# > Imports

[ ] \( \rightarrow 2 \) cells hidden

## Classification

```
import pandas as pd
V1 = pd.read_excel("ISAC_V1.xlsx", engine="openpyxl")
V1.head()
```

 $\overline{\mathbf{x}}$ 

	Unnamed:	Chip_Code	Chip_Type	Age	Blood_Month_sample	French_Residence
0	XPW0007	XPW0007	ISAC_V1	9	3	
1	XPW0011	XPW0011	ISAC_V1	25	5	
2	XPW0013	XPW0013	ISAC_V1	59	5	
3	XPW0017	XPW0017	ISAC_V1	49	5	
4	XPW0018	XPW0018	ISAC_V1	9	5	

5 rows x 129 columns

```
target_1 = [
    "Allergy_Present",
    "Respiratory_Allergy",
    "Food Allergy",
    "Venom_Allergy",
   "Severe_Allergy",
   "Type_of_Food_Allergy_Other",
    "Type_of_Respiratory_Allergy_IGE_Pollen_Herb",
    "Type_of_Respiratory_Allergy_IGE_Pollen_Tree",
    "Type_of_Respiratory_Allergy_IGE_Dander_Animals",
    "Type_of_Respiratory_Allergy_IGE_Mite_Cockroach",
    "Type_of_Respiratory_Allergy_IGE_Molds_Yeast",
    "Type_of_Respiratory_Allergy_ARIA",
    "Type_of_Respiratory_Allergy_CONJ",
    "Type_of_Respiratory_Allergy_IGE_Pollen_Gram",
    "Type_of_Respiratory_Allergy_GINA",
    "Type of Food Allergy Aromatics",
```

```
"Type_of_Food_Allergy_Cereals_&_Seeds",
    "Type_of_Food_Allergy_Egg",
    "Type of Food Allergy Fish",
    "Type_of_Food_Allergy_Fruits_and_Vegetables",
    "Type_of_Food_Allergy_Mammalian_Milk",
    "Type of Food Allergy Oral Syndrom",
    "Type_of_Food_Allergy_Other_Legumes",
    "Type_of_Food_Allergy_Peanut",
    "Type_of_Food_Allergy_Shellfish",
    "Type_of_Food_Allergy_TP0",
    "Type_of_Food_Allergy_Tree_Nuts",
    "Type_of_Venom_Allergy_ATCD_Venom",
    "Type_of_Venom_Allergy_IGE_Venom",
1
extra_columns = [
    "Chip Type",
    "Chip_Code",
    "French_Region",
    "French Residence Department"
extra = ['History_of_food_anaphylaxis','First_degree_family_history_of_atopy',
         'History of hymenoptera venom anaphylaxis', 'Mammalian meat']
extra_1 = ["Conjunctivitis", "Oral_Syndrom", "Cardiovascular_symptoms", "Respir
Gina = ["GINA_(asthma)_0", "GINA_(asthma)_1", "GINA_(asthma)_2", "GINA_(asthma)
inconnu = ["Treatment_of_athsma_9", "Treatment_of_rhinitis_9", "General_cofactor
           "Age_of_onsets_9", "ARIA_(rhinitis)_9", "GINA_(asthma)_9", "Treatmer
Aria = ["ARIA_(rhinitis)_9", "ARIA_(rhinitis)_0", "ARIA_(rhinitis)_1", "ARIA_(r
import pandas as pd
import numpy as np
from sklearn.model_selection import StratifiedKFold
from sklearn.ensemble import RandomForestClassifier
from sklearn.linear_model import LogisticRegression
from sklearn.svm import SVC
from xgboost import XGBClassifier
from sklearn.metrics import (
    f1_score, accuracy_score, recall_score,
    precision_score, confusion_matrix, roc_auc_score, roc_curve
from imblearn.over_sampling import SMOTE
import plotly.graph_objects as go
targets = ["Allergy_Present", "Respiratory_Allergy", "Food_Allergy", "Venom_All
```

```
models = {
    "RandomForest": RandomForestClassifier(random_state=42),
    "XGBoost": XGBClassifier(random_state=42, eval_metric="logloss", use_label_
    "LogisticRegression": LogisticRegression(max_iter=1000, random_state=42),
    "SVM": SVC(probability=True, random_state=42)
}
X = V1.copy()
X.drop(target_1, axis=1, inplace=True)
X.drop(extra_columns, axis=1, inplace=True)
X.drop(extra, axis=1, inplace=True)
X.drop(inconnu, axis=1, inplace=True)
X = X.iloc[:, 1:]
results = []
kfold = StratifiedKFold(n_splits=10, shuffle=True, random_state=42)
for target in targets:
    y = V1[target]
    for model_name, base_model in models.items():
        f1_class0_scores, f1_class1_scores = [], []
        precision_scores, acc_scores, recall_scores, auc_scores = [], [], [], [
        for train_idx, test_idx in kfold.split(X, y):
            X_train, X_test = X.iloc[train_idx], X.iloc[test_idx]
            y_train, y_test = y.iloc[train_idx], y.iloc[test_idx]
            smote = SMOTE(random state=42)
            X_train_res, y_train_res = smote.fit_resample(X_train, y_train)
            base_model.fit(X_train_res, y_train_res)
            y_pred = base_model.predict(X_test)
            acc_scores.append(accuracy_score(y_test, y_pred))
            recall_scores.append(recall_score(y_test, y_pred, zero_division=0))
            precision_scores.append(precision_score(y_test, y_pred, average='we
            f1_class0_scores.append(f1_score(y_test, y_pred, pos_label=0, zero_
            f1_class1_scores.append(f1_score(y_test, y_pred, pos_label=1, zero_
            if hasattr(base_model, "predict_proba"):
                y_proba = base_model.predict_proba(X_test)[:, 1]
                auc_scores.append(roc_auc_score(y_test, y_proba))
        base_model.fit(X, y)
        y_pred_full = base_model.predict(X)
        y_proba_full = base_model.predict_proba(X)[:, 1] if hasattr(base_model,
        matrix = confusion_matrix(y, y_pred_full)
```

```
print(f"\nQ Target: {target} | Model: {model_name}")
        print(f" Accuracy: {np.mean(acc_scores):.4f}")
        print(f''@ F1 (0): {np.mean(f1_class0_scores):.4f} | F1 (1): {np.mean(
        print(f" Precision: {np.mean(precision_scores):.4f} | AUC: {np.mean(
        print(" Confusion Matrix:\n", matrix)
        if y_proba_full is not None:
           fpr, tpr, _ = roc_curve(y, y_proba_full)
           fig = go.Figure()
           fig.add_trace(go.Scatter(x=fpr, y=tpr, mode='lines', name=f"{model_
           fig.add_trace(go.Scatter(x=[0, 1], y=[0, 1], mode='lines', name='Ra
            fig.update layout(
               title=f"ROC Curve - {target} - {model_name}",
               xaxis_title="False Positive Rate",
               yaxis_title="True Positive Rate",
               width=700, height=500
            fig.show()
        results.append({
           "Target": target,
           "Model": model name,
           "F1 Class 0": np.mean(f1 class0 scores),
           "F1_Class_1": np.mean(f1_class1_scores),
           "Precision": np.mean(precision_scores),
           "Accuracy": np.mean(acc_scores),
           "Recall": np.mean(recall_scores),
           "AUC_ROC": np.mean(auc_scores) if auc_scores else np.nan
        })
pd.DataFrame(results).to_csv("results_V1_Allergie.csv", index=False)
       Target: Allergy Present | Model: RandomForest
     ✓ Accuracy: 0.9426
      F1 (0): 0.9400 | F1 (1): 0.9449
    Precision: 0.9432 | AUC: 0.981794232370263
    Confusion Matrix:
     [[1121 0]
        2 1228]]
```

## ROC Curve - Allergy\_Present - RandomForest



```
0.8
0.6
0.4
0.2
0.2
0.4
0.6
0.8
1
False Positive Rate
```

```
Target: Allergy_Present | Model: XGBoost
Accuracy: 0.9400

F1 (0): 0.9386 | F1 (1): 0.9414

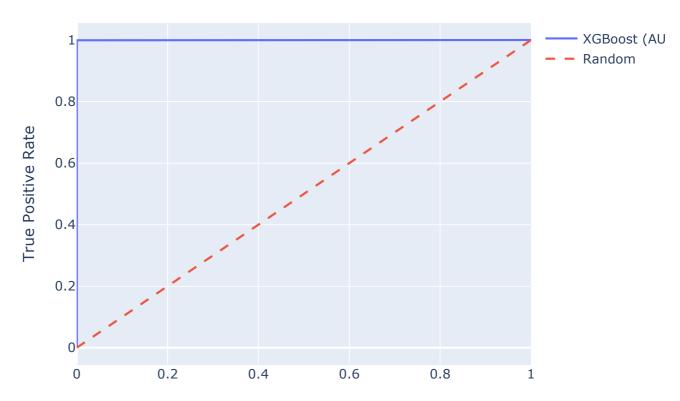
Precision: 0.9414 | AUC: 0.975914184473703

Confusion Matrix:

[[1121 0]

[ 3 1227]]
```

## ROC Curve - Allergy\_Present - XGBoost

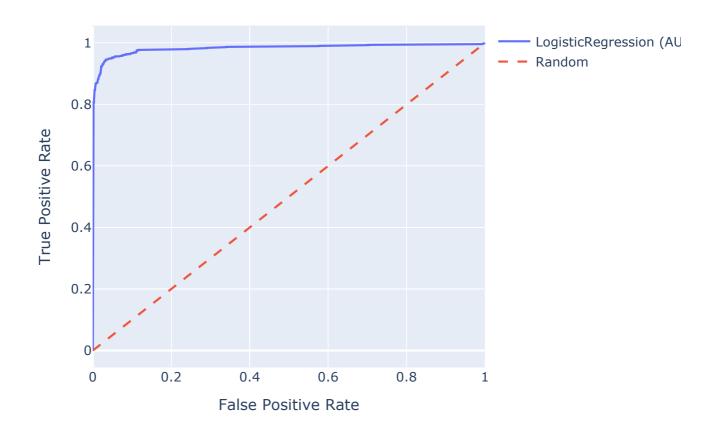


#### False Positive Rate

```
Target: Allergy_Present | Model: LogisticRegression
Accuracy: 0.9409

F1 (0): 0.9402 | F1 (1): 0.9415
Precision: 0.9437 | AUC: 0.9786801851109536
Confusion Matrix:
[[1095 26]
[ 87 1143]]
```

#### ROC Curve - Allergy\_Present - LogisticRegression



```
Target: Allergy_Present | Model: SVM

Accuracy: 0.8532

F1 (0): 0.8598 | F1 (1): 0.8458

Precision: 0.8676 | AUC: 0.9411421556535414

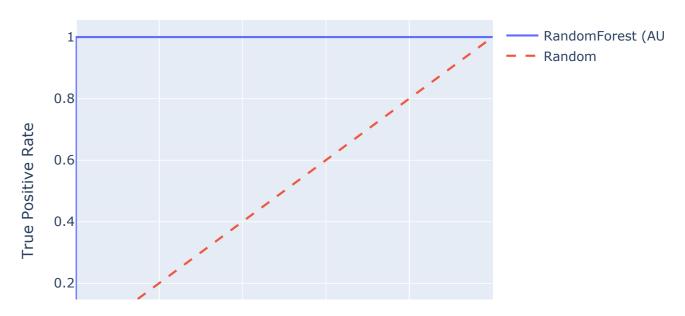
Confusion Matrix:

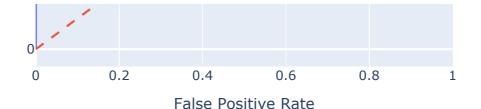
[[1071 50]
[ 246 984]]
```

ROC Curve - Allergy\_Present - SVM

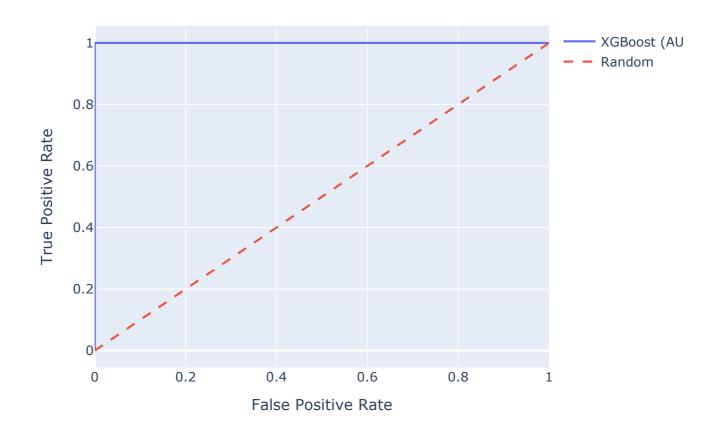


## ROC Curve - Respiratory\_Allergy - RandomForest





#### ROC Curve - Respiratory\_Allergy - XGBoost



```
Target: Respiratory_Allergy | Model: LogisticRegression Accuracy: 0.9230

F1 (0): 0.9308 | F1 (1): 0.9132

Precision: 0.9236 | AUC: 0.9700775202341848

Confusion Matrix:

[[1234 54]
[ 117 946]]
```

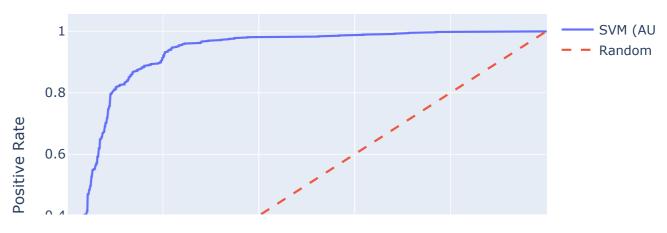
## ROC Curve - Respiratory\_Allergy - LogisticRegression

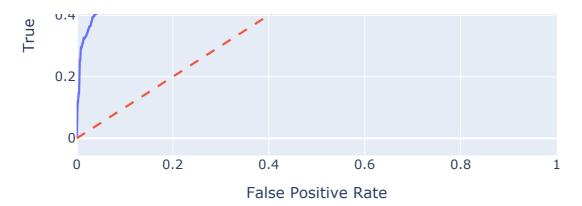


```
Target: Respiratory_Allergy | Model: SVM
Accuracy: 0.8350

F1 (0): 0.8531 | F1 (1): 0.8117
Precision: 0.8354 | AUC: 0.9091027731529515
Confusion Matrix:
[[1166 122]
[ 213 850]]
```

## ROC Curve - Respiratory\_Allergy - SVM





```
Target: Food_Allergy | Model: RandomForest Accuracy: 0.9039

F1 (0): 0.9223 | F1 (1): 0.8739

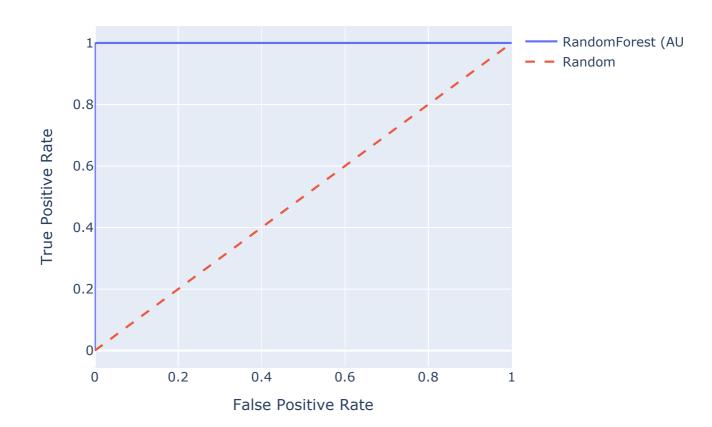
Precision: 0.9044 | AUC: 0.962290753115054

Confusion Matrix:

[[1457 0]

[ 1 893]]
```

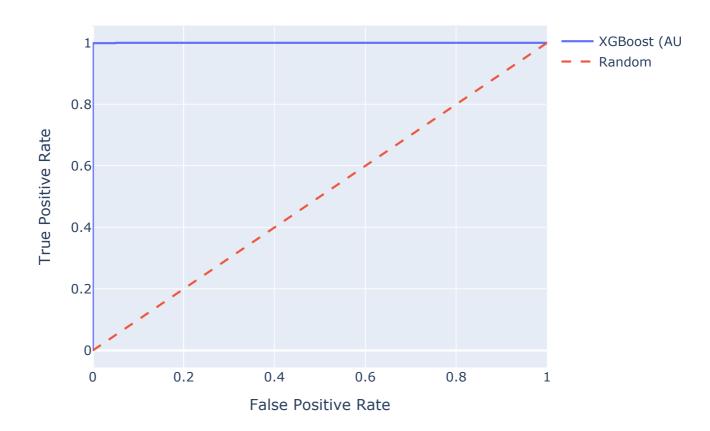
### ROC Curve - Food\_Allergy - RandomForest



Target: Food\_Allergy | Model: XGBoost
Accuracy: 0.9009
 F1 (0): 0.9208 | F1 (1): 0.8676

```
Precision: 0.9008 | AUC: 0.9569938798750026
Confusion Matrix:
[[1457 0]
[ 1 893]]
```

#### ROC Curve - Food\_Allergy - XGBoost

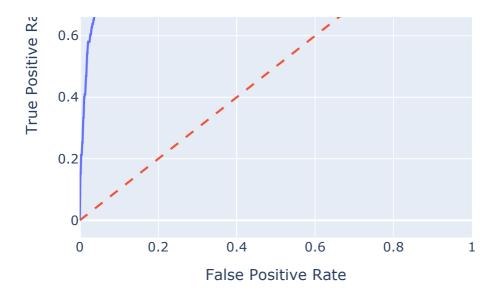


```
Target: Food_Allergy | Model: LogisticRegression
Accuracy: 0.8622

F1 (0): 0.8886 | F1 (1): 0.8190
Precision: 0.8634 | AUC: 0.9215525792334454
Confusion Matrix:
[[1346 111]
[ 179 715]]
```

## ROC Curve - Food\_Allergy - LogisticRegression





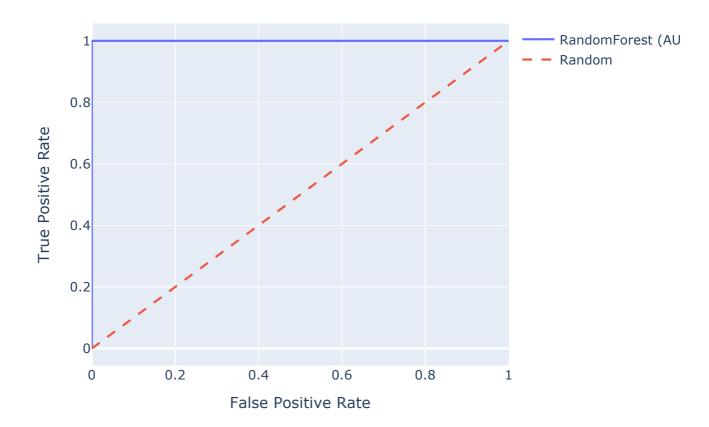
```
Target: Food_Allergy | Model: SVM
Accuracy: 0.8103
F1 (0): 0.8453 | F1 (1): 0.7545
Precision: 0.8121 | AUC: 0.8829788679361001
Confusion Matrix:
[[1324 133]
[ 261 633]]
```

## ROC Curve - Food\_Allergy - SVM



```
Target: Venom_Allergy | Model: RandomForest
Accuracy: 0.9792
F1 (0): 0.9892 | F1 (1): 0.7224
Precision: 0.9783 | AUC: 0.9542713864306785
Confusion Matrix:
[[2254 0]
[ 0 97]]
```

## ROC Curve - Venom\_Allergy - RandomForest



```
Target: Venom_Allergy | Model: XGBoost

Accuracy: 0.9821

F1 (0): 0.9907 | F1 (1): 0.7764

Precision: 0.9819 | AUC: 0.9512725663716814

Confusion Matrix:

[[2254 0]

0 97]]
```

## ROC Curve - Venom\_Allergy - XGBoost





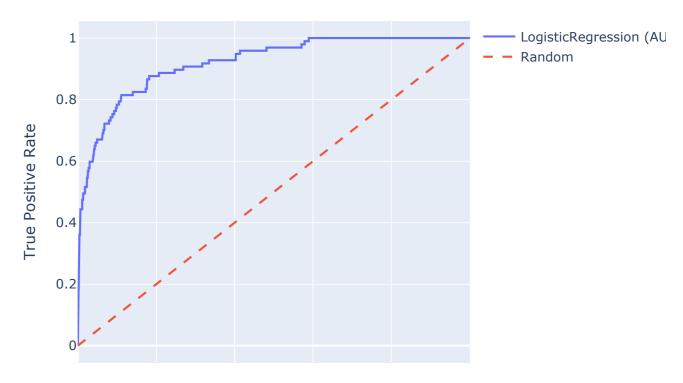
```
Target: Venom_Allergy | Model: LogisticRegression
Accuracy: 0.9145

F1 (0): 0.9543 | F1 (1): 0.3347

Precision: 0.9488 | AUC: 0.8160660766961652

Confusion Matrix:
[[2249 5]
[71 26]]
```

## ROC Curve - Venom\_Allergy - LogisticRegression



0 0.2 0.4 0.6 0.8 1

False Positive Rate

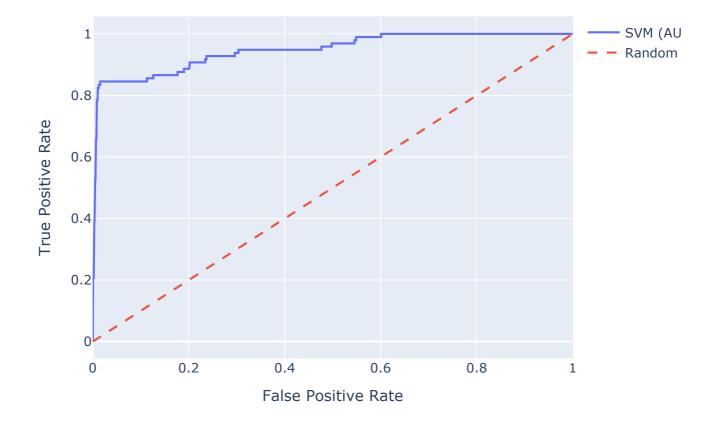
```
Target: Venom_Allergy | Model: SVM
Accuracy: 0.8252

F1 (0): 0.9015 | F1 (1): 0.2187

Precision: 0.9444 | AUC: 0.7936481809242871

Confusion Matrix:
[[2254 0]
[ 97 0]]
```

#### ROC Curve - Venom\_Allergy - SVM

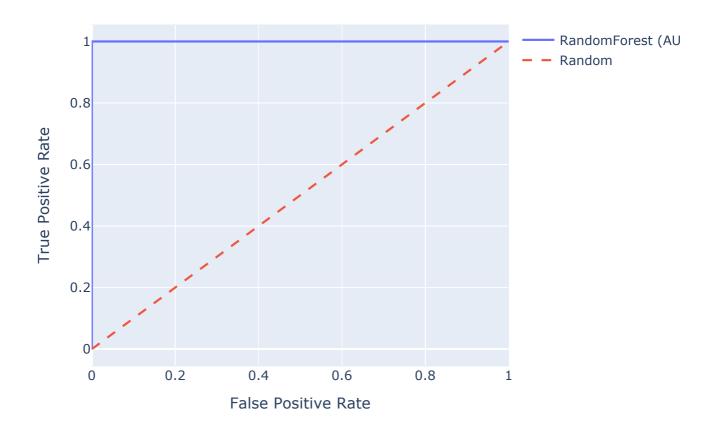


```
import pandas as pd
import numpy as np
from sklearn.model_selection import StratifiedKFold
from sklearn.ensemble import RandomForestClassifier
from sklearn.linear_model import LogisticRegression
from sklearn.svm import SVC
from xgboost import XGBClassifier
from sklearn.metrics import (
```

```
f1_score, accuracy_score, recall_score,
    precision_score, confusion_matrix, roc_auc_score, roc_curve
)
from imblearn.over_sampling import SMOTE
import plotly.graph_objects as go
V1_sev = V1[V1["Allergy_Present"] == 1]
targets = ["Severe_Allergy"]
models = {
    "RandomForest": RandomForestClassifier(random_state=42),
    "XGBoost": XGBClassifier(random_state=42, eval_metric="logloss", use_label_
    "LogisticRegression": LogisticRegression(max_iter=1000, random_state=42),
    "SVM": SVC(probability=True, random_state=42)
X = V1_sev_copy()
X.drop(target_1, axis=1, inplace=True)
X.drop(extra_columns, axis=1, inplace=True)
X.drop(extra, axis=1, inplace=True)
X.drop(inconnu, axis=1, inplace=True)
X = X.iloc[:, 1:]
results_severe = []
kfold = StratifiedKFold(n_splits=10, shuffle=True, random_state=42)
for target in targets:
    y = V1_sev[target]
    for model_name, base_model in models.items():
        f1_class0_scores, f1_class1_scores = [], []
        precision_scores, acc_scores, recall_scores, auc_scores = [], [], [], [
        for train_idx, test_idx in kfold.split(X, y):
            X_train, X_test = X.iloc[train_idx], X.iloc[test_idx]
            y_train, y_test = y.iloc[train_idx], y.iloc[test_idx]
            smote = SMOTE(random_state=42)
            X_train_res, y_train_res = smote.fit_resample(X_train, y_train)
            base_model.fit(X_train_res, y_train_res)
            y_pred = base_model.predict(X_test)
            acc_scores.append(accuracy_score(y_test, y_pred))
            recall_scores.append(recall_score(y_test, y_pred, zero_division=0))
            precision_scores.append(precision_score(y_test, y_pred, average='we
            f1_class0_scores.append(f1_score(y_test, y_pred, pos_label=0, zero_
            f1_class1_scores.append(f1_score(y_test, y_pred, pos_label=1, zero_
            if hasattr(base_model, "predict_proba"):
                y_proba = base_model.predict_proba(X_test)[:, 1]
```

```
auc_scores.append(roc_auc_score(y_test, y_proba))
        base model.fit(X, y)
        y_pred_full = base_model.predict(X)
        y_proba_full = base_model.predict_proba(X)[:, 1] if hasattr(base_model,
        matrix = confusion matrix(y, y pred full)
        print(f"\nQ Target: {target} | Model: {model_name}")
        print(f" Accuracy: {np.mean(acc_scores):.4f}")
        print(f''@ F1 (0): {np.mean(f1_class0_scores):.4f} | F1 (1): {np.mean(
        print(f" Precision: {np.mean(precision_scores):.4f} | AUC: {np.mean(≀
        print(" Confusion Matrix:\n", matrix)
        if y_proba_full is not None:
            fpr, tpr, _ = roc_curve(y, y_proba_full)
            fig = go.Figure()
            fig.add_trace(go.Scatter(x=fpr, y=tpr, mode='lines', name=f"{model_
            fig.add_trace(go.Scatter(x=[0, 1], y=[0, 1], mode='lines', name='Ra
            fig.update_layout(
                title=f"ROC Curve - {target} - {model_name}",
                xaxis_title="False Positive Rate",
                yaxis_title="True Positive Rate",
               width=700, height=500
            )
            fig.show()
        results_severe.append({
            "Target": target,
            "Model": model name,
            "F1_Class_0": np.mean(f1_class0_scores),
            "F1_Class_1": np.mean(f1_class1_scores),
            "Precision": np.mean(precision_scores),
            "Accuracy": np.mean(acc_scores),
            "Recall": np.mean(recall scores),
            "AUC_ROC": np.mean(auc_scores) if auc_scores else np.nan
        })
pd.DataFrame(results_severe).to_csv("results_V1_severe.csv", index=False)
    Target: Severe Allergy | Model: RandomForest
     ✓ Accuracy: 0.8447
    6 F1 (0): 0.7878 | F1 (1): 0.8771
    Precision: 0.8528 | AUC: 0.9318168604651162
    Confusion Matrix:
     [[430 0]
     [ 0 800]]
```

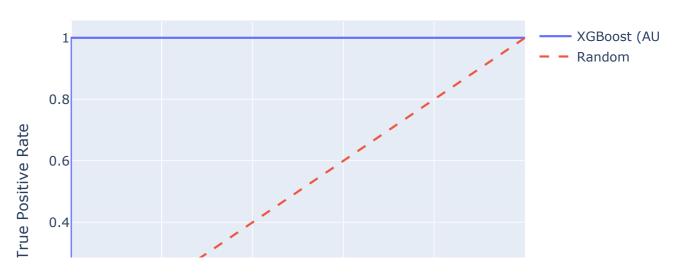
#### KUC Curve - Severe\_Allergy - KandomForest

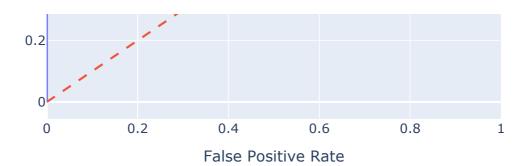


```
Target: Severe_Allergy | Model: XGBoost
Accuracy: 0.8431

F1 (0): 0.7830 | F1 (1): 0.8767
Precision: 0.8484 | AUC: 0.9257848837209302
Confusion Matrix:
[[430 0]
[ 0 800]]
```

## ROC Curve - Severe\_Allergy - XGBoost





```
Target: Severe_Allergy | Model: LogisticRegression Accuracy: 0.8195

F1 (0): 0.7652 | F1 (1): 0.8527

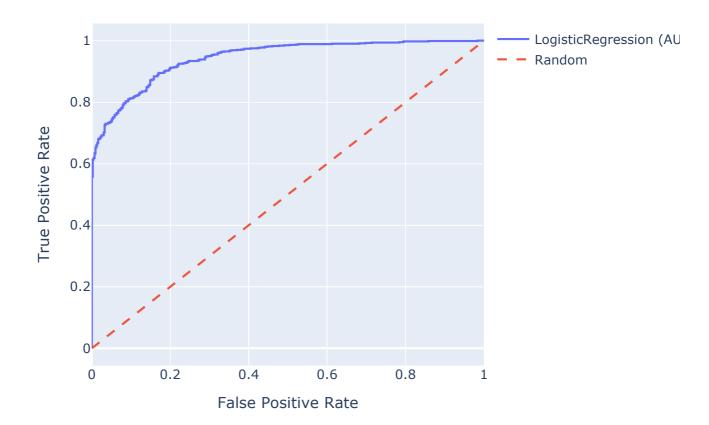
Precision: 0.8359 | AUC: 0.9182558139534883

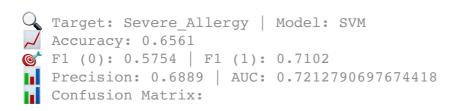
Confusion Matrix:

[[362 68]

[100 700]]
```

## ROC Curve - Severe\_Allergy - LogisticRegression





[[121 309] [ 37 763]]

#### ROC Curve - Severe\_Allergy - SVM



```
import pandas as pd
import numpy as np
from sklearn.model_selection import StratifiedKFold
from sklearn.ensemble import RandomForestClassifier
from sklearn.linear model import LogisticRegression
from sklearn.svm import SVC
from xgboost import XGBClassifier
from sklearn.metrics import (
    f1_score, accuracy_score, recall_score,
    precision_score, confusion_matrix, roc_auc_score, roc_curve
)
from imblearn.over_sampling import SMOTE
import plotly.graph_objects as go
# Données respiratoires
V1_res = V1[V1["Respiratory_Allergy"] == 1]
targets = ["Type_of_Respiratory_Allergy_IGE_Pollen_Herb",
```

```
"Type_of_Respiratory_Allergy_IGE_Pollen_Tree",
    "Type_of_Respiratory_Allergy_IGE_Dander_Animals",
    "Type_of_Respiratory_Allergy_IGE_Mite_Cockroach",
    "Type_of_Respiratory_Allergy_IGE_Molds_Yeast",
    "Type_of_Respiratory_Allergy_ARIA",
    "Type of Respiratory Allergy CONJ",
    "Type_of_Respiratory_Allergy_IGE_Pollen_Gram",
    "Type_of_Respiratory_Allergy_GINA"]
models = {
    "RandomForest": RandomForestClassifier(random_state=42),
    "XGBoost": XGBClassifier(random_state=42, eval_metric="logloss", use_label_
    "LogisticRegression": LogisticRegression(max_iter=1000, random_state=42),
    "SVM": SVC(probability=True, random state=42)
}
X = V1 res.copy()
X.drop(target_1, axis=1, inplace=True)
X.drop(extra_columns, axis=1, inplace=True)
X.drop(extra, axis=1, inplace=True)
X.drop(inconnu, axis=1, inplace=True)
X = X.iloc[:, 1:]
results res = []
kfold = StratifiedKFold(n_splits=10, shuffle=True, random_state=42)
# Boucle principale
for target in targets:
    v = V1 res[target]
    for model_name, base_model in models.items():
        f1_class0_scores, f1_class1_scores = [], []
        precision_scores, acc_scores, recall_scores, auc_scores = [], [], [], []
        print(f"\nQ Target: {target} | Model: {model_name}")
        for train_idx, test_idx in kfold.split(X, y):
            X_train, X_test = X.iloc[train_idx], X.iloc[test_idx]
            y_train, y_test = y.iloc[train_idx], y.iloc[test_idx]
            # Application de SMOTE sur les données d'entraînement
            smote = SMOTE(random state=42)
            X_train_res, y_train_res = smote.fit_resample(X_train, y_train)
            base_model.fit(X_train_res, y_train_res)
            y_pred = base_model.predict(X_test)
            acc_scores.append(accuracy_score(y_test, y_pred))
```

```
recall_scores.append(recall_score(y_test, y_pred, zero_division=0))
            precision_scores.append(precision_score(y_test, y_pred, average='we
            f1_class0_scores.append(f1_score(y_test, y_pred, pos_label=0, zero_
            f1_class1_scores.append(f1_score(y_test, y_pred, pos_label=1, zero_
            if hasattr(base_model, "predict_proba"):
                y_proba = base_model.predict_proba(X_test)[:, 1]
                auc_scores.append(roc_auc_score(y_test, y_proba))
        # Entraînement final sur tout X (sans SMOTE ici, car prédiction globale
        base_model.fit(X, y)
        y_pred_full = base_model.predict(X)
        y_proba_full = base_model.predict_proba(X)[:, 1] if hasattr(base_model,
        matrix = confusion_matrix(y, y_pred_full)
        print(f" Accuracy: {np.mean(acc_scores):.4f}")
        print(f''@ F1 (0): {np.mean(f1_class0_scores):.4f} | F1 (1): {np.mean(
        print(f" Precision: {np.mean(precision_scores):.4f} | AUC: {np.mean(≀
        print(" Confusion Matrix:\n", matrix)
        if y_proba_full is not None:
            fpr, tpr, _ = roc_curve(y, y_proba_full)
            fig = go.Figure()
            fig.add_trace(go.Scatter(x=fpr, y=tpr, mode='lines', name=f"{model_
            fig.add_trace(go.Scatter(x=[0, 1], y=[0, 1], mode='lines', name='Ra
            fig.update_layout(
                title=f"ROC Curve - {target} - {model_name}",
                xaxis_title="False Positive Rate",
                yaxis_title="True Positive Rate",
                width=700, height=500
            )
            fig.show()
        results res.append({
            "Target": target,
            "Model": model_name,
            "F1_Class_0": np.mean(f1_class0_scores),
            "F1_Class_1": np.mean(f1_class1_scores),
            "Precision": np.mean(precision_scores),
            "Accuracy": np.mean(acc scores),
            "Recall": np.mean(recall_scores),
            "AUC_ROC": np.mean(auc_scores) if auc_scores else np.nan
        })
pd.DataFrame(results_res).to_csv("results_V1_respiratoire.csv", index=False)
```



Target: Type\_of\_Respiratory\_Allergy\_IGE\_Pollen\_Herb | Model: RandomFores

```
F1 (0): 0.7659 | F1 (1): 0.7081

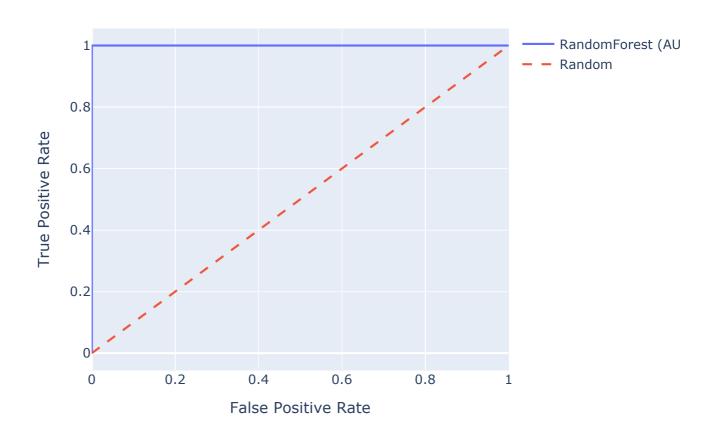
Precision: 0.7456 | AUC: 0.8438125216057385

Confusion Matrix:

[[579 0]

[ 0 484]]
```

#### ROC Curve - Type\_of\_Respiratory\_Allergy\_IGE\_Pollen\_Herb - Randor



```
Target: Type_of_Respiratory_Allergy_IGE_Pollen_Herb | Model: XGBoost  
Accuracy: 0.7516

F1 (0): 0.7695 | F1 (1): 0.7289

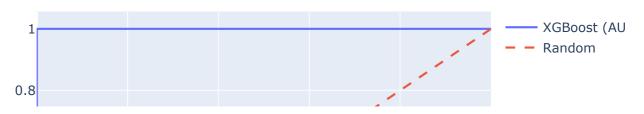
Precision: 0.7556 | AUC: 0.8460580715952443

Confusion Matrix:

[[579 0]

[ 0 484]]
```

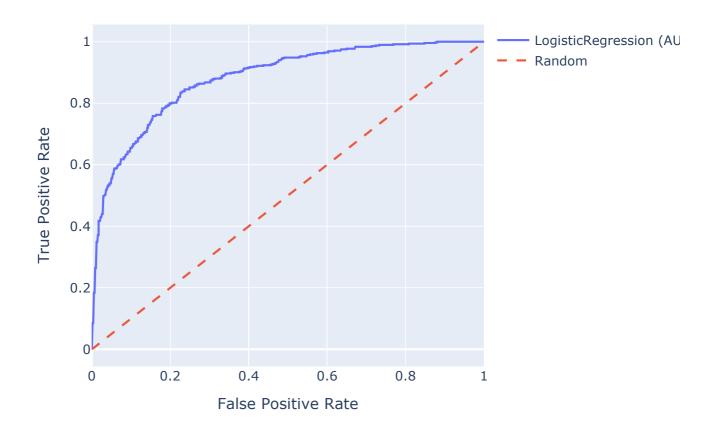
## ROC Curve - Type\_of\_Respiratory\_Allergy\_IGE\_Pollen\_Herb - XGBoos





```
Target: Type_of_Respiratory_Allergy_IGE_Pollen_Herb | Model: LogisticReg
Accuracy: 0.7498
F1 (0): 0.7793 | F1 (1): 0.7098
Precision: 0.7522 | AUC: 0.8320215949356149
Confusion Matrix:
[[509 70]
[152 332]]
```

## ROC Curve - Type\_of\_Respiratory\_Allergy\_IGE\_Pollen\_Herb - Logistic



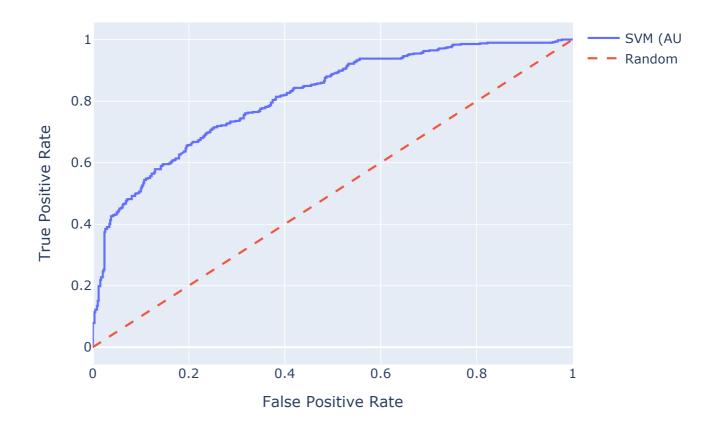
```
Target: Type_of_Respiratory_Allergy_IGE_Pollen_Herb | Model: SVM Accuracy: 0.7140

F1 (0): 0.7395 | F1 (1): 0.6800

Precision: 0.7166 | AUC: 0.7858488585997012

Confusion Matrix:
[[475 104]
[183 301]]
```

#### ROC Curve - Type\_of\_Respiratory\_Allergy\_IGE\_Pollen\_Herb - SVM

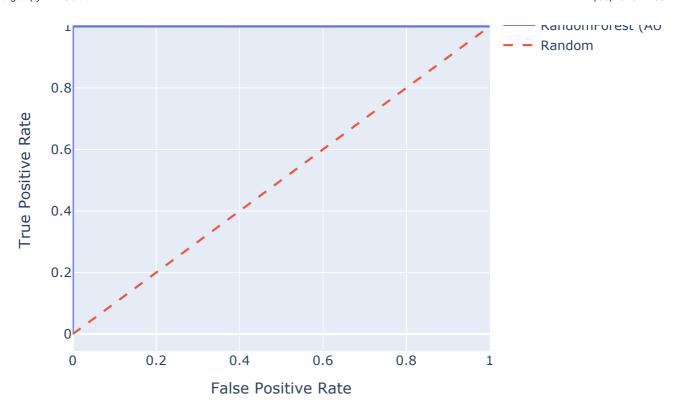


```
Target: Type_of_Respiratory_Allergy_IGE_Pollen_Tree | Model: RandomFores  
Accuracy: 0.8984

F1 (0): 0.7999 | F1 (1): 0.9317
Precision: 0.8980 | AUC: 0.9353242543907101
Confusion Matrix:
[[279 0]
[ 0 784]]
```

### ROC Curve - Type\_of\_Respiratory\_Allergy\_IGE\_Pollen\_Tree - Randon

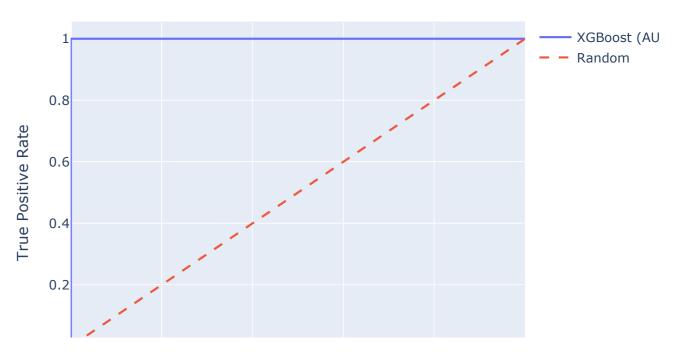
--- DandamEaract (All

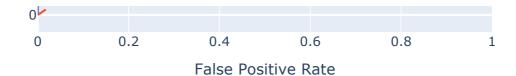


```
Target: Type_of_Respiratory_Allergy_IGE_Pollen_Tree | Model: XGBoost Accuracy: 0.8984

F1 (0): 0.7975 | F1 (1): 0.9321
Precision: 0.8972 | AUC: 0.9320739504283807
Confusion Matrix:
[[279 0]
[ 0 784]]
```

ROC Curve - Type\_of\_Respiratory\_Allergy\_IGE\_Pollen\_Tree - XGBoos





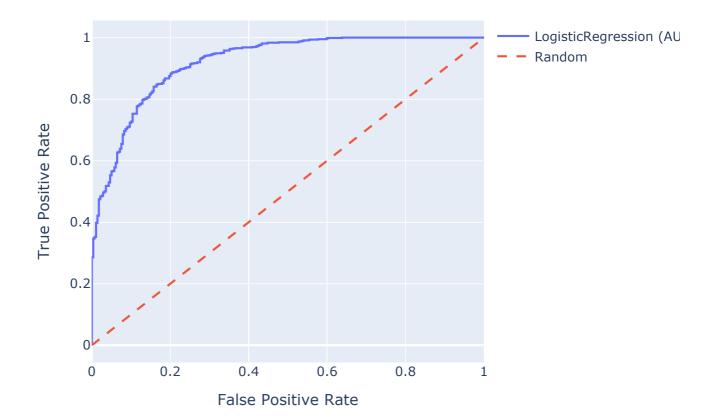
```
Target: Type_of_Respiratory_Allergy_IGE_Pollen_Tree | Model: LogisticReg  
Accuracy: 0.8156

F1 (0): 0.6510 | F1 (1): 0.8745

Precision: 0.8176 | AUC: 0.8584135098375605

Confusion Matrix:
[[185 94]
[ 35 749]]
```

#### ROC Curve - Type\_of\_Respiratory\_Allergy\_IGE\_Pollen\_Tree - Logistic



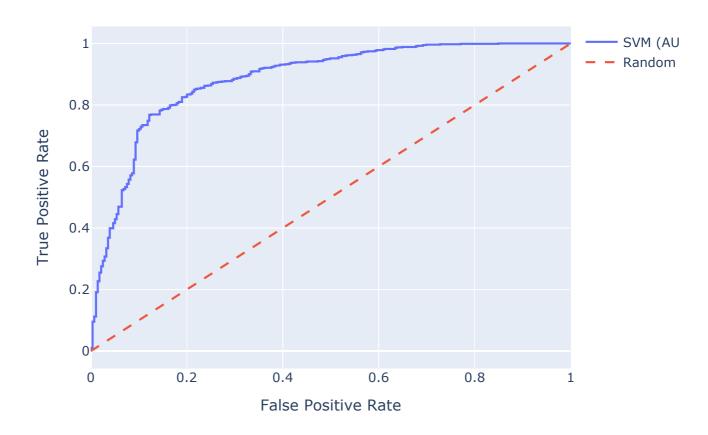
```
Target: Type_of_Respiratory_Allergy_IGE_Pollen_Tree | Model: SVM Accuracy: 0.7761

F1 (0): 0.6663 | F1 (1): 0.8310

Precision: 0.8337 | AUC: 0.8905201104568194

Confusion Matrix:
[[171 108]
[ 57 727]]
```

#### KUC Curve - Type\_or\_kespiratory\_Allergy\_IGE\_Pollen\_Tree - SVM



```
Target: Type_of_Respiratory_Allergy_IGE_Dander_Animals | Model: RandomFc Accuracy: 0.8251

F1 (0): 0.7834 | F1 (1): 0.8531

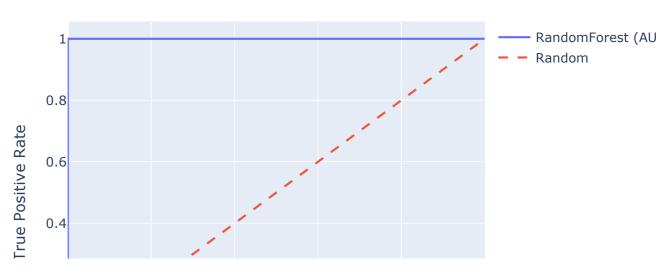
Precision: 0.8268 | AUC: 0.8851984834085689

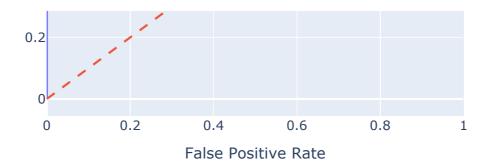
Confusion Matrix:

[[458 0]

[ 0 605]]
```

## ROC Curve - Type\_of\_Respiratory\_Allergy\_IGE\_Dander\_Animals - Ra





```
Target: Type_of_Respiratory_Allergy_IGE_Dander_Animals | Model: XGBoost Accuracy: 0.8072

F1 (0): 0.7619 | F1 (1): 0.8379

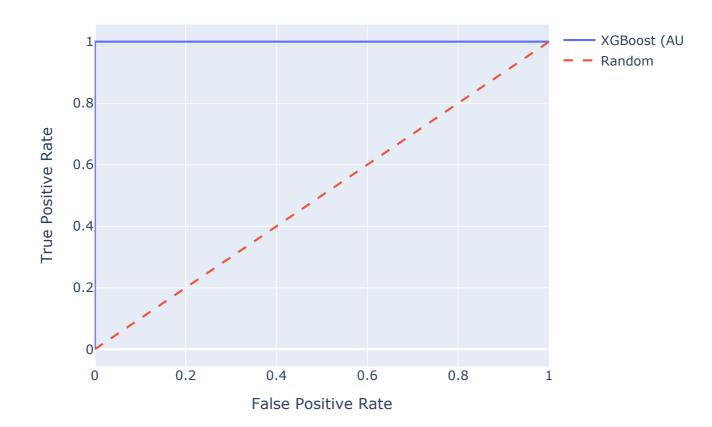
Precision: 0.8086 | AUC: 0.8649671339193791

Confusion Matrix:

[[458 0]

[ 0 605]]
```

#### ROC Curve - Type\_of\_Respiratory\_Allergy\_IGE\_Dander\_Animals - XG



```
Target: Type_of_Respiratory_Allergy_IGE_Dander_Animals | Model: Logistic Accuracy: 0.7169

F1 (0): 0.6911 | F1 (1): 0.7363

Precision: 0.7282 | AUC: 0.7952817375465272

Confusion Matrix:
```

[[334 124] [120 485]]

#### ROC Curve - Type\_of\_Respiratory\_Allergy\_IGE\_Dander\_Animals - Log



```
Target: Type_of_Respiratory_Allergy_IGE_Dander_Animals | Model: SVM Accuracy: 0.7357

F1 (0): 0.7157 | F1 (1): 0.7523

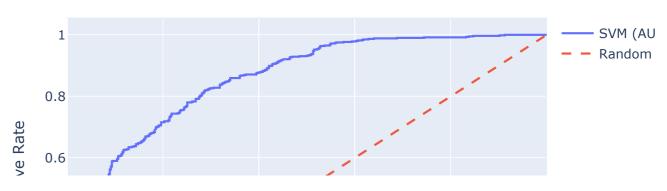
Precision: 0.7463 | AUC: 0.8254104300308862

Confusion Matrix:

[[339 119]

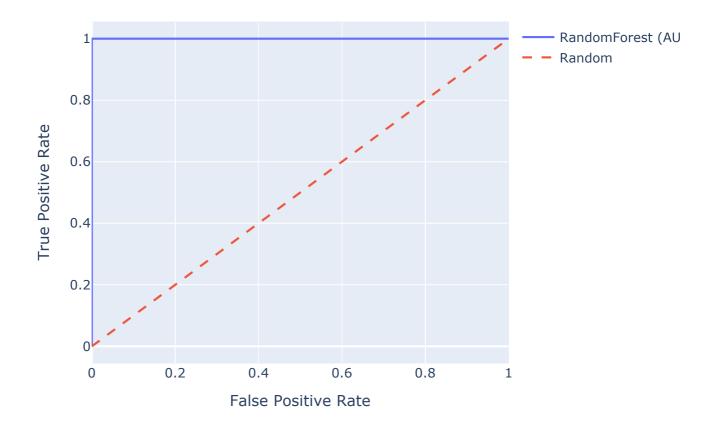
[132 473]]
```

## ROC Curve - Type\_of\_Respiratory\_Allergy\_IGE\_Dander\_Animals - SV





## ROC Curve - Type\_of\_Respiratory\_Allergy\_IGE\_Mite\_Cockroach - Rar



Target: Type of Respiratory Allergy TGE Mite Cockroach | Model: XGBoost

```
Accuracy: 0.8627

F1 (0): 0.8414 | F1 (1): 0.8786

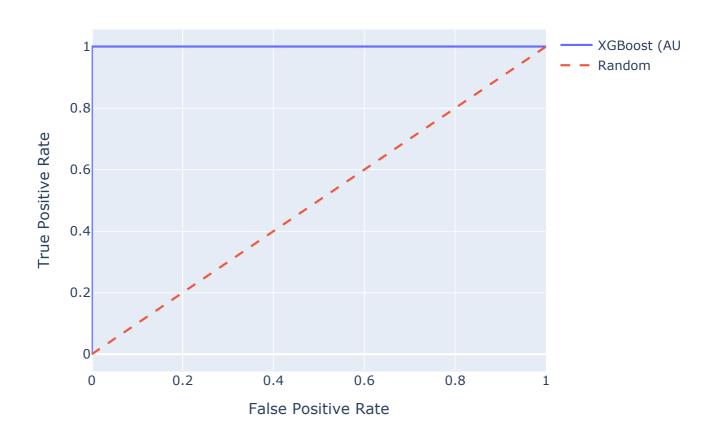
Precision: 0.8644 | AUC: 0.9362430946445277

Confusion Matrix:

[[464 0]

[ 0 599]]
```

#### ROC Curve - Type\_of\_Respiratory\_Allergy\_IGE\_Mite\_Cockroach - XGI



```
Target: Type_of_Respiratory_Allergy_IGE_Mite_Cockroach | Model: Logistic
Accuracy: 0.7977

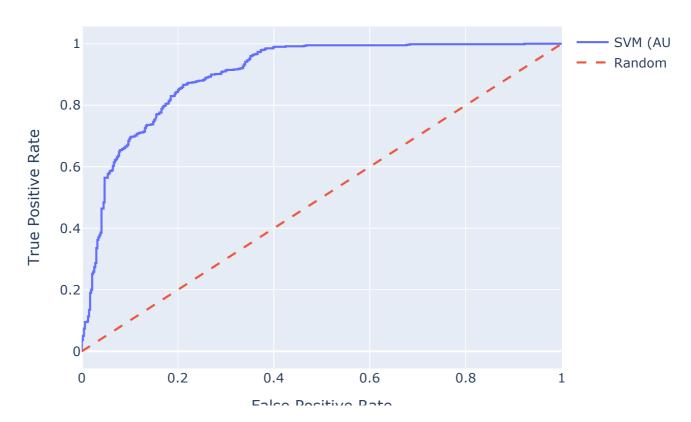
F1 (0): 0.7954 | F1 (1): 0.7988
Precision: 0.8222 | AUC: 0.8859456090562723
Confusion Matrix:
[[432 32]
[132 467]]
```

## ROC Curve - Type\_of\_Respiratory\_Allergy\_IGE\_Mite\_Cockroach - Log



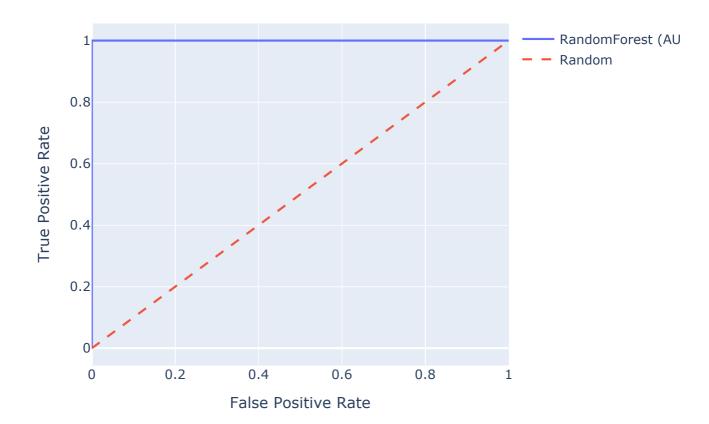


## ROC Curve - Type\_of\_Respiratory\_Allergy\_IGE\_Mite\_Cockroach - SVI

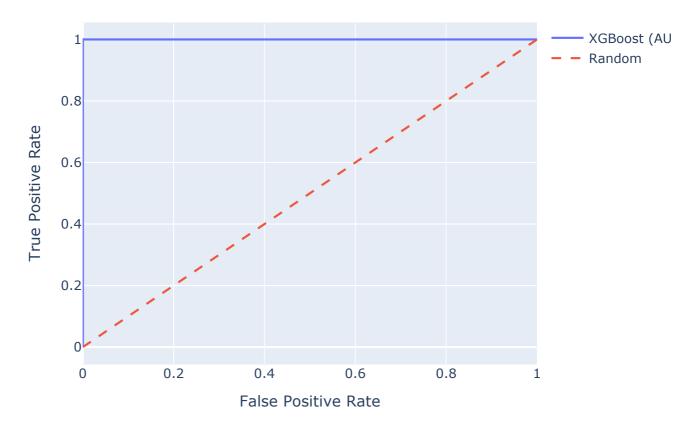


#### raise rusitive rate

#### ROC Curve - Type\_of\_Respiratory\_Allergy\_IGE\_Molds\_Yeast - Randor



ROC Curve - Type\_of\_Respiratory\_Allergy\_IGE\_Molds\_Yeast - XGBoo



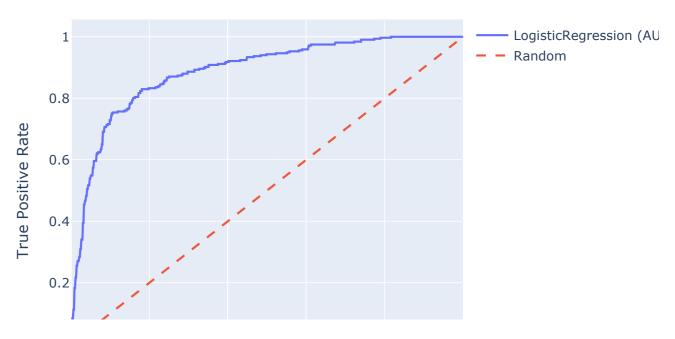
```
Target: Type_of_Respiratory_Allergy_IGE_Molds_Yeast | Model: LogisticReg Accuracy: 0.7996

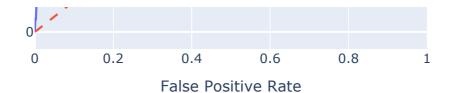
F1 (0): 0.8602 | F1 (1): 0.6443

Precision: 0.7962 | AUC: 0.8110391601278698

Confusion Matrix:
[[711 35]
[149 168]]
```

#### ROC Curve - Type\_of\_Respiratory\_Allergy\_IGE\_Molds\_Yeast - Logistic





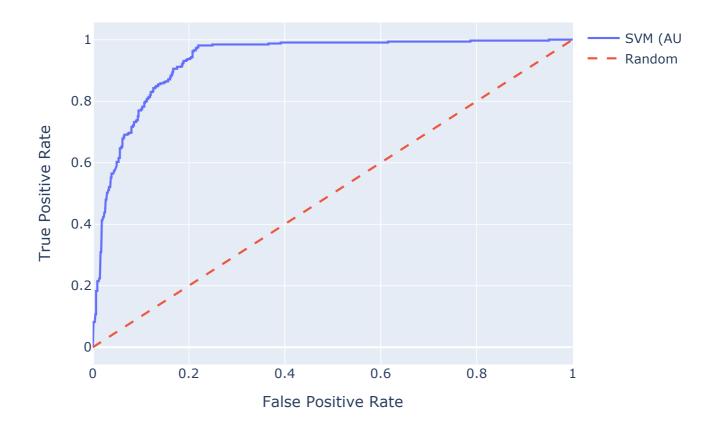
```
Target: Type_of_Respiratory_Allergy_IGE_Molds_Yeast | Model: SVM Accuracy: 0.8230

F1 (0): 0.8734 | F1 (1): 0.7037

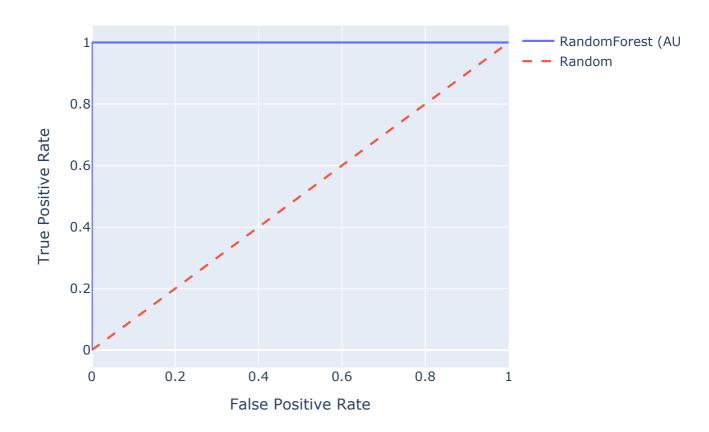
Precision: 0.8259 | AUC: 0.8923271941296136

Confusion Matrix:
[[718 28]
[144 173]]
```

### ROC Curve - Type\_of\_Respiratory\_Allergy\_IGE\_Molds\_Yeast - SVM



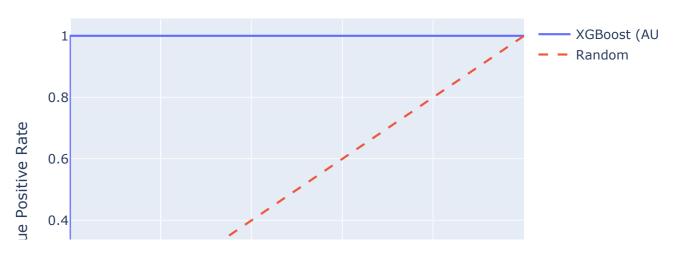
#### ROC Curve - Type\_of\_Respiratory\_Allergy\_ARIA - RandomForest

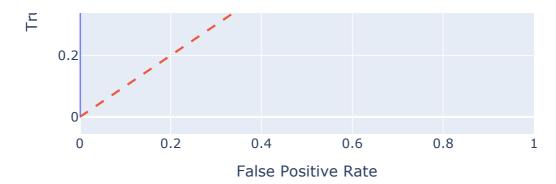


```
Target: Type_of_Respiratory_Allergy_ARIA | Model: XGBoost Accuracy: 0.9925

F1 (0): 0.9928 | F1 (1): 0.9921
Precision: 0.9926 | AUC: 0.9990056022408964
Confusion Matrix:
[[557 0]
[ 0 506]]
```

# ROC Curve - Type\_of\_Respiratory\_Allergy\_ARIA - XGBoost





```
Target: Type_of_Respiratory_Allergy_ARIA | Model: LogisticRegression Accuracy: 0.9896

F1 (0): 0.9902 | F1 (1): 0.9891

Precision: 0.9898 | AUC: 0.9996114081996434

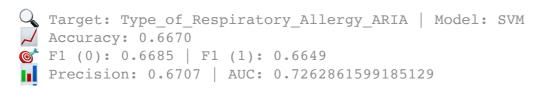
Confusion Matrix:

[[557 0]

[ 4 502]]
```

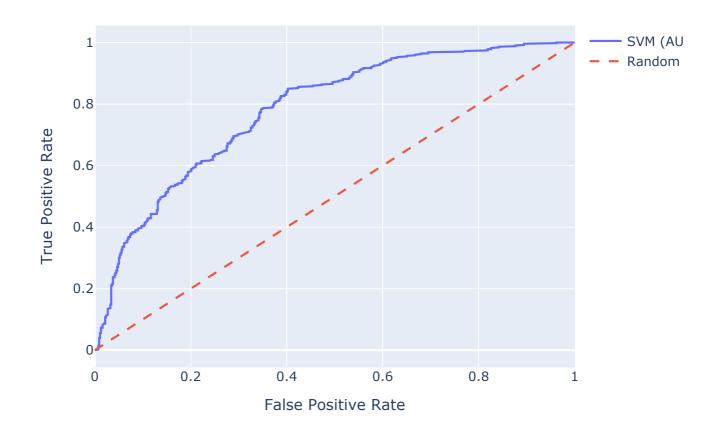
# ROC Curve - Type\_of\_Respiratory\_Allergy\_ARIA - LogisticRegression





Confusion Matrix: [[380 177] [148 358]]

# ROC Curve - Type\_of\_Respiratory\_Allergy\_ARIA - SVM



```
Target: Type_of_Respiratory_Allergy_CONJ | Model: RandomForest

Accuracy: 0.9915

F1 (0): 0.9943 | F1 (1): 0.9836

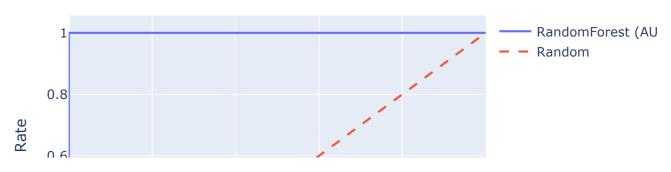
Precision: 0.9916 | AUC: 0.9996567544035898

Confusion Matrix:

[[787 0]

[ 0 276]]
```

# $ROC\ Curve\ -\ Type\_of\_Respiratory\_Allergy\_CONJ\ -\ RandomForest$





```
Target: Type_of_Respiratory_Allergy_CONJ | Model: XGBoost Accuracy: 0.9991

F1 (0): 0.9994 | F1 (1): 0.9982

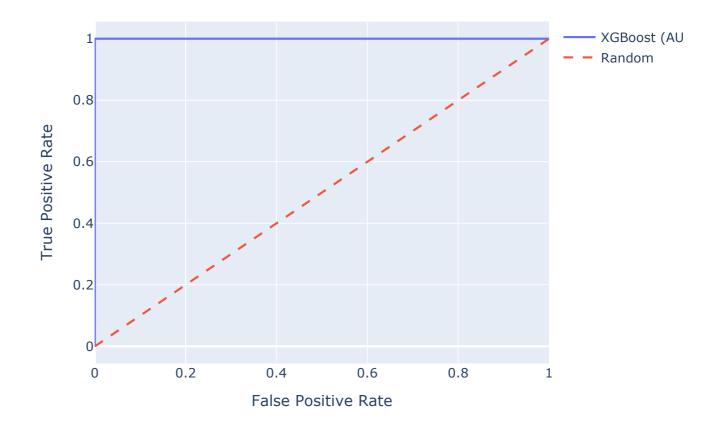
Precision: 0.9991 | AUC: 0.9996718237224567

Confusion Matrix:

[[787 0]

[ 0 276]]
```

### ROC Curve - Type\_of\_Respiratory\_Allergy\_CONJ - XGBoost



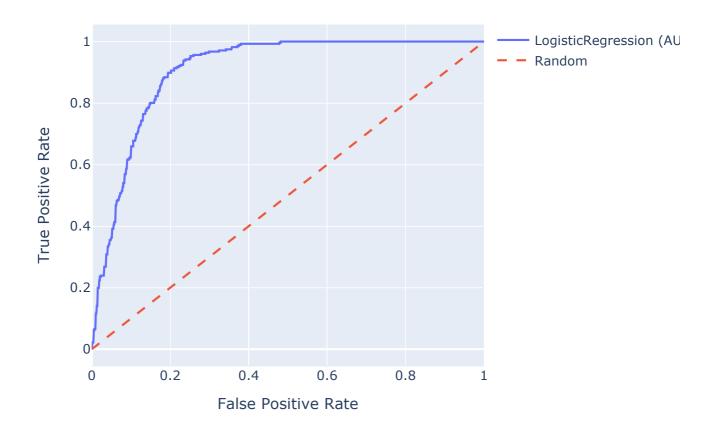
```
Target: Type_of_Respiratory_Allergy_CONJ | Model: LogisticRegression Accuracy: 0.7912

F1 (0): 0.8541 | F1 (1): 0.6295

Precision: 0.8060 | AUC: 0.848039571773749

Confusion Matrix:
[[700 87]
[ 88 188]]
```

### ROC Curve - Type\_of\_Respiratory\_Allergy\_CONJ - LogisticRegression



```
Target: Type_of_Respiratory_Allergy_CONJ | Model: SVM Accuracy: 0.6877

F1 (0): 0.7453 | F1 (1): 0.5949

Precision: 0.8104 | AUC: 0.7328171125639481

Confusion Matrix:

[[785 2]
[262 14]]
```

# ROC Curve - Type\_of\_Respiratory\_Allergy\_CONJ - SVM



```
0.8
0.6
0.4
0.2
0.2
0.4
0.6
0.8
1
False Positive Rate
```

```
Target: Type_of_Respiratory_Allergy_IGE_Pollen_Gram | Model: RandomFores Accuracy: 0.8279

F1 (0): 0.7549 | F1 (1): 0.8672

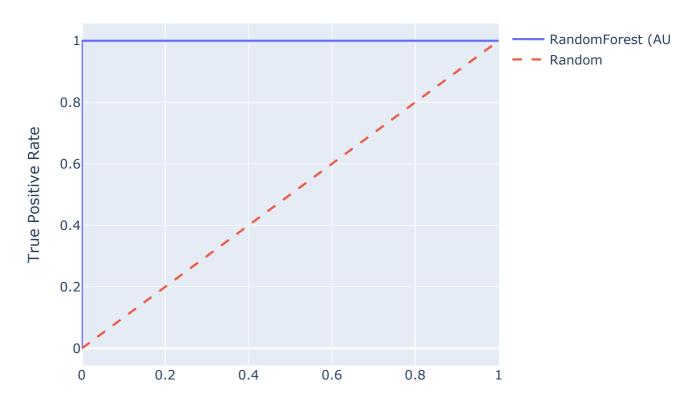
Precision: 0.8277 | AUC: 0.9112214398407023

Confusion Matrix:

[[379 0]

[ 0 684]]
```

# ROC Curve - Type\_of\_Respiratory\_Allergy\_IGE\_Pollen\_Gram - Rando



#### False Positive Rate

```
Target: Type_of_Respiratory_Allergy_IGE_Pollen_Gram | Model: XGBoost Accuracy: 0.8513

F1 (0): 0.7940 | F1 (1): 0.8835

Precision: 0.8530 | AUC: 0.9335121280251851

Confusion Matrix:
[[379 0]
   0 684]]
```

### ROC Curve - Type\_of\_Respiratory\_Allergy\_IGE\_Pollen\_Gram - XGBoo



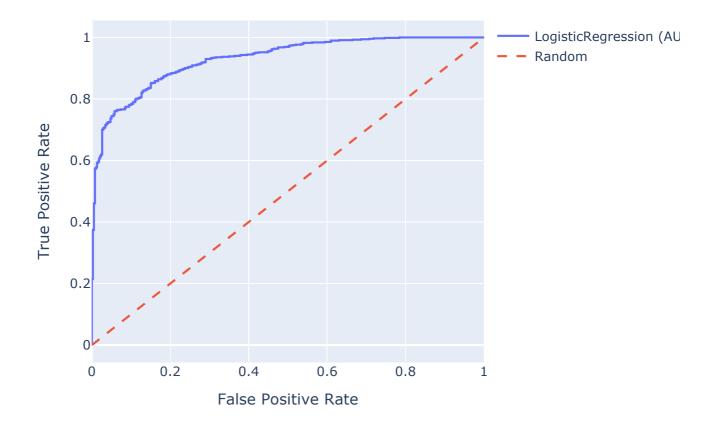
```
Target: Type_of_Respiratory_Allergy_IGE_Pollen_Gram | Model: LogisticReg  
Accuracy: 0.7921

F1 (0): 0.7342 | F1 (1): 0.8292

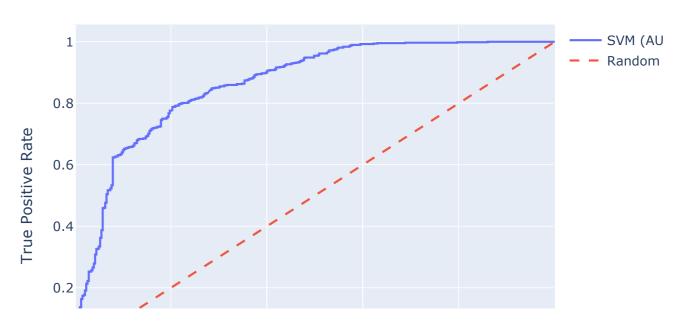
Precision: 0.8063 | AUC: 0.8787284642968449

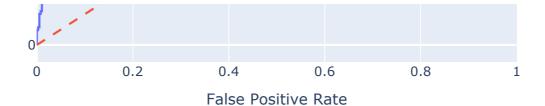
Confusion Matrix:
[[315 64]
[ 94 590]]
```

ROC Curve - Type\_of\_Respiratory\_Allergy\_IGE\_Pollen\_Gram - Logisti



ROC Curve - Type\_of\_Respiratory\_Allergy\_IGE\_Pollen\_Gram - SVM





```
Target: Type_of_Respiratory_Allergy_GINA | Model: RandomForest

Accuracy: 0.9906

F1 (0): 0.9905 | F1 (1): 0.9907

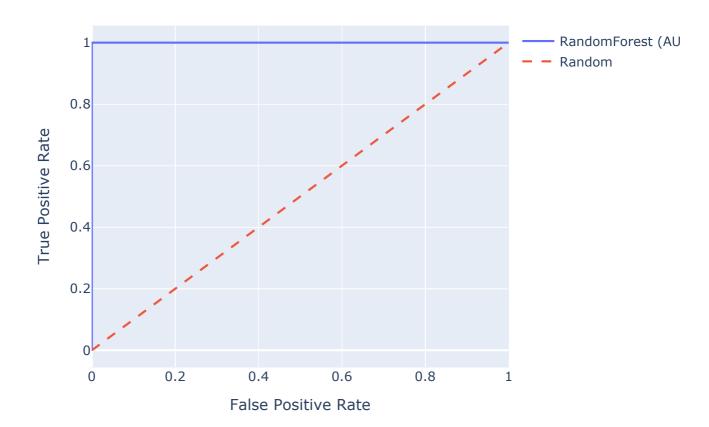
Precision: 0.9907 | AUC: 0.9991129241129242

Confusion Matrix:

[[520 0]

[ 0 543]]
```

#### ROC Curve - Type\_of\_Respiratory\_Allergy\_GINA - RandomForest



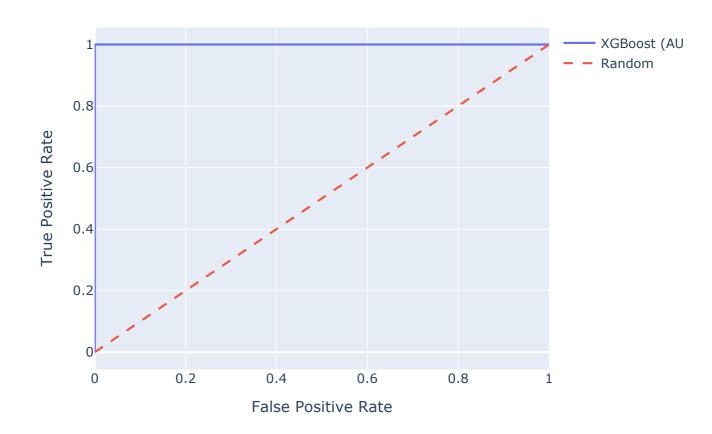
```
Target: Type_of_Respiratory_Allergy_GINA | Model: XGBoost Accuracy: 0.9887

F1 (0): 0.9887 | F1 (1): 0.9887

Precision: 0.9890 | AUC: 0.9993945868945868

Confusion Matrix:
[[520 0]
[ 0 543]]
```

#### ROC Curve - Type\_of\_Respiratory\_Allergy\_GINA - XGBoost



```
Target: Type_of_Respiratory_Allergy_GINA | Model: LogisticRegression
Accuracy: 0.9877

F1 (0): 0.9877 | F1 (1): 0.9878

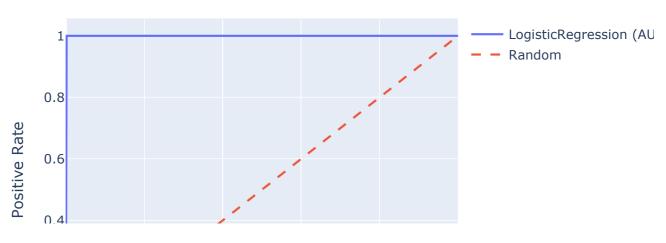
Precision: 0.9881 | AUC: 0.9996082621082621

Confusion Matrix:

[[519 1]

[ 4 539]]
```

# ROC Curve - Type\_of\_Respiratory\_Allergy\_GINA - LogisticRegression





```
Target: Type_of_Respiratory_Allergy_GINA | Model: SVM Accuracy: 0.6697

F1 (0): 0.6762 | F1 (1): 0.6614

Precision: 0.6733 | AUC: 0.7325848225848225

Confusion Matrix:
[[387 133]
[173 370]]
```

# ROC Curve - Type\_of\_Respiratory\_Allergy\_GINA - SVM



import pandas as pd
import numpy as np

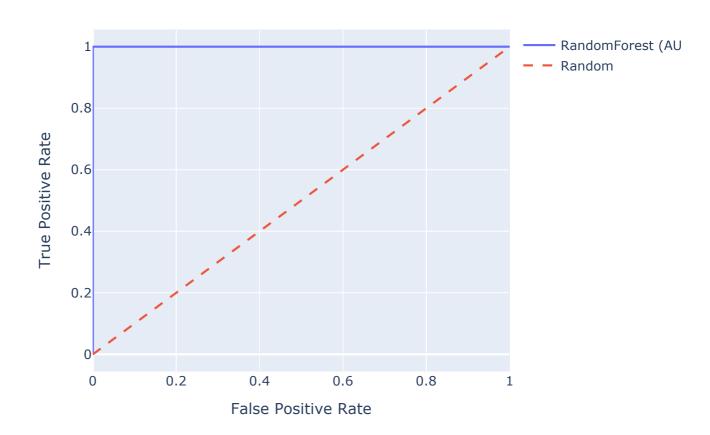
```
from sklearn.model selection import StratifiedKFold
from sklearn.ensemble import RandomForestClassifier
from sklearn.linear model import LogisticRegression
from sklearn.svm import SVC
from xgboost import XGBClassifier
from sklearn.metrics import (
    f1_score, accuracy_score, recall_score,
    precision_score, confusion_matrix, roc_auc_score, roc_curve
)
from imblearn.over sampling import SMOTE
import plotly.graph_objects as go
V1 food = V1[V1["Food Allergy"] == 1]
targets = ["Type_of_Food_Allergy_Aromatics",
    "Type_of_Food_Allergy_Cereals_&_Seeds",
    "Type_of_Food_Allergy_Egg",
    "Type of Food Allergy Fish",
    "Type_of_Food_Allergy_Fruits_and_Vegetables",
    "Type_of_Food_Allergy_Mammalian_Milk",
    "Type_of_Food_Allergy_Oral_Syndrom",
    "Type_of_Food_Allergy_Other_Legumes",
    "Type_of_Food_Allergy_Peanut",
    "Type_of_Food_Allergy_Shellfish",
    "Type of Food Allergy TPO",
    "Type_of_Food_Allergy_Tree_Nuts"]
models = {
    "RandomForest": RandomForestClassifier(random_state=42),
    "XGBoost": XGBClassifier(random_state=42, eval_metric="logloss", use_label_
    "LogisticRegression": LogisticRegression(max iter=1000, random state=42),
    "SVM": SVC(probability=True, random_state=42)
}
X=V1 food.copy()
X.drop(target_1, axis=1, inplace=True)
X.drop(extra_columns, axis=1, inplace=True)
X.drop(extra, axis=1, inplace=True)
X.drop(inconnu, axis=1, inplace=True)
X = X.iloc[:, 1:]
results food = []
kfold = StratifiedKFold(n_splits=10, shuffle=True, random_state=42)
for target in targets:
    y = V1_food[target]
    for model name, base model in models.items():
        f1_class0_scores, f1_class1_scores = [], []
```

```
precision_scores, acc_scores, recall_scores, auc_scores = [], [], [], [
for train_idx, test_idx in kfold.split(X, y):
    X_train, X_test = X.iloc[train_idx], X.iloc[test_idx]
    y_train, y_test = y.iloc[train_idx], y.iloc[test_idx]
    smote = SMOTE(random_state=42)
    X_train_res, y_train_res = smote.fit_resample(X_train, y_train)
    base_model.fit(X_train_res, y_train_res)
    y_pred = base_model.predict(X_test)
    acc_scores.append(accuracy_score(y_test, y_pred))
    recall_scores.append(recall_score(y_test, y_pred, zero_division=0))
    precision_scores.append(precision_score(y_test, y_pred, average='we
    f1_class0_scores.append(f1_score(y_test, y_pred, pos_label=0, zero_
    f1_class1_scores.append(f1_score(y_test, y_pred, pos_label=1, zero_
    if hasattr(base_model, "predict_proba"):
        y_proba = base_model.predict_proba(X_test)[:, 1]
        auc_scores.append(roc_auc_score(y_test, y_proba))
base_model.fit(X, y)
y pred full = base model.predict(X)
y_proba_full = base_model.predict_proba(X)[:, 1] if hasattr(base_model,
matrix = confusion_matrix(y, y_pred_full)
print(f"\nQ Target: {target} | Model: {model_name}")
print(f" Accuracy: {np.mean(acc_scores):.4f}")
print(f''@ F1 (0): {np.mean(f1_class0_scores):.4f} | F1 (1): {np.mean(
print(f" Precision: {np.mean(precision_scores):.4f} | AUC: {np.mean(≀
print(" Confusion Matrix:\n", matrix)
if y proba full is not None:
    fpr, tpr, _ = roc_curve(y, y_proba_full)
    fig = go.Figure()
    fig.add_trace(go.Scatter(x=fpr, y=tpr, mode='lines', name=f"{model_
    fig.add_trace(go.Scatter(x=[0, 1], y=[0, 1], mode='lines', name='Ra
    fig.update_layout(
        title=f"ROC Curve - {target} - {model_name}",
        xaxis_title="False Positive Rate",
        yaxis_title="True Positive Rate",
       width=700, height=500
    fig.show()
results food.append({
    "Target": target,
```

0 16]]

```
"Model": model_name,
            "F1_Class_0": np.mean(f1_class0_scores),
            "F1_Class_1": np.mean(f1_class1_scores),
            "Precision": np.mean(precision_scores),
            "Accuracy": np.mean(acc_scores),
            "Recall": np.mean(recall scores),
            "AUC_ROC": np.mean(auc_scores) if auc_scores else np.nan
        })
pd.DataFrame(results_food).to_csv("results_V1_food.csv", index=False)
→
       Target: Type of Food Allergy_Aromatics | Model: RandomForest
       Accuracy: 0.9776
      F1 (0): 0.9887 | F1 (1): 0.0667
       Precision: 0.9662 | AUC: 0.8009273772204806
    Confusion Matrix:
     [[878 0]
```

# ROC Curve - Type\_of\_Food\_Allergy\_Aromatics - RandomForest

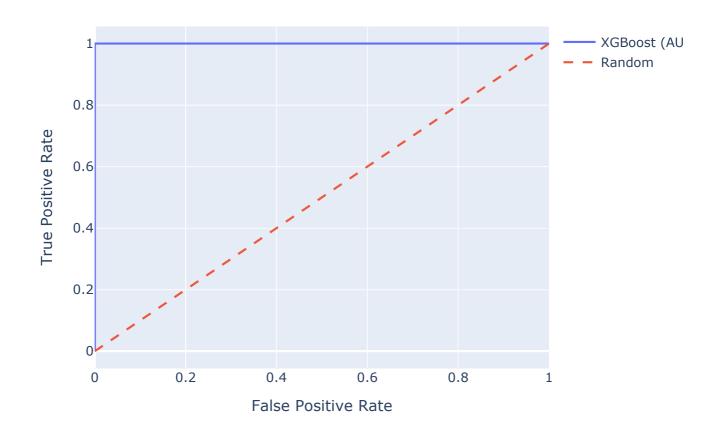


```
Target: Type_of_Food_Allergy_Aromatics | Model: XGBoost Accuracy: 0.9754

F1 (0): 0.9875 | F1 (1): 0.1900
```

```
Precision: 0.9727 | AUC: 0.7526907001044931
Confusion Matrix:
[[878 0]
0 16]]
```

### ROC Curve - Type\_of\_Food\_Allergy\_Aromatics - XGBoost



```
Target: Type_of_Food_Allergy_Aromatics | Model: LogisticRegression Accuracy: 0.9351

F1 (0): 0.9662 | F1 (1): 0.0832

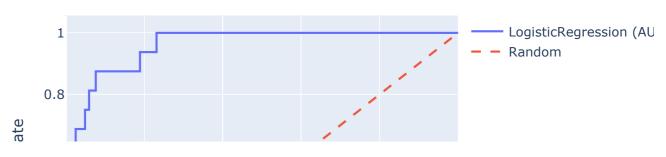
Precision: 0.9684 | AUC: 0.7120885579937305

Confusion Matrix:

[[878   0]

[ 15   1]]
```

### ROC Curve - Type\_of\_Food\_Allergy\_Aromatics - LogisticRegression





```
Target: Type_of_Food_Allergy_Aromatics | Model: SVM Accuracy: 0.7998

F1 (0): 0.8872 | F1 (1): 0.0868

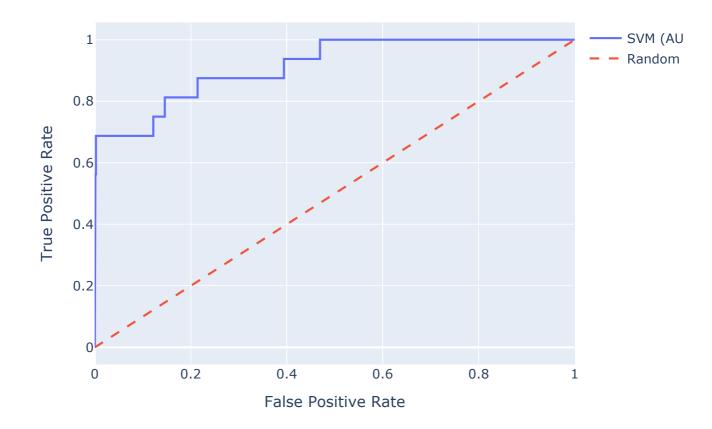
Precision: 0.9737 | AUC: 0.7221068443051202

Confusion Matrix:

[[878 0]

[ 16 0]]
```

# ROC Curve - Type\_of\_Food\_Allergy\_Aromatics - SVM



```
Target: Type_of_Food_Allergy_Cereals_&_Seeds | Model: RandomForest Accuracy: 0.9608

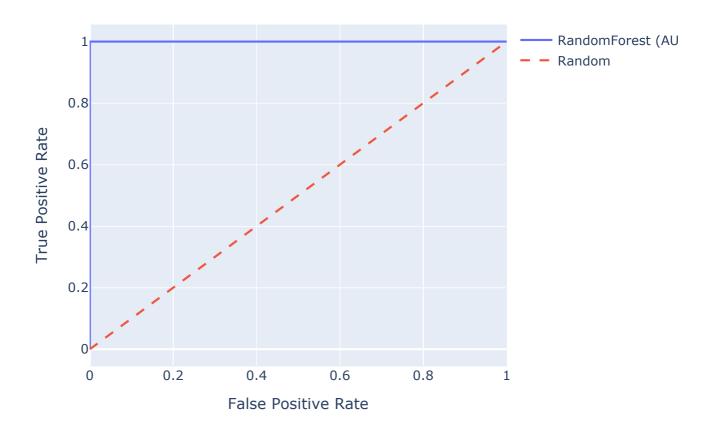
F1 (0): 0.9800 | F1 (1): 0.0000

Precision: 0.9403 | AUC: 0.7197986278178738

Confusion Matrix:

[[867 0]
[ 0 27]]
```

### ROC Curve - Type\_of\_Food\_Allergy\_Cereals\_&\_Seeds - RandomFores



```
Target: Type_of_Food_Allergy_Cereals_&_Seeds | Model: XGBoost  
Accuracy: 0.9507

F1 (0): 0.9747 | F1 (1): 0.0000

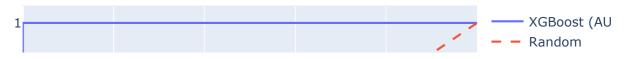
Precision: 0.9400 | AUC: 0.6871380201372181

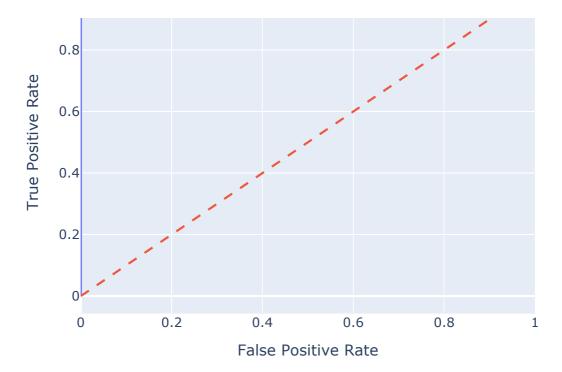
Confusion Matrix:

[[867 0]

0 27]]
```

## ROC Curve - Type\_of\_Food\_Allergy\_Cereals\_&\_Seeds - XGBoost





```
Target: Type_of_Food_Allergy_Cereals_&_Seeds | Model: LogisticRegression Accuracy: 0.9038

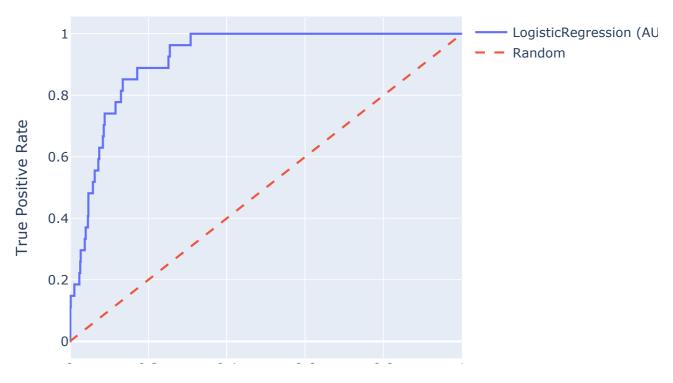
F1 (0): 0.9491 | F1 (1): 0.0832

Precision: 0.9447 | AUC: 0.5341441682259646

Confusion Matrix:

[[867 0]
[27 0]]
```

# ROC Curve - Type\_of\_Food\_Allergy\_Cereals\_&\_Seeds - LogisticRegre



0 0.2 0.4 0.6 0.8 1

False Positive Rate

```
Target: Type_of_Food_Allergy_Cereals_&_Seeds | Model: SVM Accuracy: 0.7605

F1 (0): 0.8612 | F1 (1): 0.1009

Precision: 0.9522 | AUC: 0.5535685645549318

Confusion Matrix:
[[867 0]
[27 0]]
```

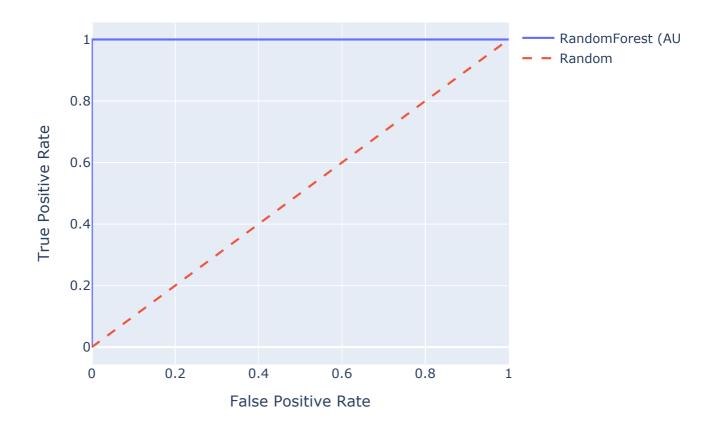
### ROC Curve - Type\_of\_Food\_Allergy\_Cereals\_&\_Seeds - SVM



```
Target: Type_of_Food_Allergy_Egg | Model: RandomForest
Accuracy: 0.9844

F1 (0): 0.9921 | F1 (1): 0.0667
Precision: 0.9707 | AUC: 0.90227272727271
Confusion Matrix:
[[880 0]
[ 0 14]]
```

ROC Curve - Type\_of\_Food\_Allergy\_Egg - RandomForest



```
Target: Type_of_Food_Allergy_Egg | Model: XGBoost Accuracy: 0.9833

F1 (0): 0.9915 | F1 (1): 0.3500

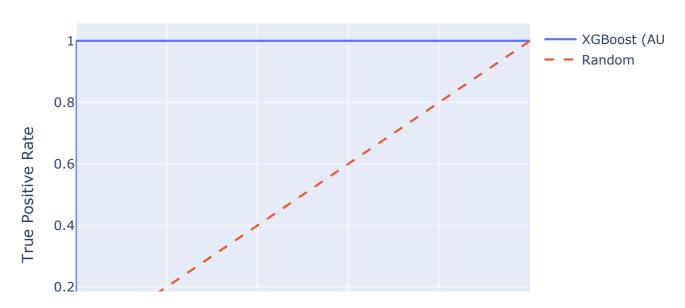
Precision: 0.9800 | AUC: 0.9335227272727273

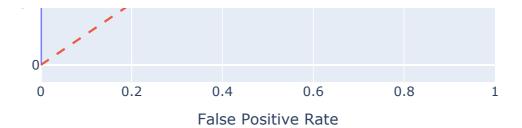
Confusion Matrix:

[[880 0]

[ 0 14]]
```

# ROC Curve - Type\_of\_Food\_Allergy\_Egg - XGBoost





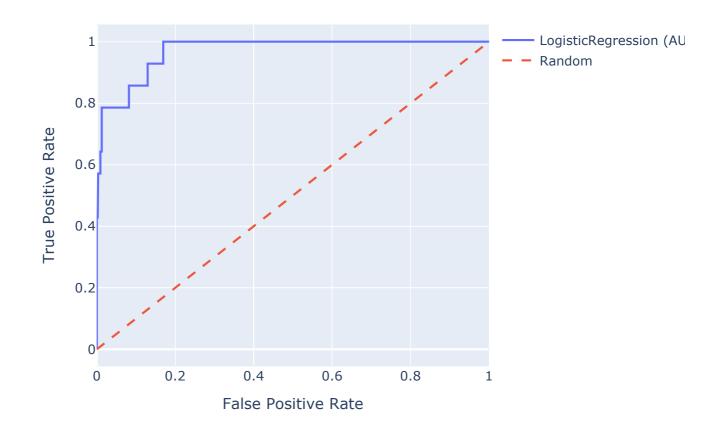
```
Target: Type_of_Food_Allergy_Egg | Model: LogisticRegression Accuracy: 0.9519

F1 (0): 0.9752 | F1 (1): 0.0933

Precision: 0.9726 | AUC: 0.77727272727273

Confusion Matrix:
[[880 0]
[ 12 2]]
```

# ROC Curve - Type\_of\_Food\_Allergy\_Egg - LogisticRegression



```
Target: Type_of_Food_Allergy_Egg | Model: SVM
Accuracy: 0.8222

F1 (0): 0.9013 | F1 (1): 0.0520
Precision: 0.9716 | AUC: 0.6448863636363636
Confusion Matrix:
[[880 0]
[14 0]]
```

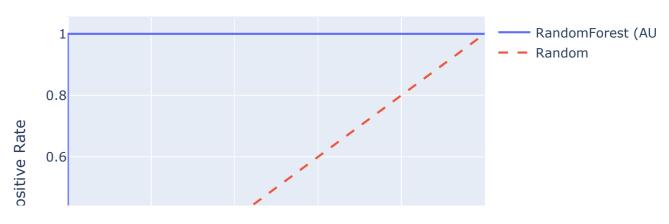
#### ROC Curve - Type\_of\_Food\_Allergy\_Egg - SVM



```
Target: Type_of_Food_Allergy_Fish | Model: RandomForest
Accuracy: 0.9731

F1 (0): 0.9864 | F1 (1): 0.0000
Precision: 0.9557 | AUC: 0.6779127481713689
Confusion Matrix:
[[874 0]
[ 0 20]]
```

# ROC Curve - Type\_of\_Food\_Allergy\_Fish - RandomForest





```
Target: Type_of_Food_Allergy_Fish | Model: XGBoost Accuracy: 0.9698

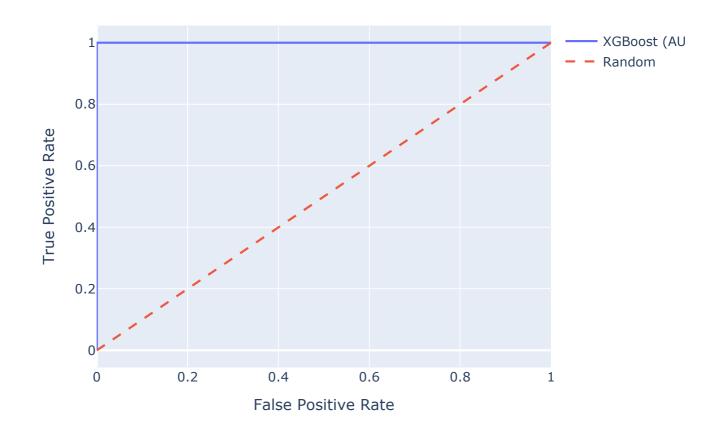
F1 (0): 0.9846 | F1 (1): 0.0667

Precision: 0.9589 | AUC: 0.6398968129571577

Confusion Matrix:

[[874 0]
[ 0 20]]
```

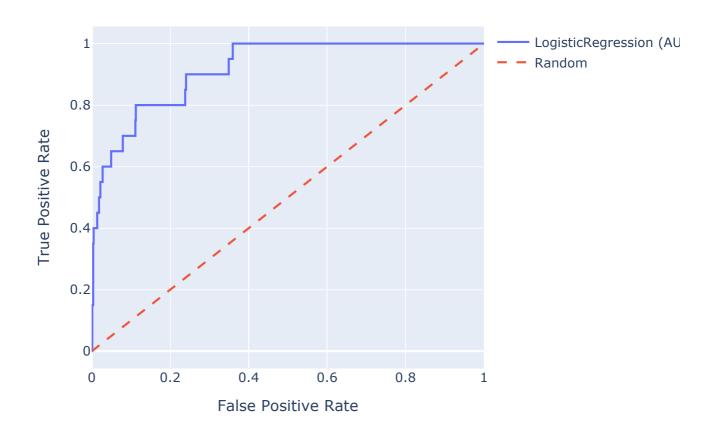
#### ROC Curve - Type\_of\_Food\_Allergy\_Fish - XGBoost



Target: Type\_of\_Food\_Allergy\_Fish | Model: LogisticRegression Accuracy: 0.9239

```
F1 (U): U.96U1 | F1 (1): U.U66/
Precision: 0.9589 | AUC: 0.550208986415883
Confusion Matrix:
[[874 0]
[ 20 0]]
```

# ROC Curve - Type\_of\_Food\_Allergy\_Fish - LogisticRegression



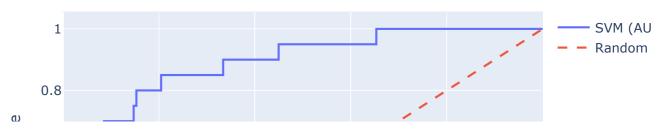
```
Target: Type_of_Food_Allergy_Fish | Model: SVM Accuracy: 0.7616

F1 (0): 0.8624 | F1 (1): 0.0838

Precision: 0.9626 | AUC: 0.6703565830721003

Confusion Matrix:
[[874 0]
[ 20 0]]
```

# ROC Curve - Type\_of\_Food\_Allergy\_Fish - SVM





```
Target: Type_of_Food_Allergy_Fruits_and_Vegetables | Model: RandomForest Accuracy: 0.9508

F1 (0): 0.9747 | F1 (1): 0.0733

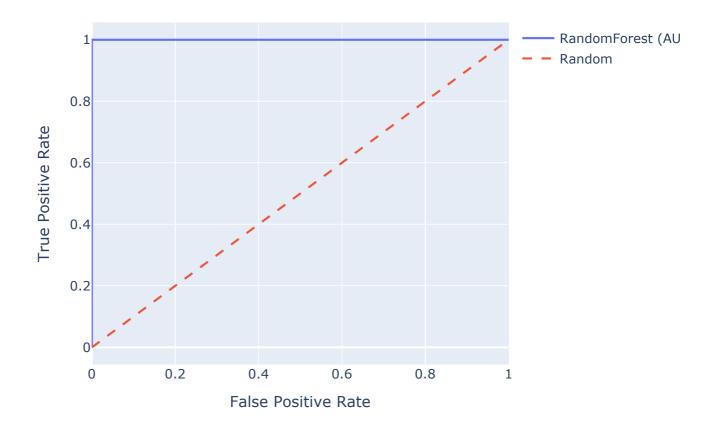
Precision: 0.9278 | AUC: 0.8438275193798448

Confusion Matrix:

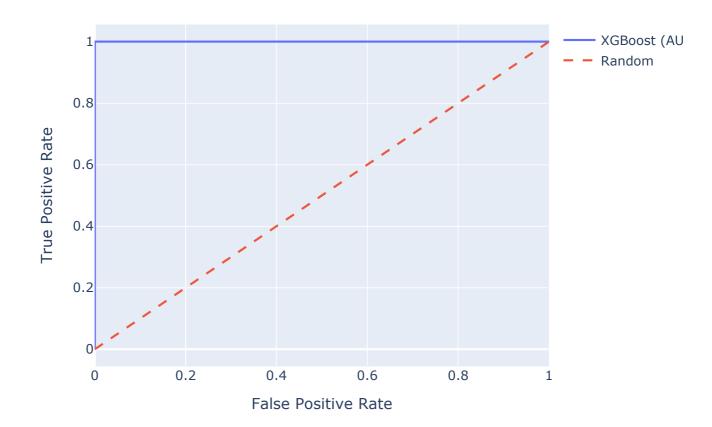
[[859 0]

[ 0 35]]
```

# ROC Curve - Type\_of\_Food\_Allergy\_Fruits\_and\_Vegetables - Random



### ROC Curve - Type\_of\_Food\_Allergy\_Fruits\_and\_Vegetables - XGBoos



```
Target: Type_of_Food_Allergy_Fruits_and_Vegetables | Model: LogisticRegr Accuracy: 0.9162

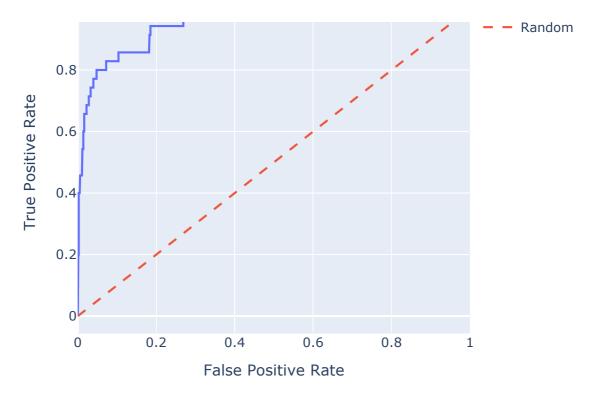
F1 (0): 0.9554 | F1 (1): 0.2003

Precision: 0.9395 | AUC: 0.7425227998176015

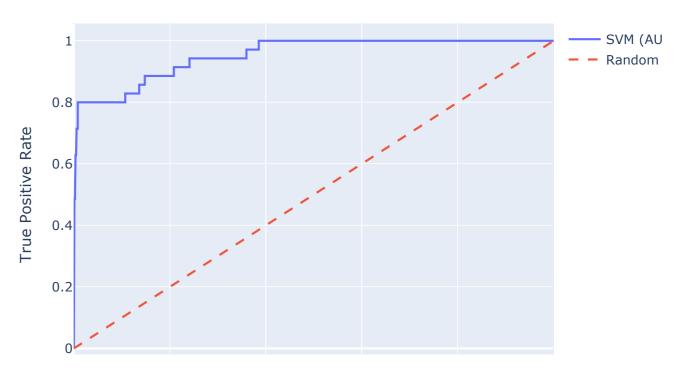
Confusion Matrix:
[[858 1]
[ 28 7]]
```

# ROC Curve - Type\_of\_Food\_Allergy\_Fruits\_and\_Vegetables - LogisticI





ROC Curve - Type\_of\_Food\_Allergy\_Fruits\_and\_Vegetables - SVM





```
Target: Type_of_Food_Allergy_Mammalian_Milk | Model: RandomForest
Accuracy: 0.9899

F1 (0): 0.9949 | F1 (1): 0.0000
Precision: 0.9800 | AUC: nan
Confusion Matrix:
[[885 0]
[ 0 9]]
```

### ROC Curve - Type\_of\_Food\_Allergy\_Mammalian\_Milk - RandomForest

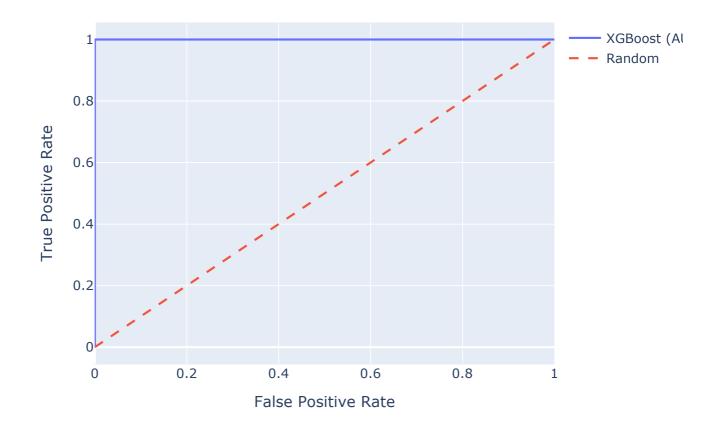


```
Target: Type_of_Food_Allergy_Mammalian_Milk | Model: XGBoost
Accuracy: 0.9855

F1 (0): 0.9927 | F1 (1): 0.0000
Precision: 0.9799 | AUC: nan
Confusion Matrix:

[[885 0]
[ 0 9]]
```

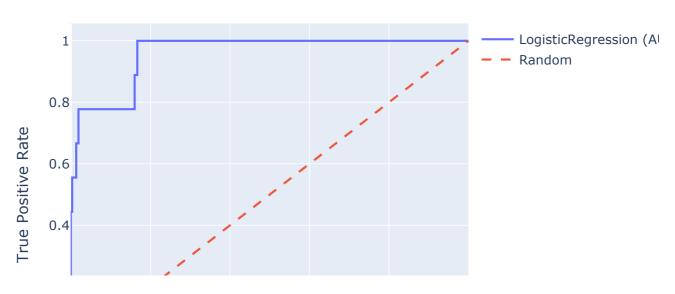
ROC Curve - Type\_of\_Food\_Allergy\_Mammalian\_Milk - XGBoost

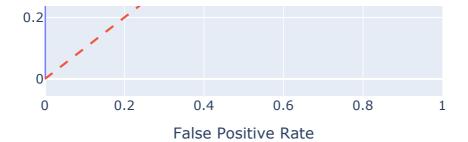


```
Target: Type_of_Food_Allergy_Mammalian_Milk | Model: LogisticRegression Accuracy: 0.9697

F1 (0): 0.9846 | F1 (1): 0.0000
Precision: 0.9798 | AUC: nan
Confusion Matrix:
[[885 0]
[ 9 0]]
```

# ROC Curve - Type\_of\_Food\_Allergy\_Mammalian\_Milk - LogisticRegres





```
Target: Type_of_Food_Allergy_Mammalian_Milk | Model: SVM Accuracy: 0.8232

F1 (0): 0.9022 | F1 (1): 0.0000

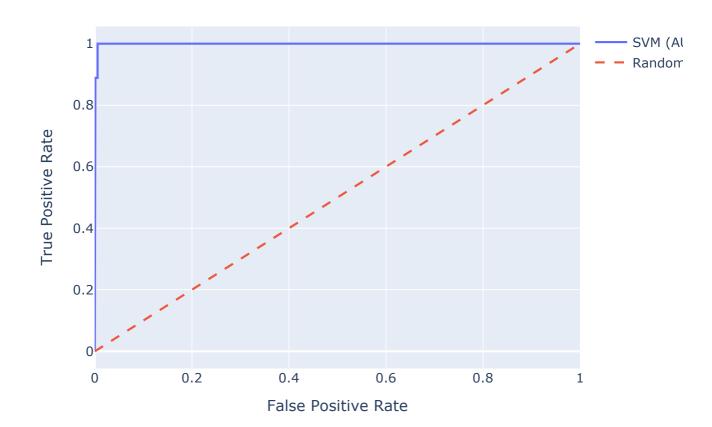
Precision: 0.9780 | AUC: nan

Confusion Matrix:

[[885 0]

[ 9 0]]
```

## ROC Curve - Type\_of\_Food\_Allergy\_Mammalian\_Milk - SVM



```
Target: Type_of_Food_Allergy_Oral_Syndrom | Model: RandomForest
Accuracy: 0.9732

F1 (0): 0.9849 | F1 (1): 0.8815

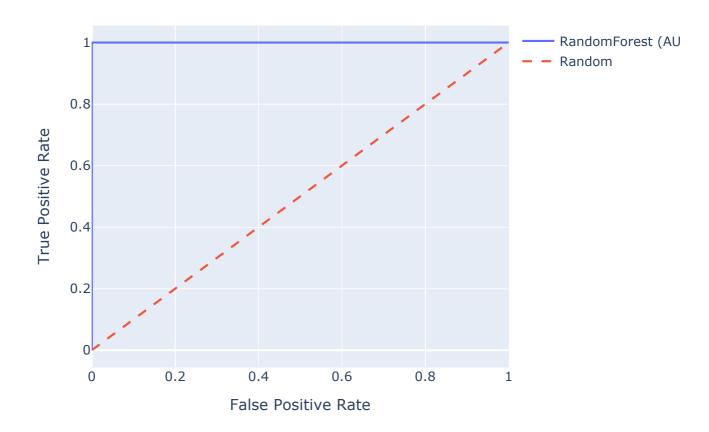
Precision: 0.9742 | AUC: 0.9989073426573427

Confusion Matrix:

[[777 0]
```

[ 0 117]]

### ROC Curve - Type\_of\_Food\_Allergy\_Oral\_Syndrom - RandomForest



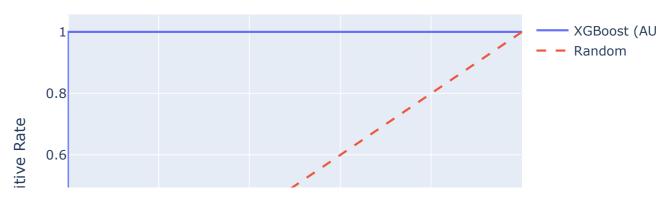
```
Target: Type_of_Food_Allergy_Oral_Syndrom | Model: XGBoost Accuracy: 0.9989

F1 (0): 0.9994 | F1 (1): 0.9960

Precision: 0.9990 | AUC: 0.999465811965812

Confusion Matrix:
[[777 0]
[ 0 117]]
```

# ROC Curve - Type\_of\_Food\_Allergy\_Oral\_Syndrom - XGBoost

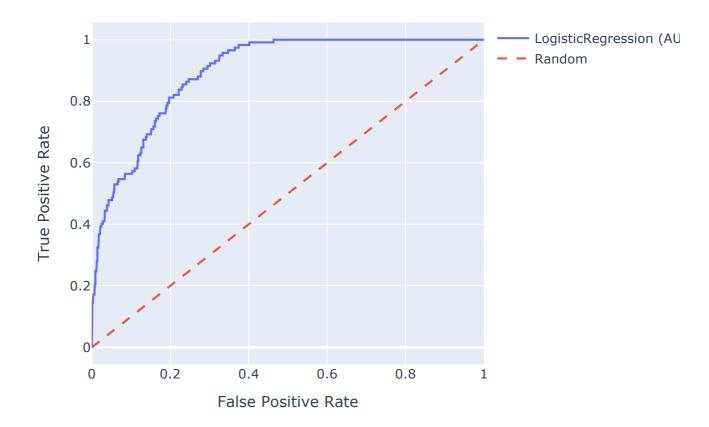




```
Target: Type_of_Food_Allergy_Oral_Syndrom | Model: LogisticRegression  
Accuracy: 0.8379

F1 (0): 0.9058 | F1 (1): 0.4118
Precision: 0.8469 | AUC: 0.8055735930735931
Confusion Matrix:
[[764 13]
[77 40]]
```

# ROC Curve - Type\_of\_Food\_Allergy\_Oral\_Syndrom - LogisticRegressic



Target: Type\_of\_Food\_Allergy\_Oral\_Syndrom | Model: SVM

```
F1 (0): 0.7741 | F1 (1): 0.3789

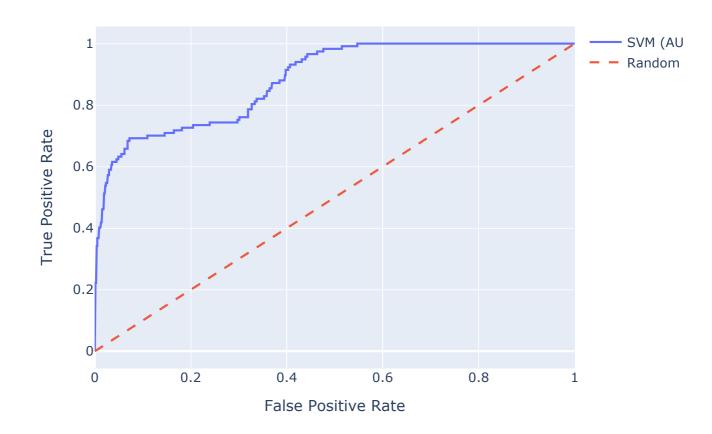
Precision: 0.8590 | AUC: 0.7192404817404817

Confusion Matrix:

[[777 0]

[116 1]]
```

# ROC Curve - Type\_of\_Food\_Allergy\_Oral\_Syndrom - SVM



```
Target: Type_of_Food_Allergy_Other_Legumes | Model: RandomForest
Accuracy: 0.9799

F1 (0): 0.9898 | F1 (1): 0.0000
Precision: 0.9646 | AUC: 0.7299536311389759
Confusion Matrix:
[[878 0]
[ 0 16]]
```

# ROC Curve - Type\_of\_Food\_Allergy\_Other\_Legumes - RandomForest





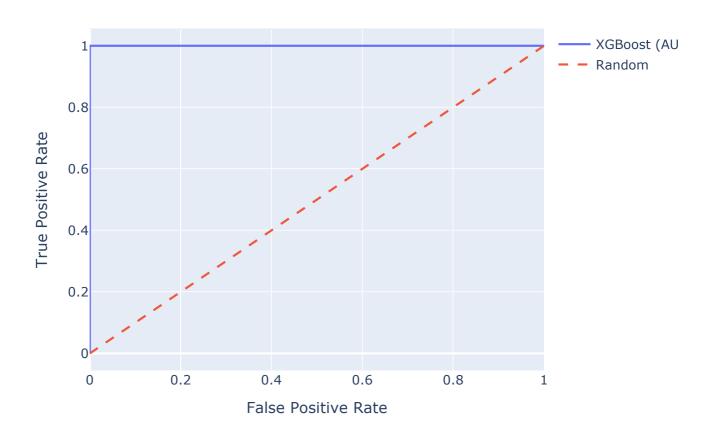
```
Target: Type_of_Food_Allergy_Other_Legumes | Model: XGBoost Accuracy: 0.9788

F1 (0): 0.9892 | F1 (1): 0.0667

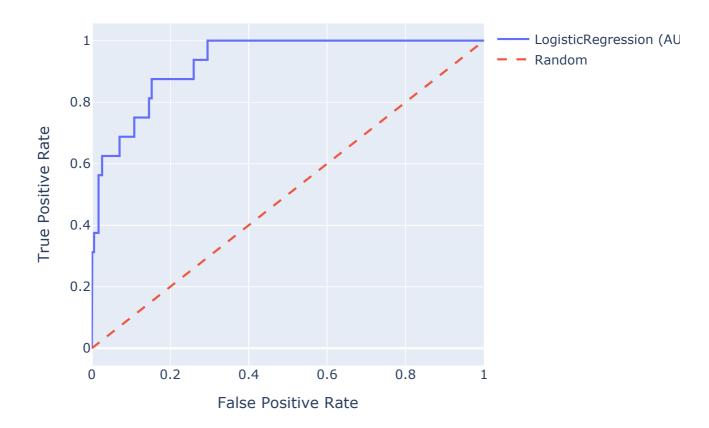
Precision: 0.9662 | AUC: 0.6641588296760711

Confusion Matrix:
[[878 0]
[ 0 16]]
```

# ROC Curve - Type\_of\_Food\_Allergy\_Other\_Legumes - XGBoost



### ROC Curve - Type\_of\_Food\_Allergy\_Other\_Legumes - LogisticRegress



```
Target: Type_of_Food_Allergy_Other_Legumes | Model: SVM Accuracy: 0.7806

F1 (0): 0.8747 | F1 (1): 0.0473

Precision: 0.9677 | AUC: 0.5754049111807733

Confusion Matrix:
[[878 0]
[ 16 0]]
```

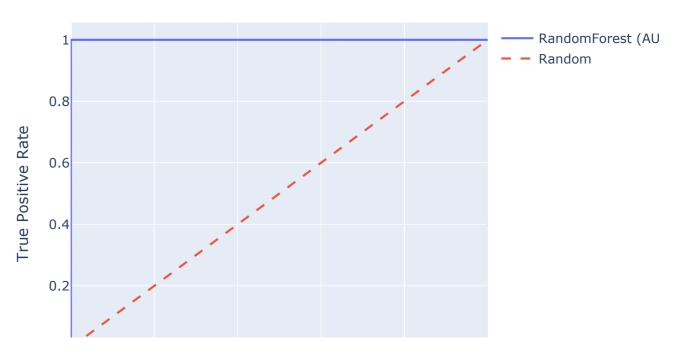
ROC Curve - Type\_of\_Food\_Allergy\_Other\_Legumes - SVM

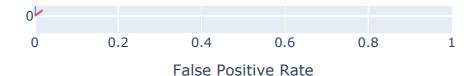


```
Target: Type_of_Food_Allergy_Peanut | Model: RandomForest
Accuracy: 0.9127

F1 (0): 0.9536 | F1 (1): 0.2453
Precision: 0.8940 | AUC: 0.8190126359094917
Confusion Matrix:
[[826 0]
[ 0 68]]
```

# ROC Curve - Type\_of\_Food\_Allergy\_Peanut - RandomForest





```
Target: Type_of_Food_Allergy_Peanut | Model: XGBoost Accuracy: 0.9072

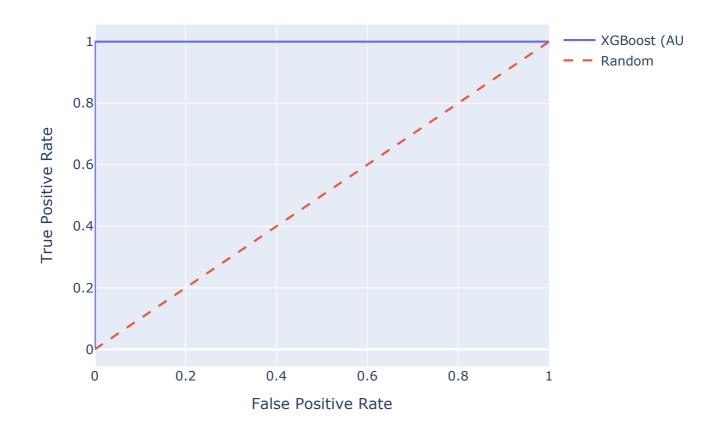
F1 (0): 0.9502 | F1 (1): 0.3065

Precision: 0.8972 | AUC: 0.8572869876719421

Confusion Matrix:

[[826 0]
   0 68]]
```

#### ROC Curve - Type\_of\_Food\_Allergy\_Peanut - XGBoost



```
Target: Type_of_Food_Allergy_Peanut | Model: LogisticRegression Accuracy: 0.8657

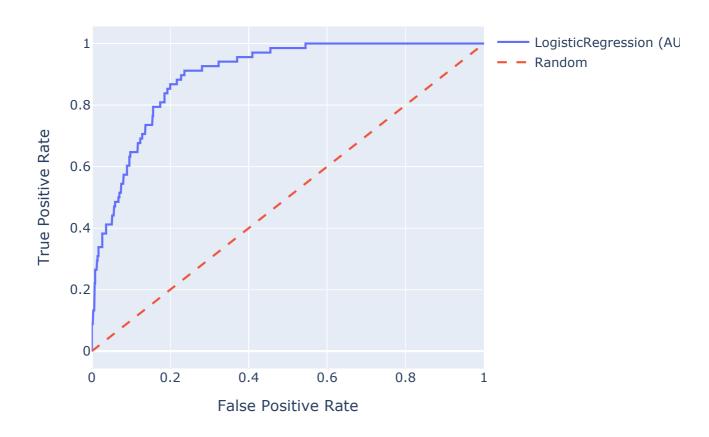
F1 (0): 0.9260 | F1 (1): 0.2537

Precision: 0.8876 | AUC: 0.6785497390257895

Confusion Matrix:

[[821 5]
[55 13]]
```

#### ROC Curve - Type\_of\_Food\_Allergy\_Peanut - LogisticRegression



```
Target: Type_of_Food_Allergy_Peanut | Model: SVM Accuracy: 0.7369

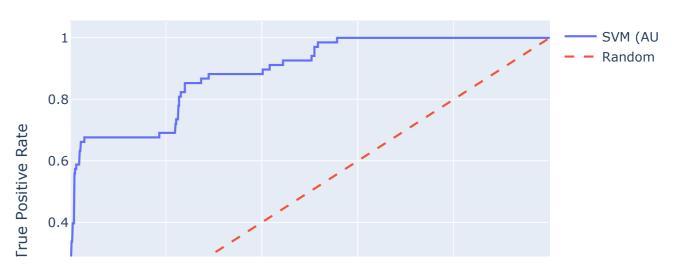
F1 (0): 0.8377 | F1 (1): 0.2782

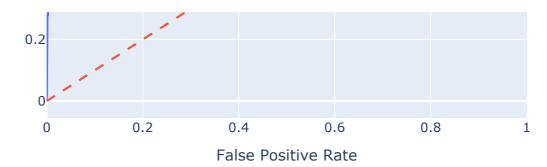
Precision: 0.9020 | AUC: 0.7270146789247581

Confusion Matrix:

[[826 0]
[68 0]]
```

### ROC Curve - Type\_of\_Food\_Allergy\_Peanut - SVM





```
Target: Type_of_Food_Allergy_Shellfish | Model: RandomForest Accuracy: 0.9698

F1 (0): 0.9847 | F1 (1): 0.0500

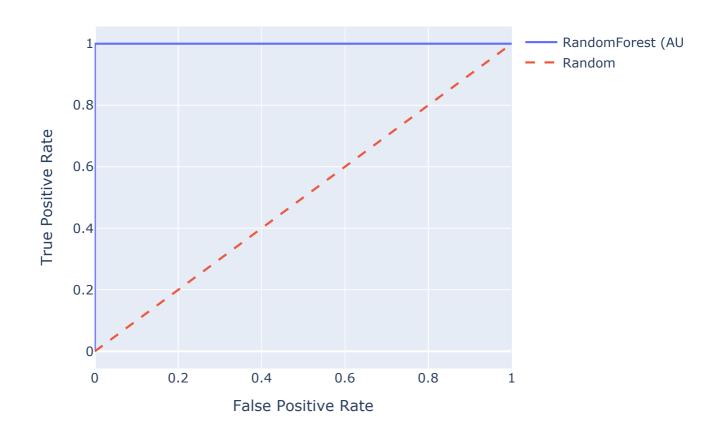
Precision: 0.9450 | AUC: 0.7830905283792213

Confusion Matrix:

[[867 0]

[ 0 27]]
```

### ROC Curve - Type\_of\_Food\_Allergy\_Shellfish - RandomForest



Target: Type\_of\_Food\_Allergy\_Shellfish | Model: XGBoost Accuracy: 0.9609

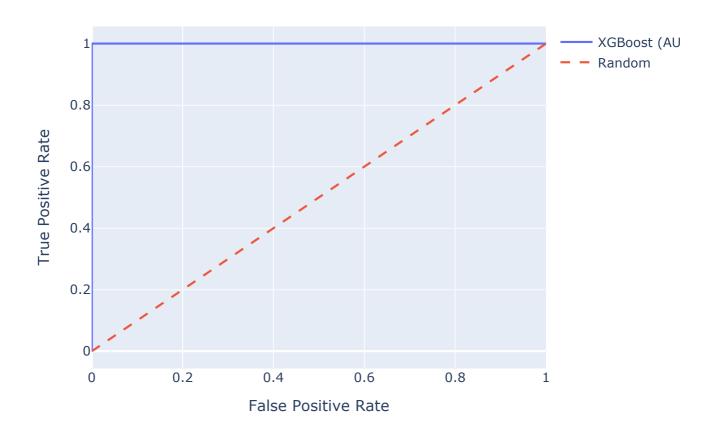
F1 (0): 0.9799 | F1 (1): 0.2000

Precision: 0.9547 | AUC: 0.8007885592087678

Confusion Matrix:

[[867 0] [ 0 27]]

### ROC Curve - Type\_of\_Food\_Allergy\_Shellfish - XGBoost



```
Target: Type_of_Food_Allergy_Shellfish | Model: LogisticRegression Accuracy: 0.9250

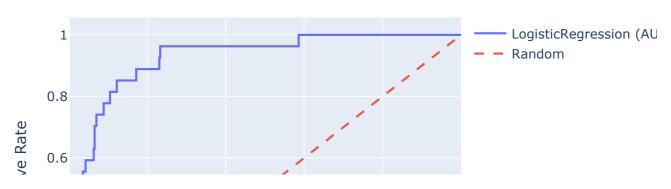
F1 (0): 0.9606 | F1 (1): 0.1430

Precision: 0.9500 | AUC: 0.7110665597433841

Confusion Matrix:

[[867 0]
[27 0]]
```

### ROC Curve - Type\_of\_Food\_Allergy\_Shellfish - LogisticRegression





```
Target: Type_of_Food_Allergy_Shellfish | Model: SVM Accuracy: 0.7638

F1 (0): 0.8619 | F1 (1): 0.1352

Precision: 0.9566 | AUC: 0.7325180433039294

Confusion Matrix:

[[867 0]
[27 0]]
```

### ROC Curve - Type\_of\_Food\_Allergy\_Shellfish - SVM



A Target. Type of Food Allergy TPO | Model. RandomForest

```
Accuracy: 0.9396

F1 (0): 0.9685 | F1 (1): 0.2317

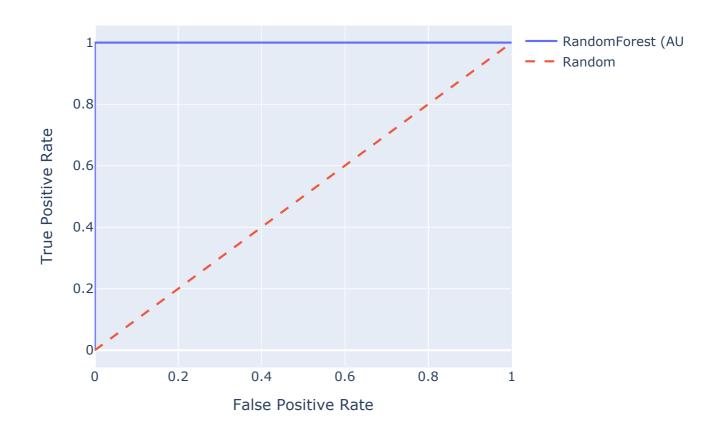
Precision: 0.9247 | AUC: 0.8500644257703082

Confusion Matrix:

[[848 0]

[ 0 46]]
```

#### ROC Curve - Type\_of\_Food\_Allergy\_TPO - RandomForest



```
Target: Type_of_Food_Allergy_TPO | Model: XGBoost Accuracy: 0.9317

F1 (0): 0.9641 | F1 (1): 0.2683

Precision: 0.9277 | AUC: 0.8226974789915967

Confusion Matrix:
[[848 0]
[ 0 46]]
```

### ROC Curve - Type\_of\_Food\_Allergy\_TPO - XGBoost

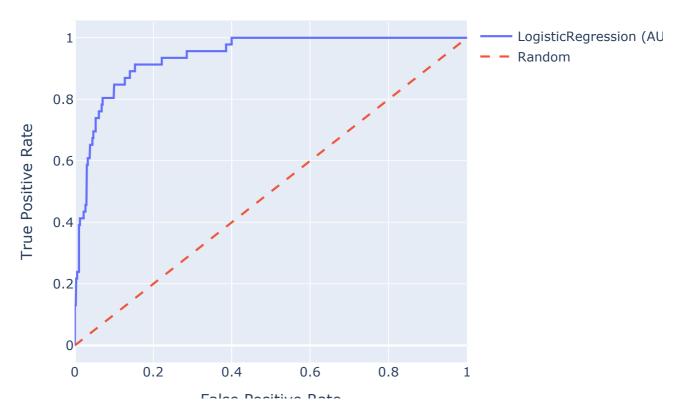




```
Target: Type_of_Food_Allergy_TPO | Model: LogisticRegression
Accuracy: 0.9127

F1 (0): 0.9536 | F1 (1): 0.2230
Precision: 0.9231 | AUC: 0.7879887955182073
Confusion Matrix:
[[845 3]
[ 36 10]]
```

### ROC Curve - Type\_of\_Food\_Allergy\_TPO - LogisticRegression



#### raise Positive Kate

```
Target: Type_of_Food_Allergy_TPO | Model: SVM Accuracy: 0.8064

F1 (0): 0.8873 | F1 (1): 0.2910

Precision: 0.9433 | AUC: 0.8249019607843138

Confusion Matrix:

[[848 0]
[46 0]]
```

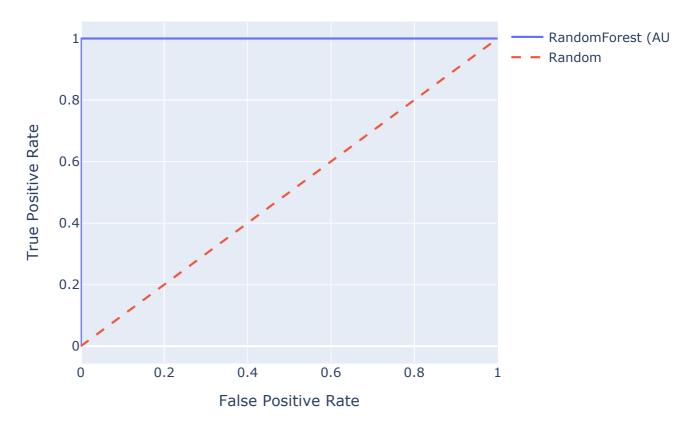
#### ROC Curve - Type\_of\_Food\_Allergy\_TPO - SVM



```
Target: Type_of_Food_Allergy_Tree_Nuts | Model: RandomForest Accuracy: 0.9172

F1 (0): 0.9560 | F1 (1): 0.2820
Precision: 0.8992 | AUC: 0.8458188153310104
Confusion Matrix:
[[820 0]
[ 0 74]]
```

ROC Curve - Type\_of\_Food\_Allergy\_Tree\_Nuts - RandomForest



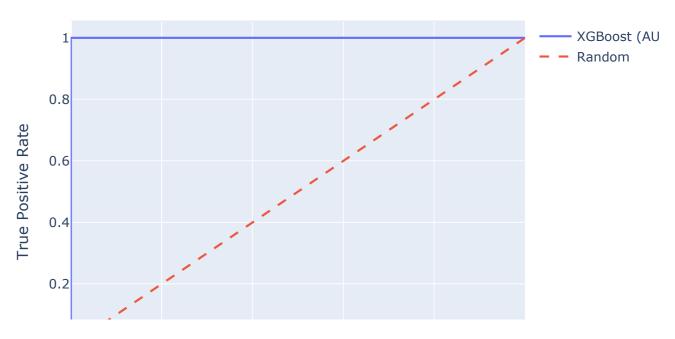
```
Target: Type_of_Food_Allergy_Tree_Nuts | Model: XGBoost Accuracy: 0.9183

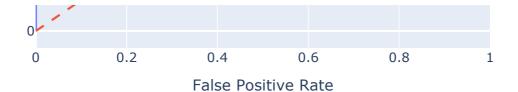
F1 (0): 0.9560 | F1 (1): 0.4085

Precision: 0.9070 | AUC: 0.8562282229965158

Confusion Matrix:
[[820 0]
[ 0 74]]
```

### ROC Curve - Type\_of\_Food\_Allergy\_Tree\_Nuts - XGBoost





```
Target: Type_of_Food_Allergy_Tree_Nuts | Model: LogisticRegression Accuracy: 0.8681

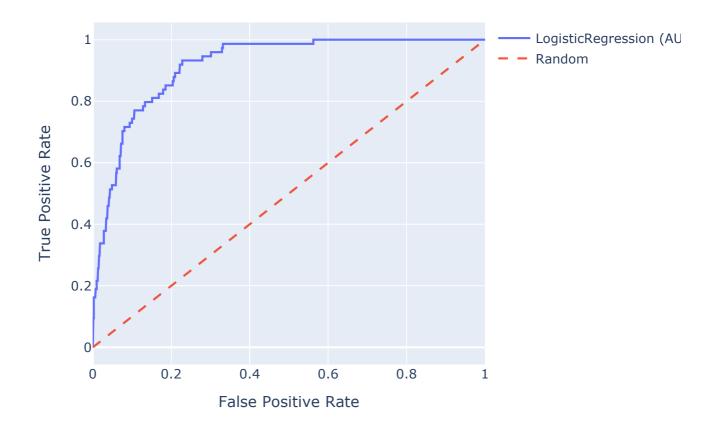
F1 (0): 0.9271 | F1 (1): 0.2944

Precision: 0.8830 | AUC: 0.7658101045296167

Confusion Matrix:

[[813 7]
[ 60 14]]
```

#### ROC Curve - Type\_of\_Food\_Allergy\_Tree\_Nuts - LogisticRegression



```
Target: Type_of_Food_Allergy_Tree_Nuts | Model: SVM Accuracy: 0.7002

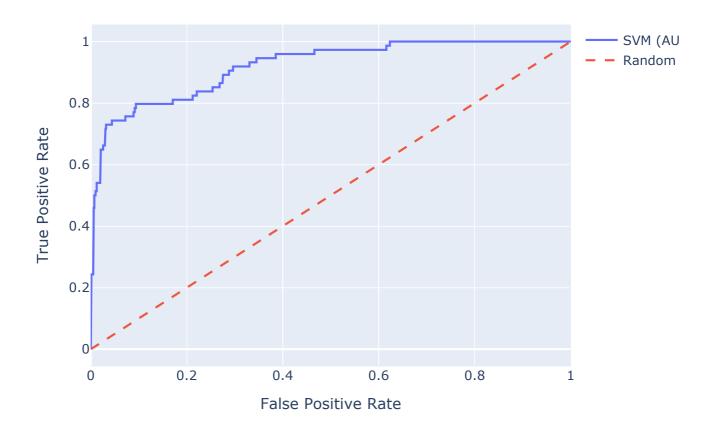
F1 (0): 0.8106 | F1 (1): 0.2641

Precision: 0.8922 | AUC: 0.7194033101045296

Confusion Matrix:

[[820 0]
[74 0]]
```

#### ROC Curve - Type\_of\_Food\_Allergy\_Tree\_Nuts - SVM

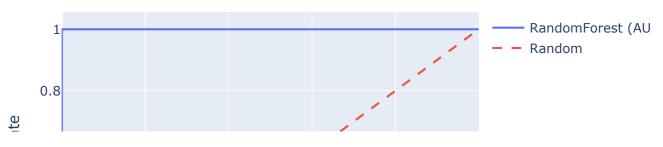


```
import pandas as pd
import numpy as np
from sklearn.model_selection import StratifiedKFold
from sklearn.ensemble import RandomForestClassifier
from sklearn.linear_model import LogisticRegression
from sklearn.svm import SVC
from xgboost import XGBClassifier
from sklearn.metrics import (
    f1_score, accuracy_score, recall_score,
    precision_score, confusion_matrix, roc_auc_score, roc_curve
)
from imblearn.over_sampling import SMOTE
import plotly.graph_objects as go
V1_venom = V1[V1["Venom_Allergy"] == 1]
targets = ["Type_of_Venom_Allergy_ATCD_Venom",
    "Type_of_Venom_Allergy_IGE_Venom"]
models = {
    "RandomForest": RandomForestClassifier(random_state=42),
    "XGBoost": XGBClassifier(random_state=42, eval_metric="logloss", use_label_
```

```
"LogisticRegression": LogisticRegression(max_iter=1000, random_state=42),
    "SVM": SVC(probability=True, random_state=42)
}
X=V1_venom_copy()
X.drop(target 1, axis=1, inplace=True)
X.drop(extra_columns, axis=1, inplace=True)
X.drop(extra, axis=1, inplace=True)
X.drop(inconnu, axis=1, inplace=True)
X = X.iloc[:, 1:]
results_venom = []
kfold = StratifiedKFold(n_splits=10, shuffle=True, random_state=42)
for target in targets:
    y = V1_venom[target]
    for model_name, base_model in models.items():
        f1_class0_scores, f1_class1_scores = [], []
        precision_scores, acc_scores, recall_scores, auc_scores = [], [], [], |
        for train_idx, test_idx in kfold.split(X, y):
            X_train, X_test = X.iloc[train_idx], X.iloc[test_idx]
            y_train, y_test = y.iloc[train_idx], y.iloc[test_idx]
            smote = SMOTE(random_state=42)
            X_train_res, y_train_res = smote.fit_resample(X_train, y_train)
            base_model.fit(X_train_res, y_train_res)
            y_pred = base_model.predict(X_test)
            acc_scores.append(accuracy_score(y_test, y_pred))
            recall_scores.append(recall_score(y_test, y_pred, zero_division=0))
            precision_scores.append(precision_score(y_test, y_pred, average='we
            f1_class0_scores.append(f1_score(y_test, y_pred, pos_label=0, zero_
            f1_class1_scores.append(f1_score(y_test, y_pred, pos_label=1, zero_
            if hasattr(base_model, "predict_proba"):
                y_proba = base_model.predict_proba(X_test)[:, 1]
                auc_scores.append(roc_auc_score(y_test, y_proba))
        base_model.fit(X, y)
        y_pred_full = base_model.predict(X)
        y_proba_full = base_model.predict_proba(X)[:, 1] if hasattr(base_model,
        matrix = confusion_matrix(y, y_pred_full)
        print(f"\nQ Target: {target} | Model: {model name}")
        print(f" Accuracy: {np.mean(acc_scores):.4f}")
```

```
print(f''@ F1 (0): {np.mean(f1_class0_scores):.4f} | F1 (1): {np.mean(
        print(f" Precision: {np.mean(precision_scores):.4f} | AUC: {np.mean(≀
        print(" Confusion Matrix:\n", matrix)
        if y_proba_full is not None:
            fpr, tpr, _ = roc_curve(y, y_proba_full)
            fig = go.Figure()
            fig.add_trace(go.Scatter(x=fpr, y=tpr, mode='lines', name=f"{model_
            fig.add_trace(go.Scatter(x=[0, 1], y=[0, 1], mode='lines', name='Ra
            fig.update layout(
                title=f"ROC Curve - {target} - {model_name}",
                xaxis_title="False Positive Rate",
                yaxis_title="True Positive Rate",
               width=700, height=500
            )
            fig.show()
        results_venom.append({
            "Target": target,
            "Model": model name,
            "F1_Class_0": np.mean(f1_class0_scores),
            "F1_Class_1": np.mean(f1_class1_scores),
            "Precision": np.mean(precision_scores),
            "Accuracy": np.mean(acc scores),
            "Recall": np.mean(recall_scores),
            "AUC_ROC": np.mean(auc_scores) if auc_scores else np.nan
       })
pd.DataFrame(results_venom).to_csv("results_V1_venom.csv", index=False)
      Target: Type_of_Venom_Allergy_ATCD_Venom | Model: RandomForest
     ✓ Accuracy: 0.8944
      F1 (0): 0.9354 | F1 (1): 0.6500
    Precision: 0.9175 | AUC: 0.9125
    Confusion Matrix:
     [[81 0]
     [ 0 16]]
```

#### ROC Curve - Type\_of\_Venom\_Allergy\_ATCD\_Venom - RandomForest

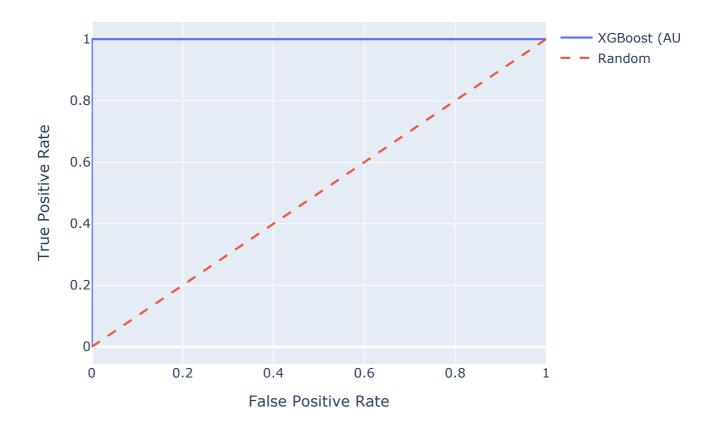




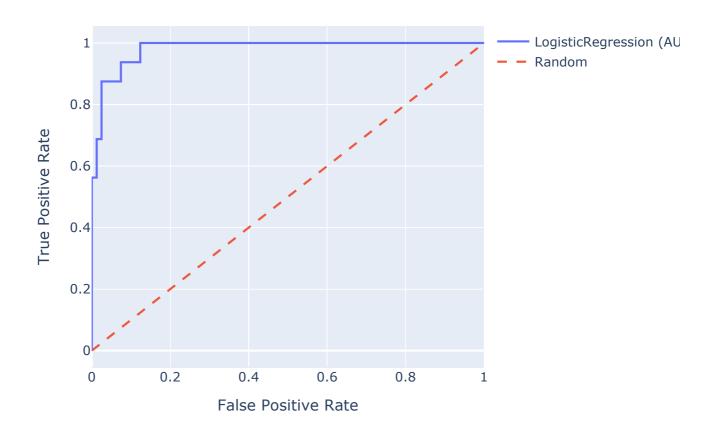
```
Target: Type_of_Venom_Allergy_ATCD_Venom | Model: XGBoost Accuracy: 0.8744

F1 (0): 0.9212 | F1 (1): 0.6200
Precision: 0.9061 | AUC: 0.90625
Confusion Matrix:
[[81 0]
[ 0 16]]
```

### ROC Curve - Type\_of\_Venom\_Allergy\_ATCD\_Venom - XGBoost



#### ROC Curve - Type\_of\_Venom\_Allergy\_ATCD\_Venom - LogisticRegress



```
Target: Type_of_Venom_Allergy_ATCD_Venom | Model: SVM Accuracy: 0.6822

F1 (0): 0.7608 | F1 (1): 0.4655

Precision: 0.8720 | AUC: 0.90277777777779

Confusion Matrix:
[[81 0]
[16 0]]
```

## ROC Curve - Type\_of\_Venom\_Allergy\_ATCD\_Venom - SVM





```
Target: Type_of_Venom_Allergy_IGE_Venom | Model: RandomForest

Accuracy: 0.9900

F1 (0): 0.9000 | F1 (1): 0.9947

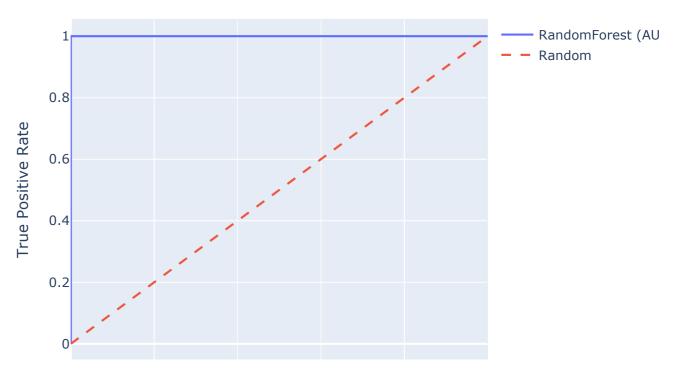
Precision: 0.9810 | AUC: 1.0

Confusion Matrix:

[[10 0]

[ 0 87]]
```

### ROC Curve - Type\_of\_Venom\_Allergy\_IGE\_Venom - RandomForest





```
Target: Type_of_Venom_Allergy_IGE_Venom | Model: XGBoost Accuracy: 1.0000

F1 (0): 1.0000 | F1 (1): 1.0000

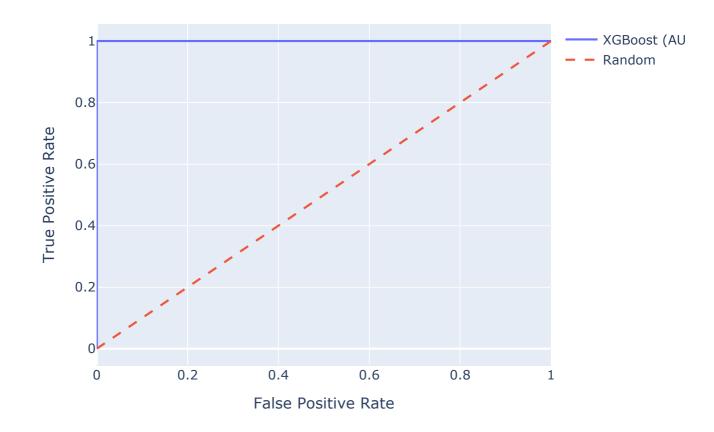
Precision: 1.0000 | AUC: 1.0

Confusion Matrix:

[[10 0]

[ 0 87]]
```

#### ROC Curve - Type\_of\_Venom\_Allergy\_IGE\_Venom - XGBoost



```
Target: Type_of_Venom_Allergy_IGE_Venom | Model: LogisticRegression Accuracy: 0.8622

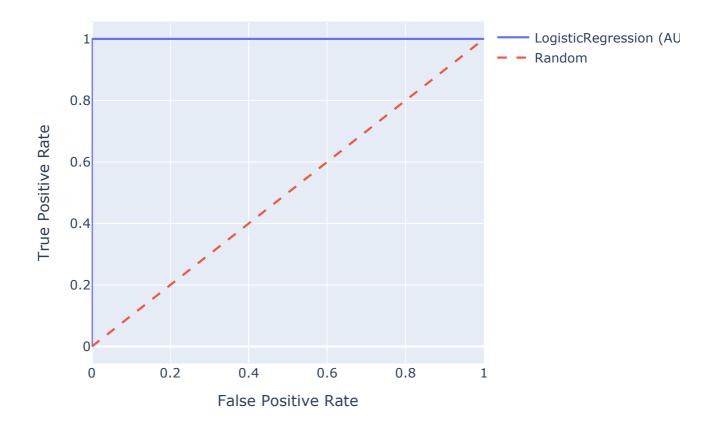
F1 (0): 0.3833 | F1 (1): 0.9198

Precision: 0.8786 | AUC: 0.90416666666666

Confusion Matrix:

[[10 0]
[ 0 87]]
```

ROC Curve - Type\_of\_Venom\_Allergy\_IGE\_Venom - LogisticRegressic



```
Target: Type_of_Venom_Allergy_IGE_Venom | Model: SVM Accuracy: 0.6756

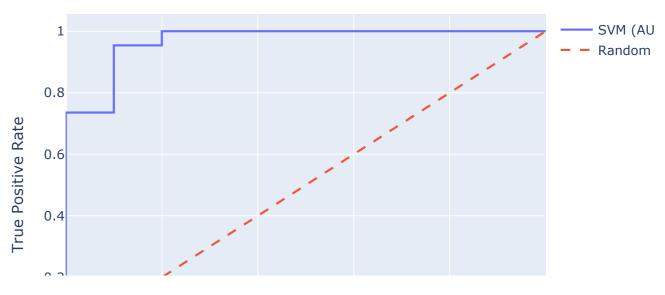
F1 (0): 0.3086 | F1 (1): 0.7755

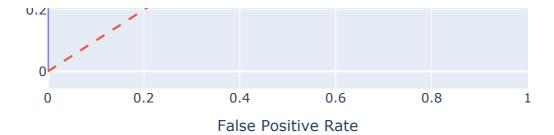
Precision: 0.8740 | AUC: 0.8458333333333333

Confusion Matrix:

[[ 0 10]
 [ 0 87]]
```

ROC Curve - Type\_of\_Venom\_Allergy\_IGE\_Venom - SVM





# Ne lancer pas cette partie, c pour la recherche des hyperparametres

[ ] → 1 cell hidden

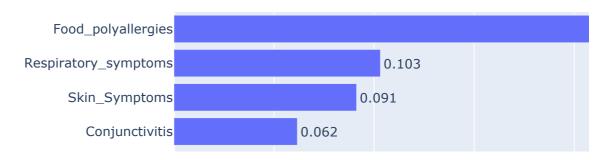
### TOP Features

```
import pandas as pd
import numpy as np
from xgboost import XGBClassifier
import plotly.graph_objects as go
targets = [
    "Allergy_Present", "Respiratory_Allergy", "Food_Allergy", "Venom_Allergy",
    "Severe_Allergy", "Type_of_Food_Allergy_Other", "Type_of_Respiratory_Allergy
    "Type_of_Respiratory_Allergy_IGE_Pollen_Tree", "Type_of_Respiratory_Allergy_
    "Type_of_Respiratory_Allergy_IGE_Mite_Cockroach", "Type_of_Respiratory_Aller
    "Type_of_Respiratory_Allergy_ARIA", "Type_of_Respiratory_Allergy_CONJ",
    "Type_of_Respiratory_Allergy_IGE_Pollen_Gram", "Type_of_Respiratory_Allergy_
    "Type_of_Food_Allergy_Aromatics", "Type_of_Food_Allergy_Cereals_&_Seeds",
    "Type_of_Food_Allergy_Egg", "Type_of_Food_Allergy_Fish", "Type_of_Food_Aller
    "Type_of_Food_Allergy_Mammalian_Milk", "Type_of_Food_Allergy_Oral_Syndrom",
    "Type_of_Food_Allergy_Other_Legumes", "Type_of_Food_Allergy_Peanut",
    "Type_of_Food_Allergy_Shellfish", "Type_of_Food_Allergy_TPO", "Type_of_Food_
    "Type_of_Venom_Allergy_ATCD_Venom", "Type_of_Venom_Allergy_IGE_Venom"
1
inconnu = ["Treatment_of_athsma_9", "Treatment_of_rhinitis_9", "General_cofactor
           "Age_of_onsets_9", "ARIA_(rhinitis)_9", "GINA_(asthma)_9", "Treatment
X = V1.copy()
X.drop(target_1, axis=1, inplace=True)
X.drop(extra_columns, axis=1, inplace=True)
```

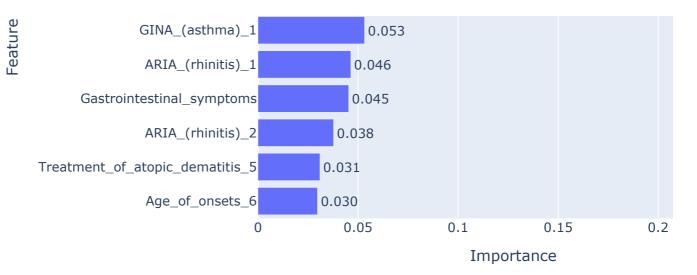
```
X.drop(extra, axis=1, inplace=True)
X.drop(inconnu, axis=1, inplace=True)
X = X.iloc[:, 1:]
def plot_top_features(model, X_sub, y_sub, target):
    if len(np.unique(y sub)) < 2:</pre>
        print(f"    Target '{target}' contient une seule classe ({np.unique(y_sub
        return
    model.fit(X_sub, y_sub)
    importances = model.feature_importances_
    top_indices = np.argsort(importances)[::-1][:10]
    features = X sub.columns[top indices]
    scores = importances[top_indices]
    fig = go.Figure(go.Bar(
        x=scores[::-1],
        y=features[::-1],
        orientation='h',
        text=[f"{s:.3f}" for s in scores[::-1]],
        textposition='outside'
    ))
    fig.update_layout(
        title=f"Top 10 Features pour la cible '{target}' (XGBoost)",
        xaxis_title="Importance",
        yaxis title="Feature",
        width=800, height=500
    fig.show()
for target in targets:
    X_{sub} = X_{copy}()
    y_sub = V1[target]
    model = XGBClassifier(random_state=42, eval_metric="logloss", use_label_enco
    plot_top_features(model, X_sub, y_sub, target)
```

**₹** 

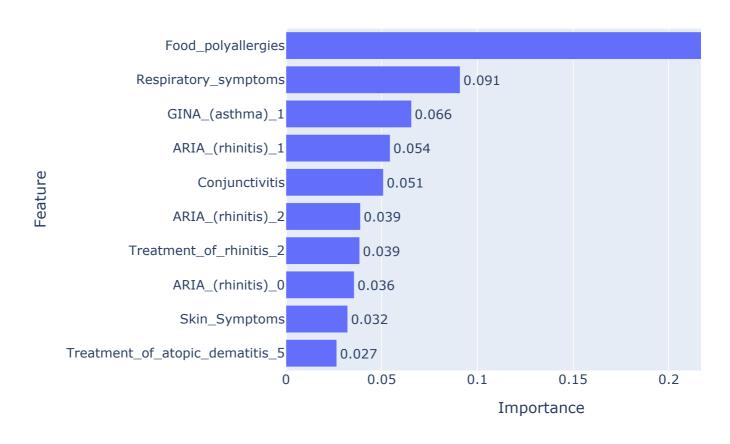
Top 10 Features pour la cible 'Allergy\_Present' (XGBoost)



24/05/2025 17:59

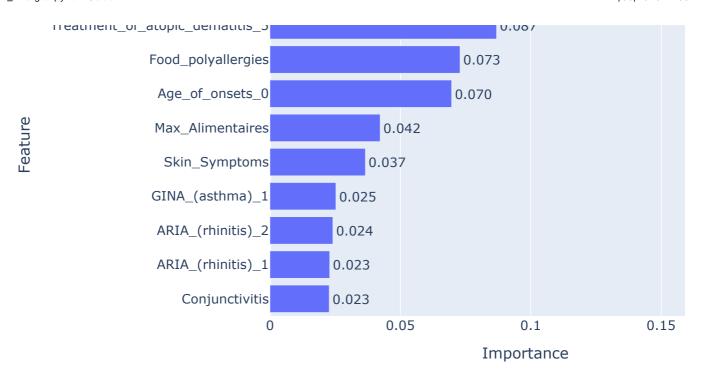


Top 10 Features pour la cible 'Respiratory\_Allergy' (XGBoost)

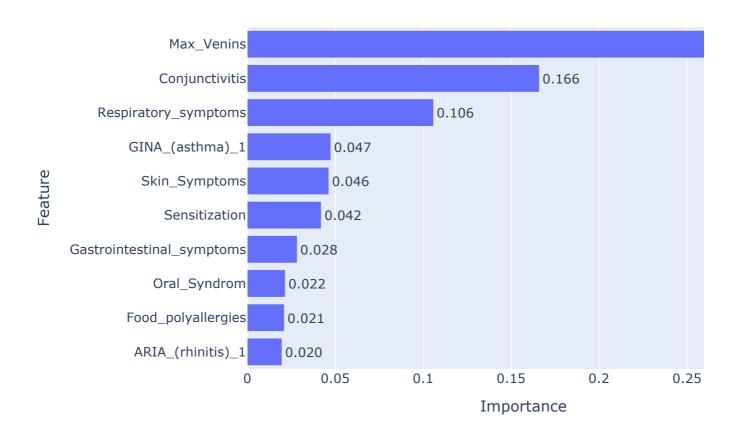


Top 10 Features pour la cible 'Food\_Allergy' (XGBoost)

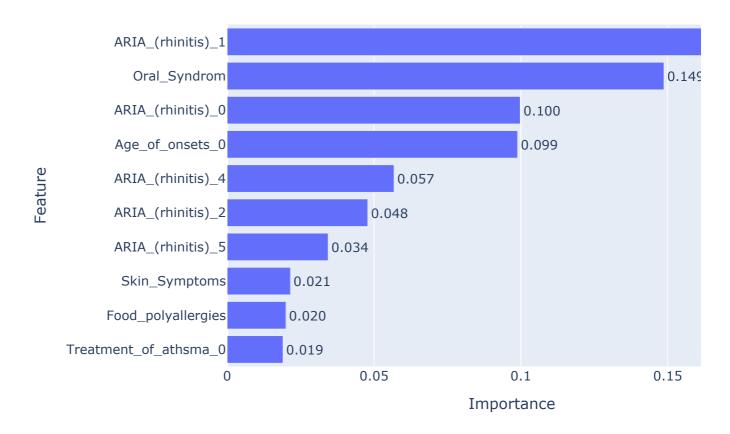




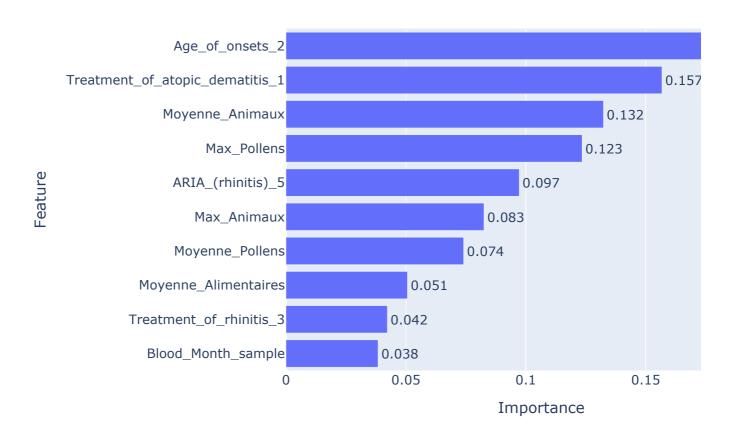
Top 10 Features pour la cible 'Venom\_Allergy' (XGBoost)



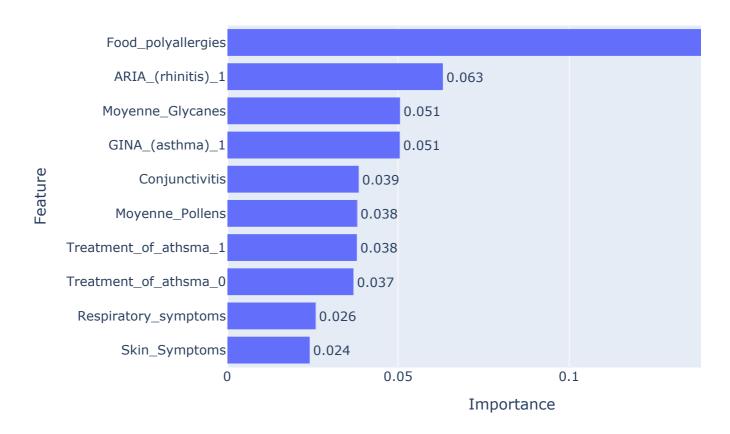
Top 10 Features pour la cible 'Severe\_Allergy' (XGBoost)



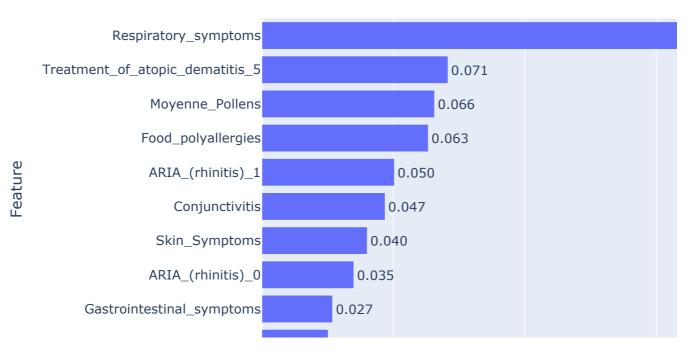
Top 10 Features pour la cible 'Type\_of\_Food\_Allergy\_Other' (XGBoos

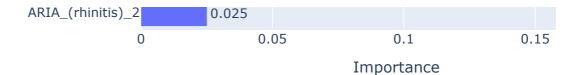


Top 10 Features pour la cible 'Type\_of\_Respiratory\_Allergy\_IGE\_Poll-

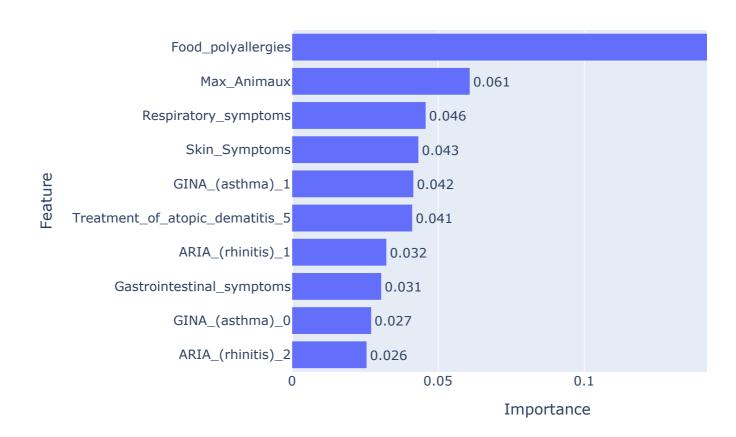


Top 10 Features pour la cible 'Type\_of\_Respiratory\_Allergy\_IGE\_Poll-

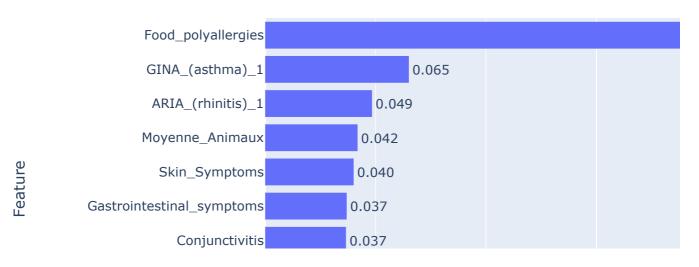


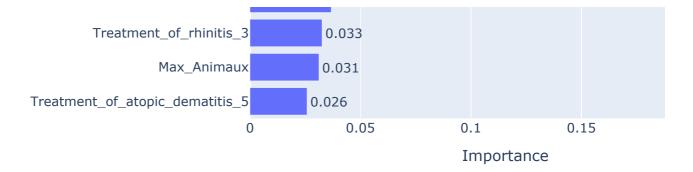


Top 10 Features pour la cible 'Type\_of\_Respiratory\_Allergy\_IGE\_Dan

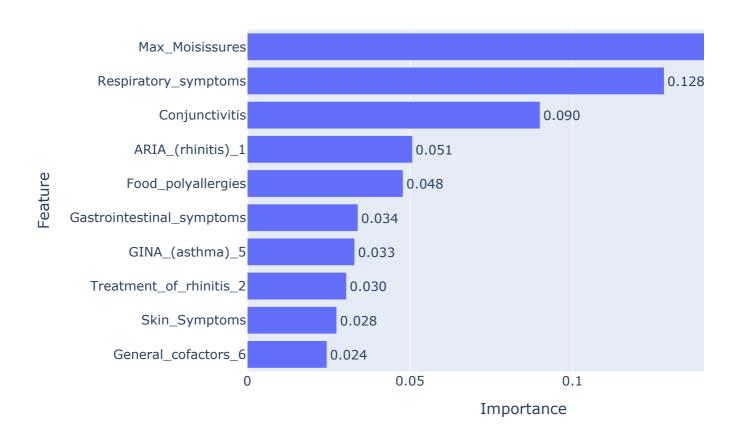


Top 10 Features pour la cible 'Type\_of\_Respiratory\_Allergy\_IGE\_Mite



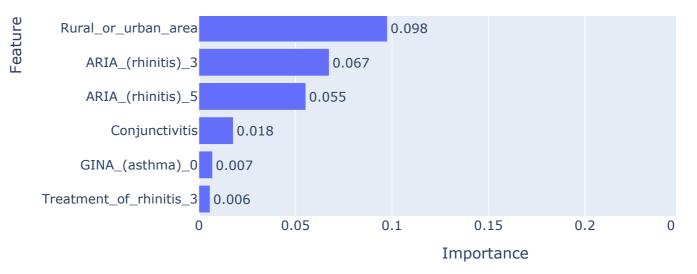


Top 10 Features pour la cible 'Type\_of\_Respiratory\_Allergy\_IGE\_Mol-

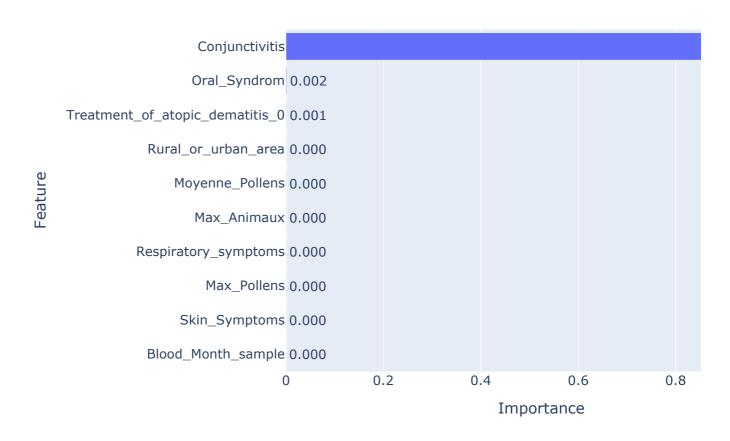


Top 10 Features pour la cible 'Type\_of\_Respiratory\_Allergy\_ARIA' (X



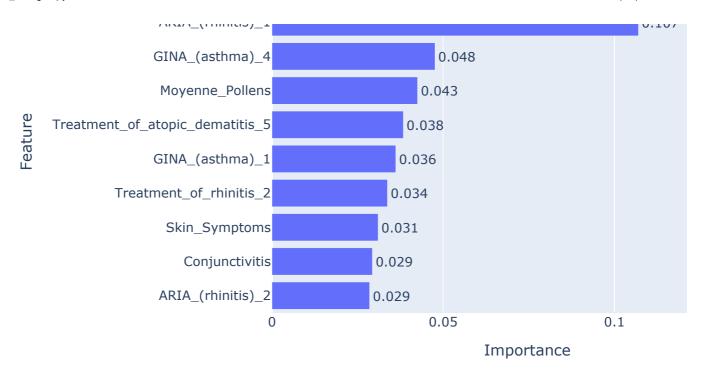


Top 10 Features pour la cible 'Type\_of\_Respiratory\_Allergy\_CONJ' (X

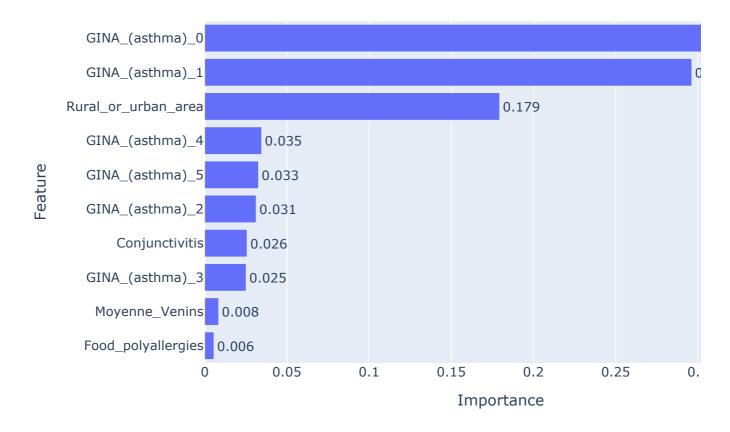


Top 10 Features pour la cible 'Type\_of\_Respiratory\_Allergy\_IGE\_Poll

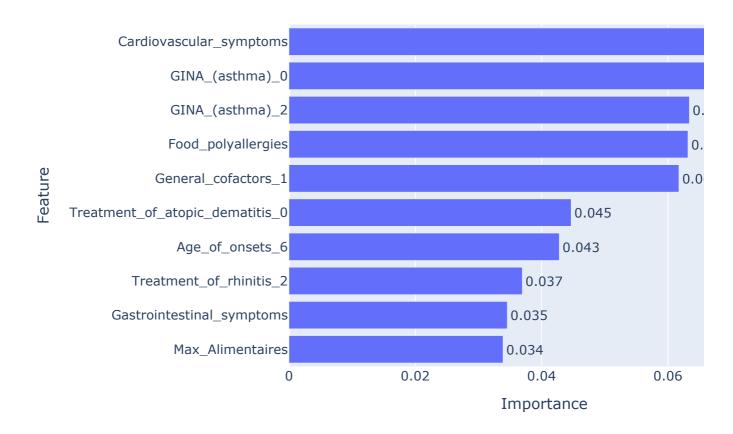




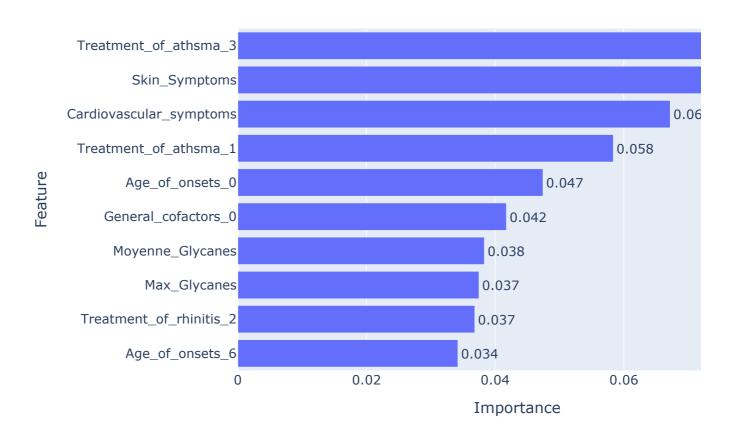
Top 10 Features pour la cible 'Type\_of\_Respiratory\_Allergy\_GINA' (X



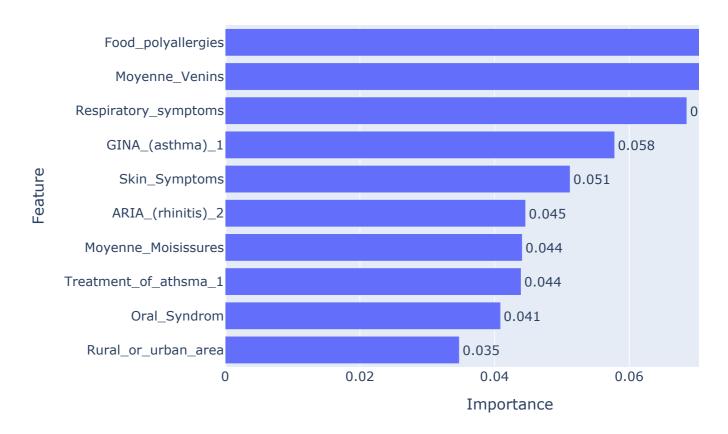
Top 10 Features pour la cible 'Type\_of\_Food\_Allergy\_Aromatics' (XG



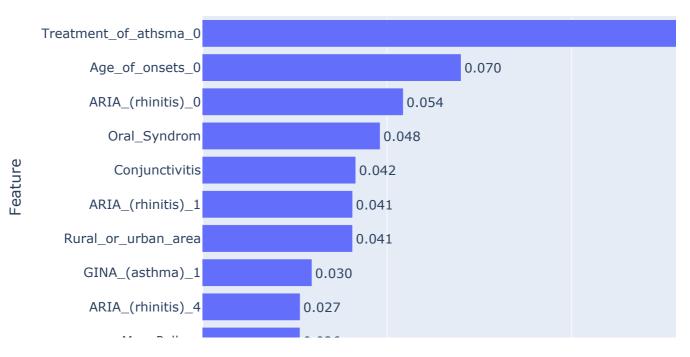
Top 10 Features pour la cible 'Type\_of\_Food\_Allergy\_Cereals\_&\_See

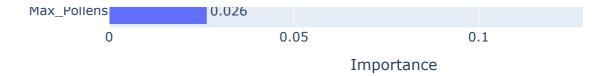


Top 10 Features pour la cible 'Type\_of\_Food\_Allergy\_Egg' (XGBoost)

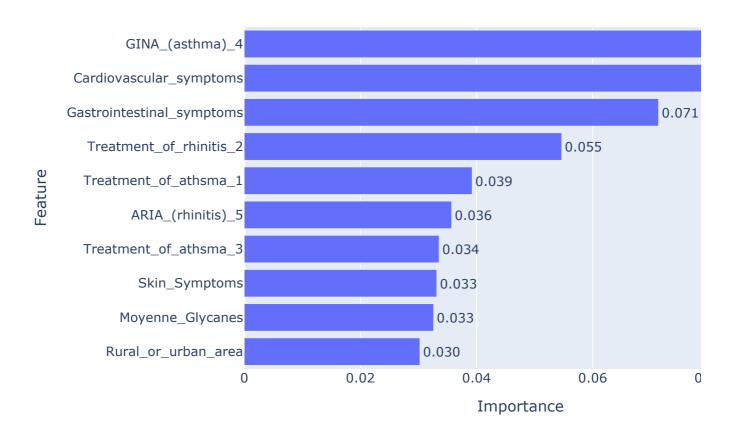


Top 10 Features pour la cible 'Type\_of\_Food\_Allergy\_Fish' (XGBoost)

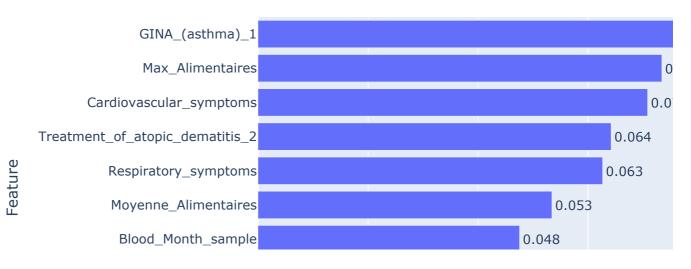


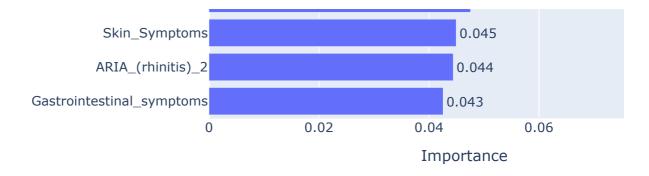


Top 10 Features pour la cible 'Type\_of\_Food\_Allergy\_Fruits\_and\_Vec

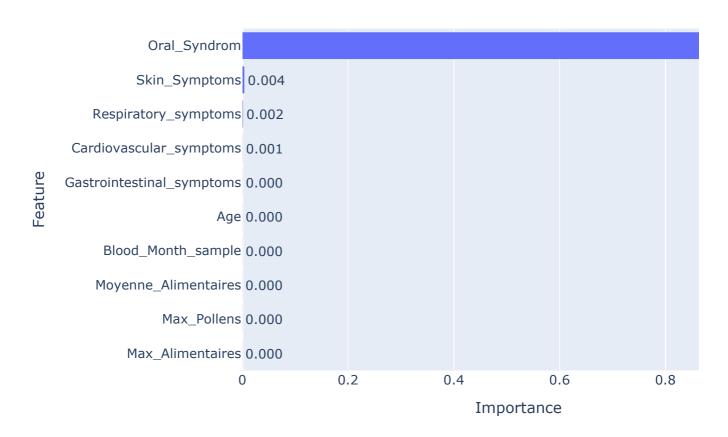


Top 10 Features pour la cible 'Type\_of\_Food\_Allergy\_Mammalian\_Mi

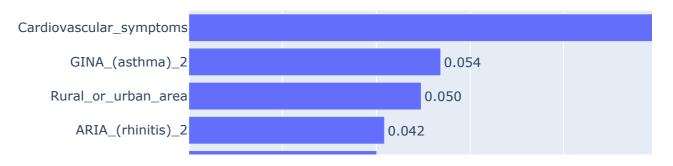


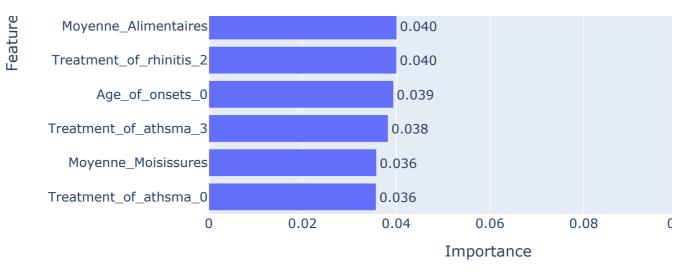


Top 10 Features pour la cible 'Type\_of\_Food\_Allergy\_Oral\_Syndrom'

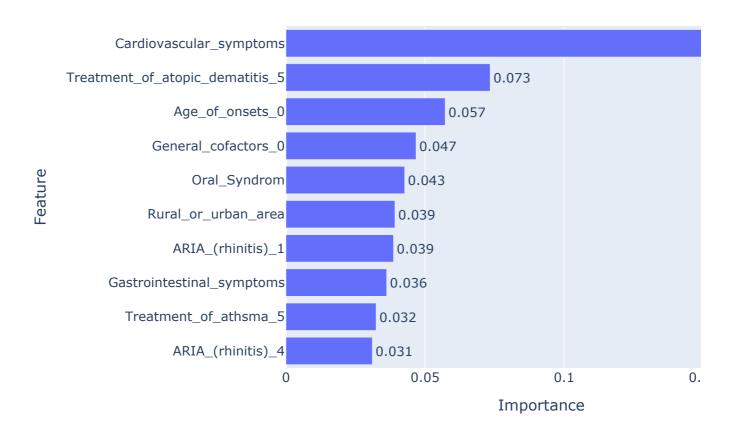


Top 10 Features pour la cible 'Type\_of\_Food\_Allergy\_Other\_Legumes



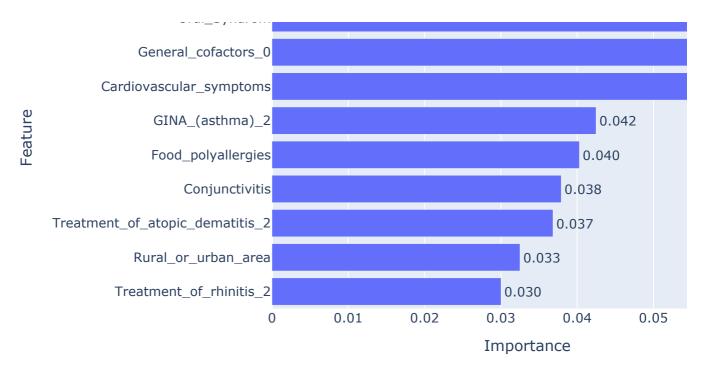


Top 10 Features pour la cible 'Type\_of\_Food\_Allergy\_Peanut' (XGBoc

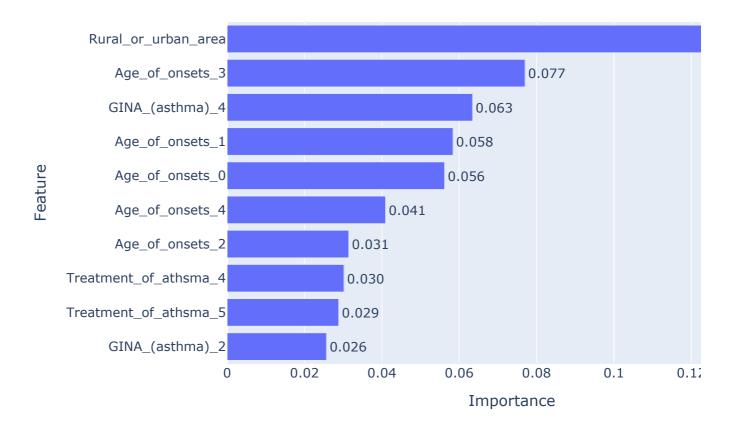


Top 10 Features pour la cible 'Type\_of\_Food\_Allergy\_Shellfish' (XGBe

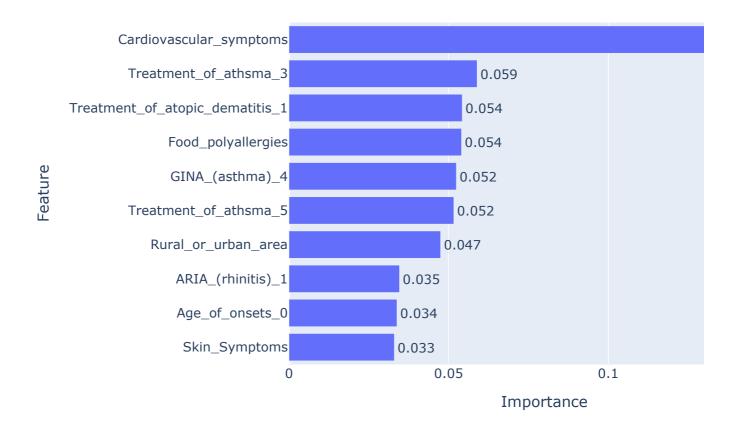




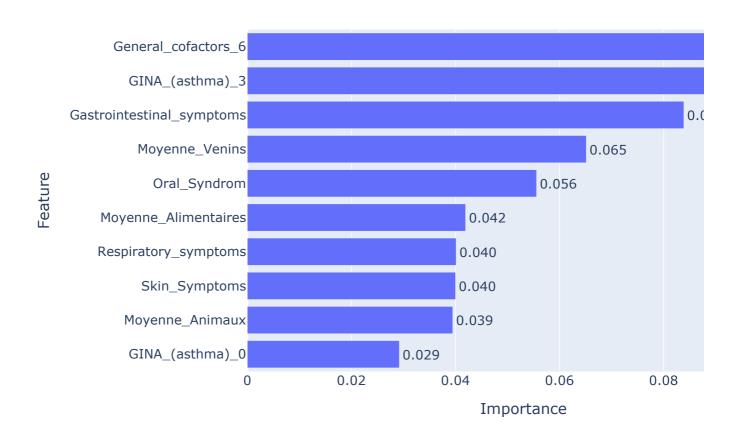
Top 10 Features pour la cible 'Type\_of\_Food\_Allergy\_TPO' (XGBoost)



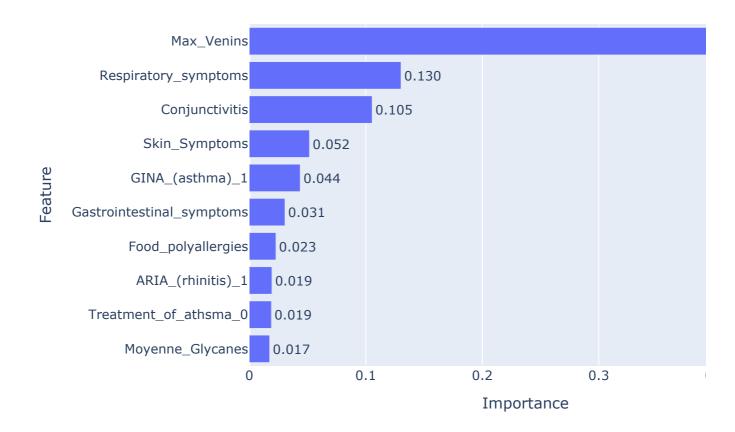
Top 10 Features pour la cible 'Type\_of\_Food\_Allergy\_Tree\_Nuts' (XG



Top 10 Features pour la cible 'Type\_of\_Venom\_Allergy\_ATCD\_Venon



Top 10 Features pour la cible 'Type\_of\_Venom\_Allergy\_IGE\_Venom'



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