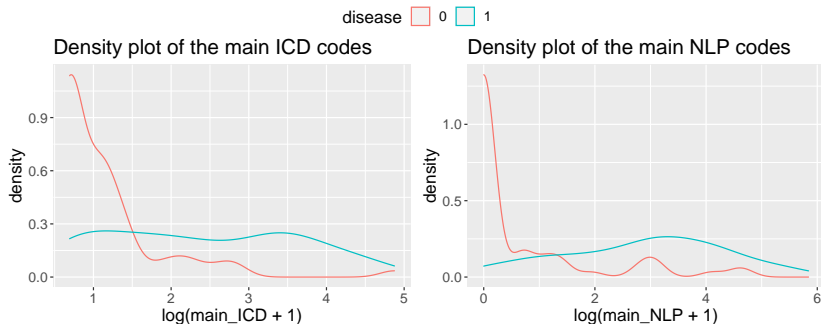


## Module 3: Semi-supervised learning (PheCAP)

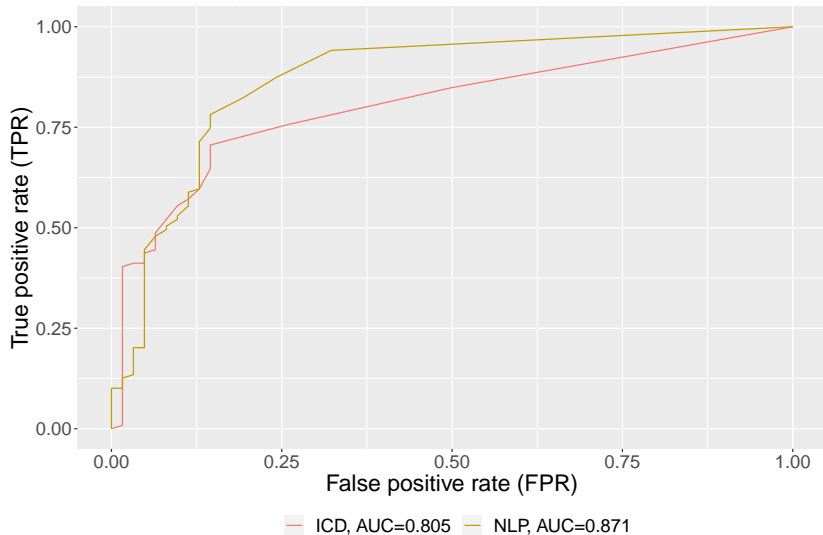
# Surrogates for CAD



The more the disease-related codes and NLP mentions, the more **likely** the patient has the disease.

# ROC Surrogates

The operating receiver characteristic (ROC) curve



## Step 1: SAFE

```
surrogates <- list(  
  PhecapSurrogate(  
    variable_names = "main_ICD",  
    lower_cutoff = 1, upper_cutoff = 10),  
  PhecapSurrogate(  
    variable_names = "main_NLP",  
    lower_cutoff = 1, upper_cutoff = 10))  
  
feature_selected <- phecap_run_feature_extraction(data, surrogates)  
feature_selected  
  
## Feature(s) selected by surrogate-assisted feature extraction (SAFE)  
## [1] "main_ICD" "main_NLP" "NLP56"      "NLP93"      "NLP274"     "NLP306"
```

## Step 2: Orthogonalization + supervised learning

```
phecap_lasso <- phecap_train_phenotyping_model(data, surrogates, feature_selected,  
                                              method = "lasso_cv")
```

```
phecap_lasso
```

```
## Phenotyping model:
```

```
## $lasso_cv
```

##	(Intercept)	main_ICD	main_NLP	healthcare_utilization
##	1.9258667	0.2157399	1.1666409	-0.9772753
##	NLP56	NLP93	NLP274	NLP306
##	0.0000000	-0.3242900	0.0000000	0.0000000

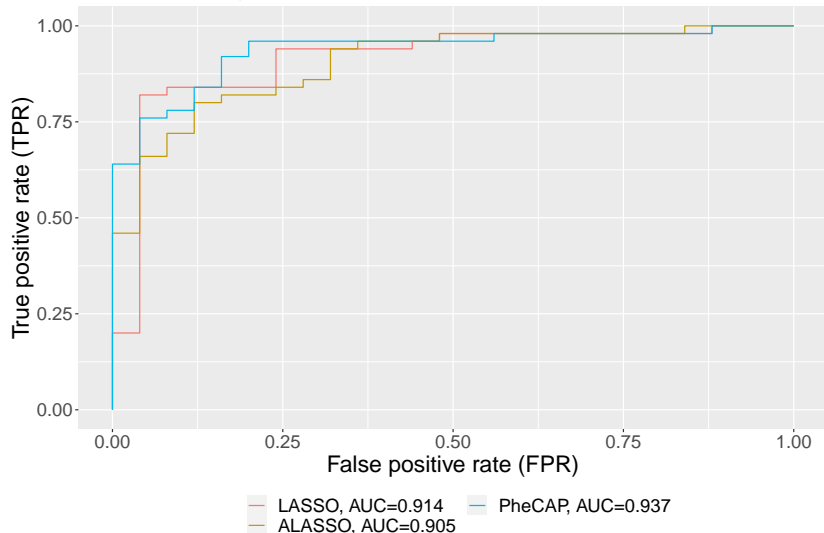
```
##
```

```
## AUC on training data: 0.93
```

```
## Average AUC on random splits: 0.889
```

# Supervised learning (LASSO, ALASSO) vs. PheCAP

The operating receiver characteristic (ROC) curve



# Supervised learning (LASSO, ALASSO) vs. PheCAP at FPR = 0.10

```
get_roc_parameter(0.1, roc_full_lasso)
```

```
##      cutoff pos.rate FPR  TPR      PPV      NPV      F1
## 1 0.8573308 0.5933333 0.1 0.84 0.9438202 0.7377049 0.8888889
```

```
get_roc_parameter(0.1, roc_full_lasso)
```

```
##      cutoff pos.rate FPR  TPR      PPV      NPV      F1
## 1 0.9228109 0.5133333 0.1 0.72 0.9350649 0.6164384 0.8135593
```

```
roc_full_phecap <- get_roc(y_true = test_y, y_score = y_hat_phecap) %>% data.frame()
get_roc_parameter(0.1, roc_full_phecap)
```

```
##      cutoff pos.rate FPR  TPR      PPV      NPV      F1
## 1 0.8342308 0.5533333 0.1 0.78 0.939759 0.6716418 0.852459
```

# Supervised learning vs. PheCAP for different training size

- ▶ Randomly sample training size = 50, 70, 90
- ▶ Use the remaining data as the test set
- ▶ Repeat 600 times

```
auc_phecap <- validate_phecap(dat = labeled_data,  
                              orig_data = ehr_data,  
                              surrogates = surrogates,  
                              feature_selected = feature_selected,  
                              nsim = 600,  
                              n.train = c(50, 70, 90))
```



# Supervised learning vs. PheCAP for different training size

Area under the ROC curve (AUC) from 600 simulations

