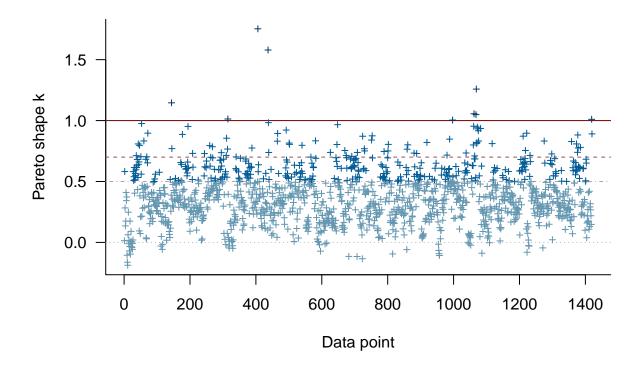
Model comp

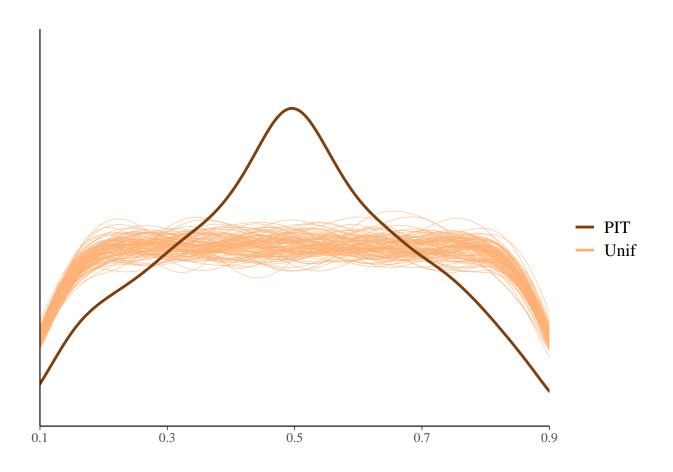
Zhengfan Wang

IN-SAMPLE

```
## [1] 4000 1420
##
## Computed from 4000 by 1420 log-likelihood matrix
##
##
            Estimate
## elpd_loo
               861.2 51.8
## p_loo
               405.4 21.1
## looic
             -1722.4 103.7
## Monte Carlo SE of elpd_loo is NA.
##
## Pareto k diagnostic values:
##
                             Count Pct.
                                           Min. n_eff
## (-Inf, 0.5]
                  (good)
                             1054 74.2%
                                            387
    (0.5, 0.7]
                  (ok)
                                   19.6%
                              279
                                            102
##
##
      (0.7, 1]
                  (bad)
                               78
                                    5.5%
                                            14
                                    0.6%
      (1, Inf)
                  (very bad)
                                            2
##
## See help('pareto-k-diagnostic') for details.
```

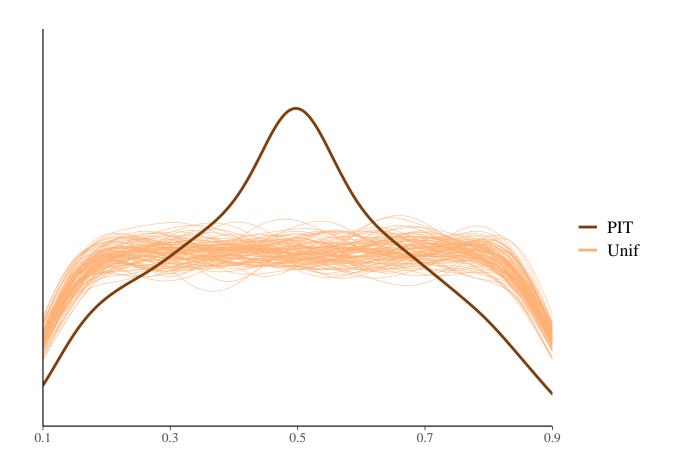
PSIS diagnostic plot





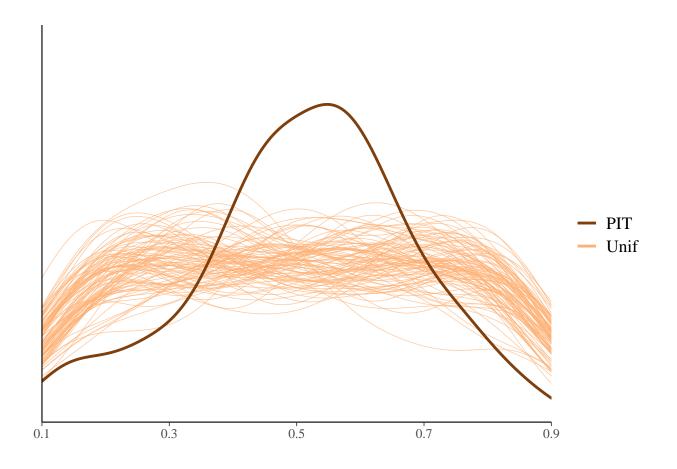
admin data

```
length(which(standata$getj_i == 1)) #num of obs
## [1] 1094
loo_by_source(source=1,dat = standata,loo=loo1)
```



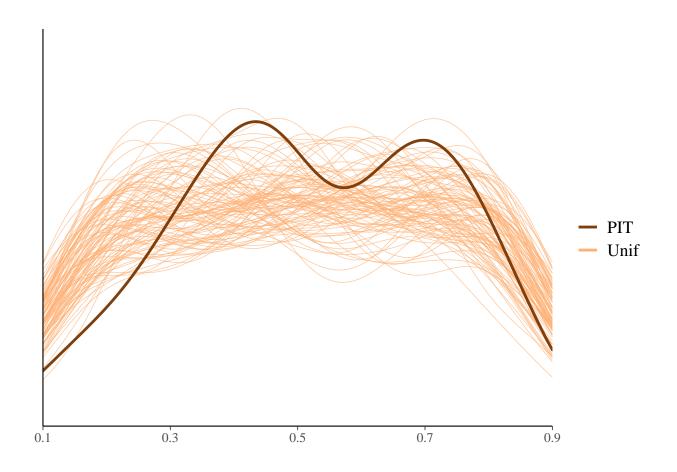
HMIS

```
length(which(standata$getj_i == 2)) #num of obs
## [1] 116
loo_by_source(source=2,dat = standata,loo=loo1)
```



subnat LR

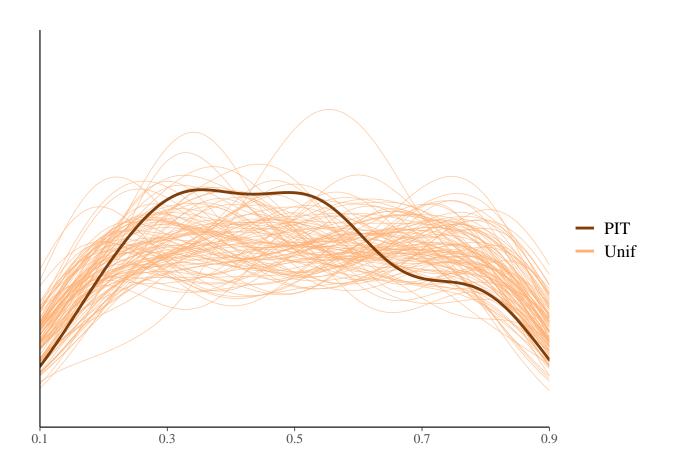
```
length(which(standata$getj_i == 3)) #num of obs
## [1] 120
loo_by_source(source=3,dat = standata,loo=loo1)
```



survey

```
length(which(standata$getj_i == 4)) #num of obs

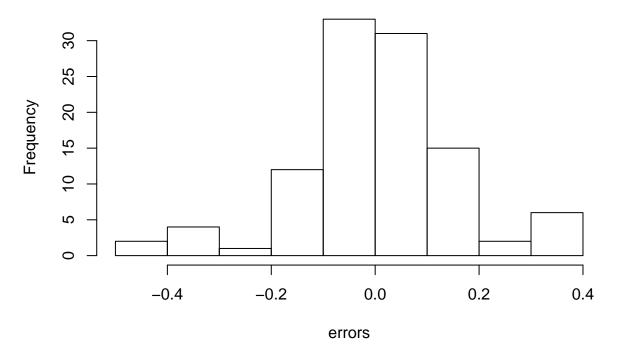
## [1] 90
loo_by_source(source=4,dat = standata,loo=loo1)
```



OUT-SAMPLE

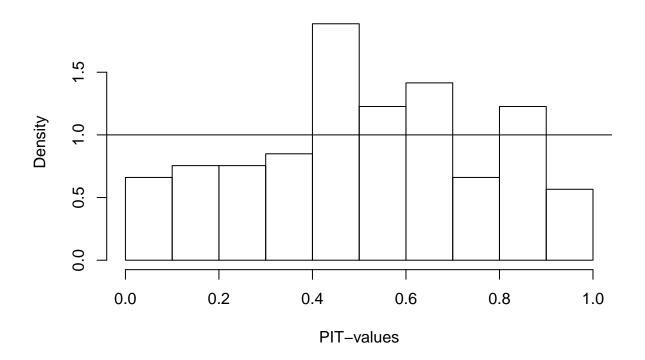
```
fit <- readRDS(file = "rdsoutput/base_val.rds")
stan.data <- readRDS(file = "output/stan_data/nhs_val.rds")
array <- rstan::extract(fit)
# indices of test set
getitest <- setdiff(seq(1,stan.data$N), stan.data$getitrain_k)
# errors
pred <- apply(array$prep,2,median)
errors <- stan.data$Y[getitest] - pred[getitest]
hist(errors)</pre>
```

Histogram of errors



```
# summarize
mean(errors)
## [1] 0.007586468
median(errors)
## [1] 0.002718375
mean(abs(errors))
## [1] 0.1088321
median(abs(errors))
## [1] 0.07437609
ntest <- length(getitest)</pre>
ntest
## [1] 106
# PIT
pit.j <- rep(NA, ntest)</pre>
for (j in 1:ntest){
  i <- getitest[j]</pre>
  yrepi.s <- array$prep[,i]</pre>
  pit.j[j] <- mean(yrepi.s <= stan.data$Y[i])</pre>
hist(pit.j, freq = F, xlab = "PIT-values", main = "Predicting last obs") # should look uniform
abline(h=1)
```

Predicting last obs



```
p < -0.1
qbinom(c(0.1, 0.9), size = ntest, prob = p)/ntest # 80% PI for prop of left out obs in one bin of PIT v
## [1] 0.06603774 0.14150943
qbinom(c(0.25, 0.75), size = ntest, prob = p)/ntest # 50% PI for prop of left out obs in one bin of PIT
## [1] 0.0754717 0.1226415
p < -0.25
qbinom(c(0.1, 0.9), size = ntest, prob = p)/ntest # 80% PI for prop of left out obs in one bin of PIT v
## [1] 0.1981132 0.3018868
qbinom(c(0.25, 0.75), size = ntest, prob = p)/ntest # 50% PI for prop of left out obs in one bin of PIT
## [1] 0.2169811 0.2735849
# coverage follows from pit
mean(pit.j < 0.025) # % below 95% PI
## [1] 0
mean(pit.j < 0.05) # % below 90% PI
## [1] 0.03773585
mean(pit.j < 0.1)
## [1] 0.06603774
```

```
mean(pit.j > 0.975) # % above 95% PI

## [1] 0
mean(pit.j > 0.95)

## [1] 0.01886792
mean(pit.j > 0.9)

## [1] 0.05660377
```