BFL synthetic PHMRC example

Yu (Zoey) Zhu

2025-05-09

```
library(rstan)
library(caret)
library(LCVA)
```

This supplementary R Markdown file provides example code for fitting Local and Bayesian Federated Learning (BFL) models under three different label shift settings: (1) No within-target label shift, (2) Mild within-target label shift, and (3) Severe within-target label shift. The goal is to evaluate model performance on a given target domain (e.g., "AP") by computing both the Top Cause Accuracy and the Cause-Specific Mortality Fraction (CSMF) Accuracy.

The "execute_balance_local_fit" and "execute_unbalance_local_fit" functions generate simulated data for each domain, fit the local LCVA models for each domain, and extract the corresponding posterior phi for subsequent BFL models. These functions also report the local model's CSMF and top-cause accuracy for the target site.

The "run_BFL_balance_model" and "run_BFL_unbalance_model" functions fit various BFL models and provide both CSMF and top-cause accuracy predictions for the target site. Additionally, the outputs also include CSMF estimates and individual-level cause predictions for the unlabeled samples.

Setup

We consider the target domain is "AP" and have 20% labeled data in the target domain (label shift case (1) & (2)).

We apply K = 5 for the latent class when fitting LCVA base models.

```
test_site_balance <- "AP"
test_site_unbalance <- "AP"
sites <- c("Mexico", "AP", "Bohol", "Dar", "Pemba", "UP")
K <- 5
miss_prop <- 0.8
model_types_balance <- c("domain", "partial", "base", "mix")
model_types_unbalance <- c("domain", "partial", "mix")</pre>
```

Case I: No within-target label shift

We evaluate the local models (local-self and local-avg), and BFL models (domain, partial, base, mix) in the 'no within-target label shift' case.

```
source("execute_balance_local_fit.R")
source("run_BFL_model.R")
source("execute balance domain.R")
source("execute_balance_partial.R")
source ("execute balance base.R")
source("execute_balance_mix.R")
```

Local Fit

```
local_res_balance <- execute_balance_local_fit(</pre>
 test_site = test_site_balance,
 sites = sites.
 K = K
 miss_prop = miss_prop
posterior phi full <- local res balance$posterior phi full
sim_data_filtered_list <- local_res_balance$sim_data_filtered_list
LCVA_local_model_test_obs_fit <- local_res_balance$LCVA_local_model_test_obs_fit
# CSMF ACC
cat("Case I : Local-self :", local_res_balance$csmf_acc_local[test_site_balance], "\n")
## Case I : Local-self : 0.6148636
cat("Case I : Local-avg :",
    mean(local res balance$csmf acc local[names(local res balance$csmf acc local) != test site balance
## Case I : Local-avg : 0.5925055
# Top Cause ACC
cat("Case I : Local-self :", local_res_balance$acc_local[test_site_balance], "\n")
## Case I : Local-self : 0.2485921
cat("Case I : Local-avg :".
    mean(local_res_balance$acc_local[names(local_res_balance$acc_local) != test_site_balance]), "\n")
## Case I : Local-avg : 0.2366854
BFL Fit
BFL_results_balance <- list()</pre>
for (model_type in model_types_balance) {
 BFL_results_balance[[model_type]] <- run_BFL_balance_model(</pre>
   model_type = model_type,
```

```
test_site = test_site_balance,
   sites = sites.
    sim_data_filtered_list = sim_data_filtered_list,
    posterior_phi_full = posterior_phi_full,
    LCVA_local_model_test_obs_fit = LCVA_local_model_test_obs_fit
  )
}
## Stacking 3 chains, with 152 data points and 1500 posterior draws;
## using stan optimizer, max iterations = 1e+05
## Total elapsed time for approximate LOO and stacking = 16.87 s
## [1] "Number of posterior draws from each chain:"
## w[1] w[2] w[3]
## 1222 1458 1320
# CSMF ACC
for (model_type in model_types_balance) {
   cat("Case I : BFL-", model_type, ":", BFL_results_balance[[model_type]]$csmf_acc, "\n")
}
## Case I : BFL- domain : 0.6893699
## Case I : BFL- partial : 0.7107892
## Case I : BFL- base : 0.5910754
## Case I : BFL- mix : 0.6998775
# Top Cause ACC
for (model_type in model_types_balance) {
  cat("Case I : BFL-", model_type, ":", BFL_results_balance[[model_type]]$acc, "\n")
}
## Case I : BFL- domain : 0.3459372
## Case I : BFL- partial : 0.3419147
## Case I : BFL- base : 0.3082368
## Case I : BFL- mix : 0.3419147
# example of CSMF estimation from BFL-domain
BFL_results_balance[["domain"]]$csmf
##
                         Cirrhosis
                                                           Epilepsy
##
                       0.035859600
                                                        0.005989274
##
                         Pneumonia
                                                               COPD
##
                       0.077476216
                                                        0.008486720
       Acute Myocardial Infarction
##
                                                              Fires
##
                       0.103730125
                                                        0.022662488
##
                     Renal Failure
                                                               AIDS
##
                       0.111075462
                                                        0.162055543
##
                       Lung Cancer
                                                           Maternal
##
                       0.001751973
                                                        0.063485187
##
                          Drowning
                                    Other Cardiovascular Diseases
##
                       0.032734139
                                                        0.014222926
```

```
## Other Non-communicable Diseases
                                                                Falls
##
                        0.018279802
                                                          0.015194428
                                                         Road Traffic
##
                             Stroke
##
                        0.063322962
                                                          0.016319234
##
           Bite of Venomous Animal
                                                             Diabetes
##
                        0.007956630
                                                          0.031037776
##
         Other Infectious Diseases
##
                        0.092282992
                                                          0.019473165
##
                            Suicide
                                                      Other Injuries
##
                        0.009946256
                                                          0.015708609
##
                    Cervical Cancer
                                                              Malaria
                        0.001363701
                                                          0.006491573
##
##
                             Asthma
                                                  Diarrhea/Dysentery
                        0.003211193
                                                          0.012320153
##
##
                  Colorectal Cancer
                                                             Homicide
##
                        0.001450121
                                                          0.024795011
##
                      Breast Cancer
                                                  Leukemia/Lymphomas
                                                          0.001278824
##
                        0.001295666
##
                                                     Prostate Cancer
                         Poisonings
##
                        0.014332290
                                                          0.001281151
##
                  Esophageal Cancer
                                                      Stomach Cancer
##
                        0.001267687
                                                          0.001861126
```

example of individual prediction for the unlabeled sample from BFL-domain (first 10 samples)
BFL_results_balance[["domain"]]\$cause_pred[1:10]

```
## [1] "Pneumonia" "Pneumonia"
## [3] "Other Cardiovascular Diseases" "Pneumonia"
## [5] "TB" "Maternal"
## [7] "Diarrhea/Dysentery" "Other Infectious Diseases"
## [9] "Maternal" "Maternal"
```

Case II: Mild within-target label shift

We evaluate the local models (local-self and local-avg), and BFL models (domain, partial, mix) in the 'mild within-target label shift' case.

```
source("execute_unbalance_local_fit.R")
source("run_BFL_model.R")
source("execute_unbalance_domain.R")
source("execute_unbalance_partial.R")
source("execute_unbalance_mix.R")
unbalanced_cases <- c("MILD", "SEVERE")</pre>
```

Local Fit

```
case = unbalanced_cases[1]
cat("Running Label Shift Case:", case, "\n")
```

```
## Running Label Shift Case: MILD
local_res_unbalance <- execute_unbalance_local_fit(</pre>
    test_site = test_site_unbalance,
    sites = sites,
   K = K
   miss_prop = miss_prop,
   unbalanced case = case
# CSMF ACC
cat("Case II : Local-self :", local_res_unbalance$csmf_acc_local[test_site unbalance], "\n")
## Case II : Local-self : 0.4790563
cat("Case II : Local-avg :",
   mean(local_res_unbalance$csmf_acc_local[names(local_res_unbalance$csmf_acc_local) != test_site_unba
## Case II : Local-avg : 0.5441316
# Top Cause ACC
cat("Case II : Local-self :", local_res_unbalance$acc_local[test_site_unbalance], "\n")
## Case II : Local-self : 0.2886179
cat("Case II : Local-avg :".
    mean(local_res_unbalance$acc_local[names(local_res_unbalance$acc_local) != test_site_unbalance]), "
## Case II : Local-avg : 0.2260163
BFL Fit
  BFL_results_unbalance <- list()</pre>
  for (model_type in model_types_unbalance) {
   BFL_results_unbalance[[model_type]] <- run_BFL_unbalance_model(</pre>
     model_type = model_type,
     test_site = test_site_unbalance,
      sites = sites,
      sim_data_filtered_list = local_res_unbalance$sim_data_filtered_list,
      sim_data_target_domain_list = local_res_unbalance$sim_data_target_domain_list,
      posterior phi full = local res unbalance$posterior phi full,
     LCVA_local_model_test_obs_fit = local_res_unbalance$LCVA_local_model_test_obs_fit
    )
 }
## Stacking 3 chains, with 652 data points and 1500 posterior draws;
## using stan optimizer, max iterations = 1e+05
##
```

```
## Total elapsed time for approximate LOO and stacking = 18 s
## [1] "Number of posterior draws from each chain:"
## w[1] w[2] w[3]
## 535 1130 2335
# CSMF ACC
for (model_type in model_types_unbalance) {
  cat("Case II : BFL-", model_type, ":", BFL_results_unbalance[[model_type]]$csmf_acc, "\n")
## Case II : BFL- domain : 0.5892829
## Case II : BFL- partial : 0.5029196
## Case II : BFL- mix : 0.5232224
# Top Cause ACC
for (model_type in model_types_unbalance) {
 cat("Case II : BFL-", model_type, ":", BFL_results_unbalance[[model_type]]$acc, "\n")
## Case II : BFL- domain : 0.296748
## Case II : BFL- partial : 0.2804878
## Case II : BFL- mix : 0.2764228
```

Case III: Severe within-target label shift

We evaluate the local models (local-self and local-avg), and BFL models (domain, partial, mix) in the 'severe within-target label shift' case.

```
Local Fit
case = unbalanced_cases[2]
cat("Running Label Shift Case:", case, "\n")
## Running Label Shift Case: SEVERE
local_res_unbalance <- execute_unbalance_local_fit(</pre>
   test_site = test_site_unbalance,
   sites = sites,
   K = K
   miss_prop = miss_prop,
   unbalanced case = case
)
# CSMF ACC
cat("Case III : Local-self :", local_res_unbalance$csmf_acc_local[test_site_unbalance], "\n")
## Case III : Local-self : 0.04261776
```

```
cat("Case III : Local-avg :",
    mean(local res_unbalance$csmf_acc_local[names(local res_unbalance$csmf_acc_local) != test_site_unbalance$csmf_acc_local
## Case III : Local-avg : 0.3080023
# Top Cause ACC
cat("Case III : Local-self :", local_res_unbalance$acc_local[test_site_unbalance], "\n")
## Case III : Local-self : 0.008842444
cat("Case III : Local-avg :",
   mean(local_res_unbalance$acc_local[names(local_res_unbalance$acc_local) != test_site_unbalance]), "
## Case III : Local-avg : 0.1729904
BFL Fit
  BFL_results_unbalance <- list()
  for (model_type in model_types_unbalance) {
   BFL_results_unbalance[[model_type]] <- run_BFL_unbalance_model(</pre>
     model_type = model_type,
     test_site = test_site_unbalance,
     sites = sites.
     sim data filtered list = local res unbalance$sim data filtered list,
     sim_data_target_domain_list = local_res_unbalance$sim_data_target_domain_list,
     posterior_phi_full = local_res_unbalance$posterior_phi_full,
     LCVA local model test_obs_fit = local_res_unbalance$LCVA_local_model_test_obs_fit
   )
 }
## Stacking 3 chains, with 157 data points and 1500 posterior draws;
## using stan optimizer, max iterations = 1e+05
##
## Total elapsed time for approximate LOO and stacking = 16.56 s
## [1] "Number of posterior draws from each chain:"
## w[1] w[2] w[3]
## 1620 670 1710
# CSMF ACC
for (model_type in model_types_unbalance) {
   cat("Case III : BFL-", model_type, ":", BFL_results_unbalance[[model_type]]$csmf_acc, "\n")
## Case III : BFL- domain : 0.6026971
## Case III: BFL- partial: 0.614353
## Case III : BFL- mix : 0.6067743
```

```
# Top Cause ACC
for (model_type in model_types_unbalance) {
   cat("Case III : BFL-", model_type, ":", BFL_results_unbalance[[model_type]]$acc, "\n")
}

## Case III : BFL- domain : 0.3223473
## Case III : BFL- partial : 0.3118971
## Case III : BFL- mix : 0.3191318
```