	2017-Q4	14	71	2	73
8	doub_n	n_per_q~	2	(a)	~ 0.090	2017-Q1	2017-Q4	4	70	1	71
9	half_n	n_per_q~	0.5	(a)	~ 0.090	2017-Q1	2017-Q4	4	70	1	71
10	doub_z	z	2	(a)	~ 0.090	2017-Q1	2017-Q4	4	70	1	71
11	half_z	z	0.5	(a)	~ 0.090	2017-Q1	2017-Q4	4	70	1	71
12	doub_n	n_per_q~	2	(a)	~ 0.095	2017-Q1	2017-Q4	4	69	1	70
13	half_n	n_per_q~	0.5	(a)	~ 0.095	2017-Q1	2017-Q4	4	69	1	70
14	doub_z	z	2	(a)	~ 0.095	2017-Q1	2017-Q4	4	69	1	70
15	half_z	z	0.5	(a)	~ 0.095	2017-Q1	2017-Q4	4	69	1	70
16	doub_n	n_per_q~	2	(a)	~ 0.100	2017-Q1	2017-Q4	4	69	1	70
17	half_n	n_per_q~	0.5	(a)	~ 0.100	2017-Q1	2017-Q4	4	69	1	70
18	doub_z	z	2	(a)	~ 0.100	2017-Q1	2017-Q4	4	69	1	70
19	half_z	z	0.5	(a)	~ 0.100	2017-Q1	2017-Q4	4	69	1	70

```
# i 4 more variables: qtrs <int>, n_per_qtr <dbl>, z <dbl>, cv <dbl>
```

```
# maxsprt: create llr test stats
```

```
maxsprt_dat <-
  sra_cum %>%
  mutate(
    maxllr = max_sprt_stat_(c_n = nA, n = nA + nC, z = (nC + nD) / (nA + nB)),
    rre = rr_est_(c_n = nA, n = nA + nC, z = (nC + nD) / (nA + nB))
  )
```

```
# maxsprt_dat
# maxsprt_dat %>% dplyr::filter(thresh == "0.100", substr(grps, 1, 3) == "(a)")
```

```
maxsprt_dat <-
  maxsprt_dat %>%
  left_join(
    ``,
    cv_tab %>% select(grps, thresh, cv),
    c("grps", "thresh")
  )
```

```
maxsprt_dat <-
  maxsprt_dat %>%
  mutate(
    # some cvs don't exist so those llr never reach cv
    reached_cv = if_else(is.na(cv), 0L, as.integer(maxllr > cv)),
    # create date for start of each quarter
    dte =
      as_date(paste0(
        substr(mnth, 1, 5),
        sprintf("%02.0f", (as.integer(substr(mnth, 7, 7)) - 1) * 3 + 1),
        "-01"
      ))
  )
```

```
maxsprt_dat %>% dplyr::filter(is.na(cv))
```

```
# A tibble: 26 x 13
```

	grps	dat_type	thresh	mnth	nA	nB	nC	nD	maxllr	rre	cv
	<chr>	<chr>	<chr>	<chr>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>
1	(a) pelvic_~	cumulat~	0.085	2014~	6	20	1	23	1.79	5.54	NA
2	(a) pelvic_~	cumulat~	0.085	2014~	23	22	1	30	8.79	15.8	NA

```

3 (a) pelvic_~ cumulat~ 0.085 2015~ 24 22 1 31 9.37 16.7 NA
4 (a) pelvic_~ cumulat~ 0.085 2015~ 24 22 1 31 9.37 16.7 NA
5 (a) pelvic_~ cumulat~ 0.085 2015~ 24 23 1 31 9.17 16.3 NA
6 (a) pelvic_~ cumulat~ 0.085 2015~ 24 23 1 33 9.73 17.4 NA
7 (a) pelvic_~ cumulat~ 0.085 2016~ 27 24 1 33 10.4 18 NA
8 (a) pelvic_~ cumulat~ 0.085 2016~ 31 24 1 33 11.4 19.2 NA
9 (a) pelvic_~ cumulat~ 0.085 2016~ 31 24 1 39 13.4 22.5 NA
10 (a) pelvic_~ cumulat~ 0.085 2016~ 33 26 1 39 13.5 22.4 NA
# i 16 more rows
# i 2 more variables: reached_cv <int>, dte <date>

```

```

# have a peak
maxsprt_dat %>%
  select(-dat_type) %>%
  print(., n = 25)

```

```
# A tibble: 1,707 x 12
```

	grps	thresh	mnth	nA	nB	nC	nD	maxllr	rre	cv	reached_cv
	<chr>	<chr>	<chr>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<int>
1	(a) pelvi~	0.010	2013~	3	7	1	4	0.0657	1.5	3.28	0
2	(a) pelvi~	0.010	2013~	4	10	1	5	0.129	1.71	3.28	0
3	(a) pelvi~	0.010	2013~	5	11	2	9	0.224	1.72	3.28	0
4	(a) pelvi~	0.010	2013~	9	11	2	9	0.801	2.48	3.28	0
5	(a) pelvi~	0.010	2014~	9	11	2	10	0.976	2.7	3.28	0
6	(a) pelvi~	0.010	2014~	10	12	3	12	0.885	2.27	3.28	0
7	(a) pelvi~	0.010	2014~	12	14	5	19	1.22	2.22	3.28	0
8	(a) pelvi~	0.010	2014~	30	15	7	24	4.05	2.95	3.28	1
9	(a) pelvi~	0.010	2015~	31	15	7	25	4.45	3.08	3.28	1
10	(a) pelvi~	0.010	2015~	31	15	7	25	4.45	3.08	3.28	1
11	(a) pelvi~	0.010	2015~	31	16	7	25	4.27	3.02	3.28	1
12	(a) pelvi~	0.010	2015~	31	16	9	25	3.36	2.49	3.28	1
13	(a) pelvi~	0.010	2016~	35	16	9	25	3.83	2.59	3.28	1
14	(a) pelvi~	0.010	2016~	39	16	9	25	4.27	2.68	3.28	1
15	(a) pelvi~	0.010	2016~	39	16	10	30	5.17	2.84	3.28	1
16	(a) pelvi~	0.010	2016~	41	18	10	30	5.04	2.78	3.28	1
17	(a) pelvi~	0.010	2017~	50	18	11	31	5.85	2.81	3.28	1
18	(a) pelvi~	0.010	2017~	63	19	12	33	7.04	2.88	3.28	1
19	(a) pelvi~	0.010	2017~	73	19	12	34	8.18	3.04	3.28	1
20	(a) pelvi~	0.010	2017~	82	20	12	34	8.65	3.08	3.28	1
21	(a) pelvi~	0.015	2013~	3	7	1	4	0.0657	1.5	3.28	0
22	(a) pelvi~	0.015	2013~	4	10	1	5	0.129	1.71	3.28	0

```

23 (a) pelvi~ 0.015 2013~ 5 11 2 9 0.224 1.72 3.28 0
24 (a) pelvi~ 0.015 2013~ 9 11 2 9 0.801 2.48 3.28 0
25 (a) pelvi~ 0.015 2014~ 9 11 2 10 0.976 2.7 3.28 0
# i 1,682 more rows
# i 1 more variable: dte <date>

```

```

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

nrow(maxsprt_dat)

```

[1] 1707

```

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

```

[1] 1707

```
maxsprt_dat
```

```

# A tibble: 1,707 x 14
  grps      dat_type thresh mnth    nA    nB    nC    nD maxllr  rre  cv
  <chr>    <chr>    <chr> <chr> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>
1 (a) pelvic~ cumulat~ 0.010 2013~ 3     7     1     4 0.0657 1.5  3.28
2 (a) pelvic~ cumulat~ 0.010 2013~ 4    10     1     5 0.129 1.71 3.28
3 (a) pelvic~ cumulat~ 0.010 2013~ 5    11     2     9 0.224 1.72 3.28

```

```

4 (a) pelvic_~ cumulat~ 0.010 2013~ 9 11 2 9 0.801 2.48 3.28
5 (a) pelvic_~ cumulat~ 0.010 2014~ 9 11 2 10 0.976 2.7 3.28
6 (a) pelvic_~ cumulat~ 0.010 2014~ 10 12 3 12 0.885 2.27 3.28
7 (a) pelvic_~ cumulat~ 0.010 2014~ 12 14 5 19 1.22 2.22 3.28
8 (a) pelvic_~ cumulat~ 0.010 2014~ 30 15 7 24 4.05 2.95 3.28
9 (a) pelvic_~ cumulat~ 0.010 2015~ 31 15 7 25 4.45 3.08 3.28
10 (a) pelvic_~ cumulat~ 0.010 2015~ 31 15 7 25 4.45 3.08 3.28
# i 1,697 more rows
# i 3 more variables: reached_cv <int>, dte <date>, dte_reach_sig <date>

```

```

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

# these are where the maxllr has dropped under the CV after exceeding it previously
maxsprt_dat %>%
  dplyr::filter(
    is.na(reach_sig) |
    is.na(reached_cv) |
    (as.logical(reached_cv) != reach_sig)
  )

```

```

# A tibble: 9 x 15
  grps      dat_type thresh mnth    nA    nB    nC    nD maxllr  rre  cv
<chr>    <chr>    <chr> <chr> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>
1 (b) pelvic_m~ cumulat~ 0.020 2016~ 41 18 38 58 3.10 1.76 3.18
2 (b) pelvic_m~ cumulat~ 0.025 2016~ 41 18 38 58 3.10 1.76 3.18
3 (b) pelvic_m~ cumulat~ 0.035 2015~ 31 16 28 51 2.82 1.86 3.04
4 (c) pelvic_m~ cumulat~ 0.055 2014~ 6 14 417 3779 2.60 3.02 2.73
5 (c) pelvic_m~ cumulat~ 0.085 2013~ 3 7 84 1449 2.61 5.48 2.88
6 (c) pelvic_m~ cumulat~ 0.085 2013~ 3 11 118 2071 1.87 3.98 2.88
7 (c) pelvic_m~ cumulat~ 0.085 2013~ 4 16 169 3316 2.61 4.12 2.88
8 (c) pelvic_m~ cumulat~ 0.085 2014~ 4 16 220 3976 2.38 3.81 2.88
9 (e) hernia_m~ cumulat~ 0.065 2015~ 14 60 772 7436 2.71 2.01 2.71
# i 4 more variables: reached_cv <int>, dte <date>, dte_reach_sig <date>,
# reach_sig <lgl>

```

```
maxsprt_dat <-  
  maxsprt_dat %>%  
  select(-reached_cv)  
  
maxsprt_dat %>%  
  write_parquet(., sink = "out/sra_cum_maxsprt.parquet")
```

### 3 Ready plot data

```
maxsprt_dat <- read_parquet("out/sra_cum_maxsprt.parquet")
bcpnn_dat <- read_parquet("out/sra_cum_bcpnn_mc_adj.parquet")
prrr_dat <- read_parquet("out/sra_cum_prr_mc_adj.parquet")

plt_dat <-
  bind_rows(
    bcpnn_dat %>%
      mutate(
        cv = 0,
        stat = "IC (BCPNN, Lower 95% CI)"
      ) %>%
      select(stat, grps, thresh, dte, cv, val = ci_lo, reach_sig, dte_reach_sig),
    prrr_dat %>%
      mutate(
        cv = 1,
        stat = "RR (PRR, Lower 95% CI)"
      ) %>%
      select(stat, grps, thresh, dte, cv, val = ci_lo, reach_sig, dte_reach_sig),
    maxsprt_dat %>%
      mutate(stat = "MaxSPRT (max LLR)") %>%
      select(stat, grps, thresh, dte, cv, val = maxllr, reach_sig, dte_reach_sig)
  )

sig_reach_dat <-
  plt_dat %>%
  arrange(stat, grps, thresh, dte) %>%
  group_by(stat, grps, thresh) %>%
  dplyr::filter(reach_sig == 1) %>%
  dplyr::filter(row_number() == 1) %>%
  select(stat, grps, thresh, dte_reached = dte) %>%
  # now create separation between reached CV values when it occurs
  group_by(stat, thresh, dte_reached) %>%
  mutate(rep_dte = 1:n()) %>%
  ungroup() %>%
  mutate(dte_reached = dte_reached + days(10 * (rep_dte - 1))) %>%
  select(-rep_dte)
```



```

plt_dat <-
  left_join(
    plt_dat,
    sig_reach_dat,
    c("stat", "grps", "thresh")
  ) # %>%
  # dplyr::filter(dte_reach_sig != dte_reached)

plt_dat <-
  plt_dat %>%
  mutate(
    stat = fct_inorder(stat)
  )

plt_dat %>%
  dplyr::filter(thresh == "0.050", grepl("(c)", grps, fixed = TRUE))

# A tibble: 69 x 9
  stat      grps  thresh dte      cv    val reach_sig dte_reach_sig
  <fct>    <chr> <chr> <date>   <dbl>  <dbl> <lgl>    <date>
1 IC (BCPNN, Low~ (c) ~ 0.050 2012-04-01    0 -9.83 FALSE    2014-04-01
2 IC (BCPNN, Low~ (c) ~ 0.050 2012-07-01    0 -8.12 FALSE    2014-04-01
3 IC (BCPNN, Low~ (c) ~ 0.050 2012-10-01    0 -1.92 FALSE    2014-04-01
4 IC (BCPNN, Low~ (c) ~ 0.050 2013-01-01    0 -1.83 FALSE    2014-04-01
5 IC (BCPNN, Low~ (c) ~ 0.050 2013-04-01    0 -1.97 FALSE    2014-04-01
6 IC (BCPNN, Low~ (c) ~ 0.050 2013-07-01    0 -1.11 FALSE    2014-04-01
7 IC (BCPNN, Low~ (c) ~ 0.050 2013-10-01    0 -0.208 FALSE    2014-04-01
8 IC (BCPNN, Low~ (c) ~ 0.050 2014-01-01    0 -0.199 FALSE    2014-04-01
9 IC (BCPNN, Low~ (c) ~ 0.050 2014-04-01    0  0.0658 TRUE     2014-04-01
10 IC (BCPNN, Low~ (c) ~ 0.050 2014-07-01    0  0.403 TRUE     2014-04-01
# i 59 more rows
# i 1 more variable: dte_reached <date>

plt_dat %>%
  dplyr::filter(thresh == "0.030", !grepl("(e)", grps, fixed = TRUE)) %>%
  ggplot(., aes(x = dte, y = val, col = grps )) +
  geom_hline(aes(yintercept = cv), alpha = 0.5) +
  geom_vline(aes(xintercept = dte_reached, col = grps), alpha = 0.5) +
  geom_line(alpha = 0.5) +

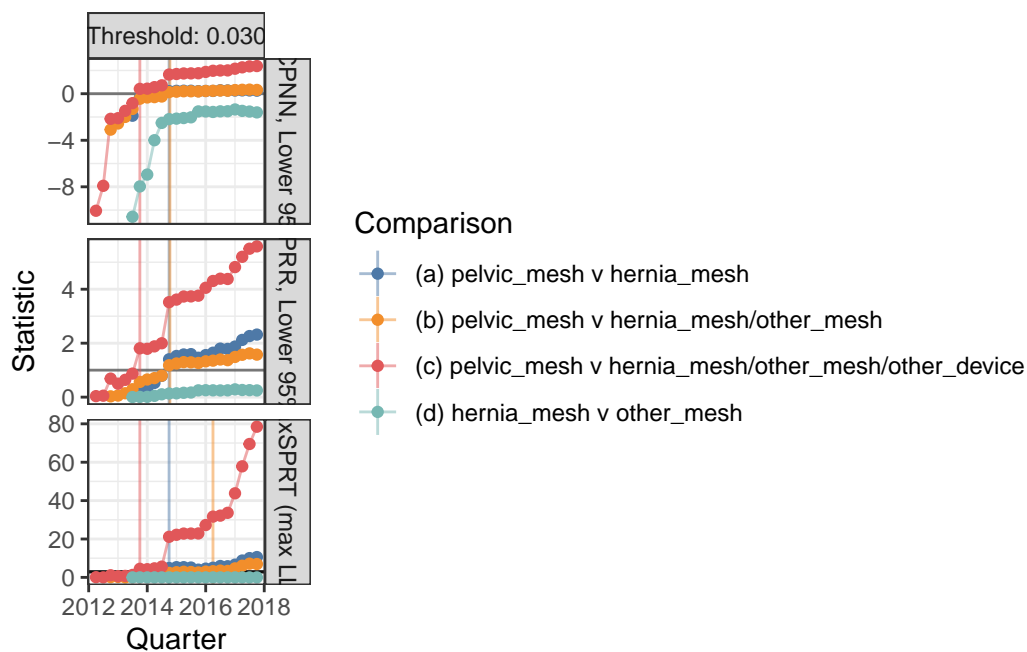
```

```

geom_point() +
# facet_wrap(thresh ~ stat, ncol = 1, scales = "free_y") +
facet_grid(
  stat ~ thresh,
  scales = "free_y",
  labeller = labeller(thresh = function(x) paste0("Threshold: ", x))
) +
scale_colour_tableau() +
theme_bw() +
labs(
  x = "Quarter",
  y = "Statistic",
  col = "Comparison"
)

```

Warning: Removed 54 rows containing missing values (`geom\_vline()`).



```

ggsave(
  filename = "fig/multi-grps_multi-test_signal-over-time_thresh-0.03.png",
  dpi = 900, width = 12, height = 8
)

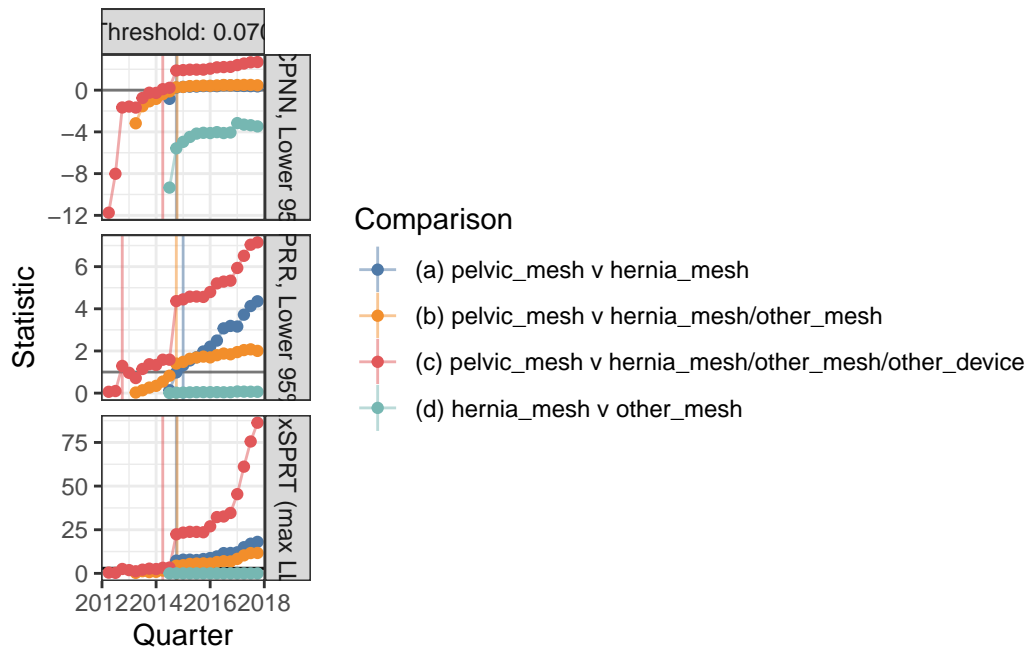
```

)

Warning: Removed 54 rows containing missing values (`geom\_vline()`).

```
plt_dat %>%
  dplyr::filter(thresh == "0.070", !grepl("(e)", grps, fixed = TRUE)) %>%
  ggplot(., aes(x = dte, y = val, col = grps )) +
  geom_hline(aes(yintercept = cv), alpha = 0.5) +
  geom_vline(aes(xintercept = dte_reached, col = grps), alpha = 0.5) +
  geom_line(alpha = 0.5) +
  geom_point() +
  # facet_wrap(thresh ~ stat, ncol = 1, scales = "free_y") +
  facet_grid(
    stat ~ thresh,
    scales = "free_y",
    labeller = labeller(thresh = function(x) paste0("Threshold: ", x))
  ) +
  scale_colour_tableau() +
  theme_bw() +
  labs(
    x = "Quarter",
    y = "Statistic",
    col = "Comparison"
  )
)
```

Warning: Removed 42 rows containing missing values (`geom\_vline()`).



```
ggsave(
  filename = "fig/multi-grps_multi-test_signal-over-time_thresh-0.07.png",
  dpi = 900, width = 12, height = 8
)
```

Warning: Removed 42 rows containing missing values (`geom\_vline()`).

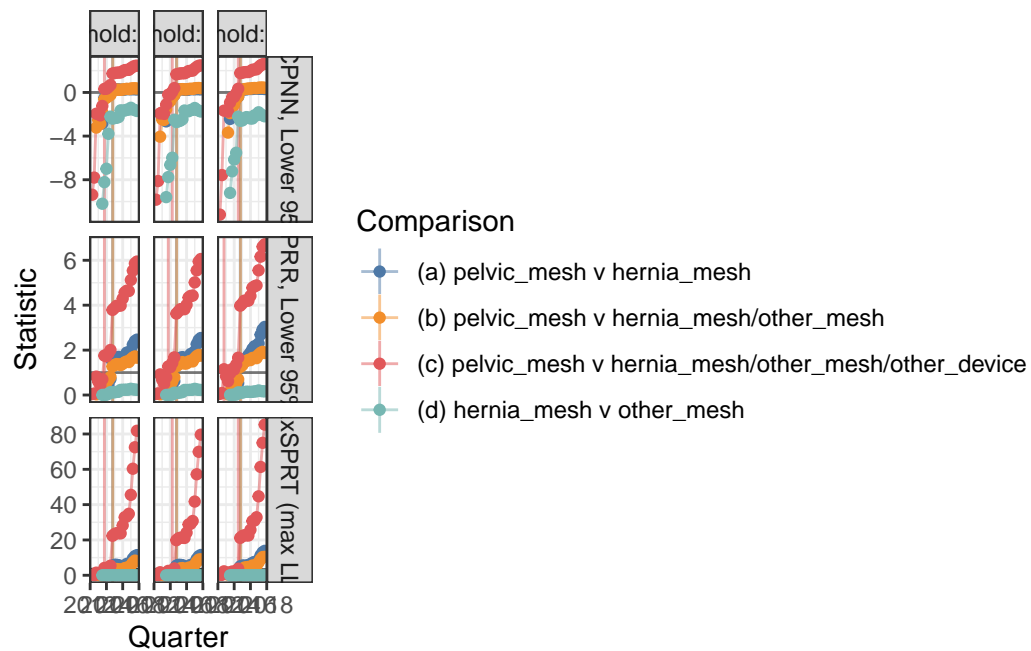
```
plt_dat %>%
  dplyr::filter(thresh %in% c("0.040", "0.050", "0.060"), !grepl("(e)", grps, fixed = TRUE)) +
  ggplot(., aes(x = dte, y = val, col = grps )) +
  geom_hline(aes(yintercept = cv), alpha = 0.5) +
  geom_vline(aes(xintercept = dte_reached, col = grps), alpha = 0.5) +
  geom_line(alpha = 0.5) +
  geom_point() +
  # facet_wrap(thresh ~ stat, ncol = 1, scales = "free_y") +
  facet_grid(
    stat ~ thresh,
    scales = "free_y",
    labeller = labeller(thresh = function(x) paste0("Threshold: ", x))
  ) +
  scale_colour_tableau() +
```

```

theme_bw() +
labs(
  x = "Quarter",
  y = "Statistic",
  col = "Comparison"
)

```

Warning: Removed 162 rows containing missing values (`geom\_vline()`).



```

ggsave(
  filename = "fig/multi-grps_multi-test_signal-over-time_thresh-range-0.04-0.06.png",
  dpi = 900, width = 12, height = 8
)

```

Warning: Removed 162 rows containing missing values (`geom\_vline()`).

## 4 Session information

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

```
# Sys.info()[!(names(Sys.info()) %in% c("login", "nodename"))] %>%
#   as.data.frame(.)
format(Sys.time(), '%d %b %Y')
```

```
[1] "04 Dec 2023"
```

```
sessionInfo()
```

```
R version 4.3.1 (2023-06-16 ucrt)
Platform: x86_64-w64-mingw32/x64 (64-bit)
Running under: Windows 10 x64 (build 19045)
```

```
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
```

```
time zone: Australia/Adelaide
```

```
tzcode source: internal
```

```
attached base packages:
```

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

```
other attached packages:
```

```
[1] pharmsignal_0.1.0 arrow_12.0.1.1   foreach_1.5.2    gsDesign_3.5.0
[5] knitr_1.43         ggthemes_5.0.0   ggrepel_0.9.3    ggplot2_3.4.2
[9] tictoc_1.2         lubridate_1.9.2  furr_0.3.1       future_1.33.0
[13] purrr_1.0.1        forcats_1.0.0    tidyr_1.3.0      dplyr_1.1.2
[17] readr_2.1.4
```

```
loaded via a namespace (and not attached):
```

```
[1] Sequential_4.3      gtable_0.3.3
```

[3] xfun_0.39	lattice_0.21-8
[5] tzdb_0.4.0	vctr_0.6.3
[7] tools_4.3.1	generics_0.1.3
[9] parallel_4.3.1	tibble_3.2.1
[11] fansi_1.0.4	pkgconfig_2.0.3
[13] Matrix_1.6-0	assertthat_0.2.1
[15] gt_0.9.0	lifecycle_1.0.3
[17] farver_2.1.1	EmpiricalCalibration_3.1.1
[19] compiler_4.3.1	stringr_1.5.0
[21] textshaping_0.3.6	MatrixModels_0.5-2
[23] mcmc_0.9-7	munSELL_0.5.0
[25] codetools_0.2-19	SparseM_1.81
[27] quantreg_5.96	htmltools_0.5.5
[29] yaml_2.3.7	pillar_1.9.0
[31] MASS_7.3-60	iterators_1.0.14
[33] boot_1.3-28.1	parallelly_1.36.0
[35] tidyselect_1.2.0	digest_0.6.33
[37] stringi_1.7.12	listenv_0.9.0
[39] labeling_0.4.2	splines_4.3.1
[41] fastmap_1.1.1	grid_4.3.1
[43] colorspace_2.1-0	cli_3.6.1
[45] magrittr_2.0.3	survival_3.5-5
[47] utf8_1.2.3	withr_2.5.0
[49] scales_1.2.1	bit64_4.0.5
[51] timechange_0.2.0	rmarkdown_2.23
[53] globals_0.16.2	bit_4.0.5
[55] ragg_1.2.5	hms_1.1.3
[57] coda_0.19-4	evaluate_0.21
[59] rlang_1.1.1	MCMCpack_1.6-3
[61] Rcpp_1.0.11	xtable_1.8-4
[63] glue_1.6.2	xml2_1.3.5
[65] rstudioapi_0.15.0	jsonlite_1.8.7
[67] R6_2.5.1	systemfonts_1.0.4