# Demo

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## **Imports**

```
#import libraries
```

## Load Model

```
clf_LR <- readRDS(file = "../models/LogisticRegressionClassifier_Full.Rds")
clf_RF <- readRDS(file = "../models/RandomForestClassifier_Full.Rds")</pre>
```

#### Load Test Set

```
test <- readRDS(file = "../data/PIMA_test.Rds")</pre>
```

## **Predicting Cases**

```
woman5 <- test %>%
  filter(row_number() == 5)

woman18 <- test %>%
  filter(row_number() == 18)

woman37 <- test %>%
  filter(row_number() == 37)
```

We can see the single woman's observation from the test set. Now we predict what's the probability she has diabetes.

## Logistic Regression Classifier

```
predict(clf_LR, type = "response", newdata = woman5)
```

```
## 1
## 0.7885501
```

And in fact we check and woman5 does have diabetes.

```
predict(clf_LR, type = "response", newdata = woman18)

##     1
## 0.1652994
```

And once again our model was correct, woman18 does not have diabetes.

```
predict(clf_LR, type = "response", newdata = woman37)
## 1
```

## 0.2049365

Our model is off here. It gives us a low probability prediction but the woman does indeed have diabetes. This is the kind of examples that we wanted to minimize by choosing a 0.3 cutoff. However, a lower cutoff would have been needed for this woman (at 0.2) to classify her correctly.

#### Random Forest Classifer

```
predict(clf_RF, type = "prob", newdata = woman5)[2]

## [1] 0.848
predict(clf_RF, type = "prob", newdata = woman18)[2]

## [1] 0.378
predict(clf_RF, type = "prob", newdata = woman37)[2]

## [1] 0.45
```

In these three instances our models are close, but the random forest is overpredicting for these specific three cases.