

```
diabetes_codes = {
    "Regular insulin dose": 33,
    "NPH insulin dose": 34,
    "UltraLente insulin dose": 35,
    "Unspecified blood glucose measurement": 48,
    "Pre-breakfast blood glucose measurement": 58,
    "Post-breakfast blood glucose measurement": 59,
    "Pre-lunch blood glucose measurement": 60,
    "Post-lunch blood glucose measurement": 61,
    "Pre-supper blood glucose measurement": 62,
    "Post-supper blood glucose measurement": 63,
    "Pre-snack blood glucose measurement": 64,
    "Hypoglycemic symptoms": 65,
    "Typical meal ingestion": 66,
    "More-than-usual meal ingestion": 67,
    "Less-than-usual meal ingestion": 68,
    "Typical exercise activity": 69,
    "More-than-usual exercise activity": 70,
    "Less-than-usual exercise activity": 71,
    "Unspecified special event": 72
}

import pandas as pd
import numpy as np
import matplotlib.pyplot as plt

from sklearn.model_selection import train_test_split
from sklearn.metrics import accuracy_score, f1_score, precision_score, recall_score
from sklearn.preprocessing import StandardScaler

df_diabetes = pd.read_csv("diabetes-clean.csv")
df_heartdis = pd.read_csv("heartdis-clean.csv")
df_thyroid = pd.read_csv("thyroid-clean.csv")
```

```
df_diabetes.info()
```

```
df_heartdis.info()
```

```
df_thyroid.info()
```

```
from sklearn.ensemble import RandomForestClassifier
```

```
# Separate features and target
```

```
X = df_diabetes[['code']]
```

```
y = df_diabetes['diabetes_type']
```

```
# Split data into train and test sets
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
```

```
# Initialize and train the SVM model
```

```
diabetes_model = RandomForestClassifier(random_state=42)
```

```
diabetes_model.fit(X_train, y_train)
```

```
# Make predictions
```

```
y_pred = diabetes_model.predict(X_test)
```

```
prob = diabetes_model.predict_proba(X_test)
```

```
# Calculate metrics
```

```
f1 = f1_score(y_test, y_pred, average='weighted')
```

```
accuracy = accuracy_score(y_test, y_pred)
```

```
precision = precision_score(y_test, y_pred, average='weighted')
```

```
recall = recall_score(y_test, y_pred, average='weighted')
```

```
# Print metrics
```

```
print("F1 Score:", f1)
```

```
print("Accuracy:", accuracy)
```

```
print("Precision:", precision)
```

```
print("Recall:", recall)
```

```
from sklearn.preprocessing import OneHotEncoder
```

```
from sklearn.compose import ColumnTransformer
```

```
from sklearn.pipeline import Pipeline
```

```
from sklearn.svm import SVC
```

```
X = df_heartdis.drop('heartdis_type', axis=1)
```

```
y = df_heartdis['heartdis_type']
```

```
categorical_cols = X.select_dtypes(include=['object']).columns.tolist()
```

```
numerical_cols = X.select_dtypes(include=['int64', 'float64']).columns.tolist()
```

```
bool_cols = X.select_dtypes(include=['bool']).columns.tolist()
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=12)
```

```
preprocessor = ColumnTransformer(
```

```
    transformers=[
```

```
        ('num', 'passthrough', numerical_cols),
```

```
        ('cat', OneHotEncoder(), categorical_cols),
```

```
        ('scaler', StandardScaler(), numerical_cols)
```

```
    ])
```

```
heartdis_model_pipeline = Pipeline(steps=[
```

```
    ('preprocessor', preprocessor),
```

```
    ('model', SVC(probability=True, kernel='rbf', C=10000, random_state=42))
```

```
])
```

```
heartdis_model_pipeline.fit(X_train, y_train)
```

```
y_pred = heartdis_model_pipeline.predict(X_test)
prob = heartdis_model_pipeline.predict_proba(X_test)

f1 = f1_score(y_test, y_pred, average='weighted')
accuracy = accuracy_score(y_test, y_pred)
precision = precision_score(y_test, y_pred, average='weighted')
recall = recall_score(y_test, y_pred, average='weighted')

print("F1 Score:", f1)
print("Accuracy:", accuracy)
print("Precision:", precision)
print("Recall:", recall)
```

```
from sklearn.preprocessing import OneHotEncoder
from sklearn.compose import ColumnTransformer
from sklearn.pipeline import Pipeline
from sklearn.svm import SVC
```

```
X = df_thyroid.drop('has_thyroid_disease', axis=1)
y = df_thyroid['has_thyroid_disease']
```

```
categorical_cols = ['sex']
bool_cols = X.select_dtypes(include=['bool']).columns.tolist()
numerical_cols = X.select_dtypes(include=['int64', 'float64']).columns.tolist()
```

```
preprocessor = ColumnTransformer(
    transformers=[
        ('num', 'passthrough', numerical_cols),
        ('cat', OneHotEncoder(), categorical_cols),
        ('scaler', StandardScaler(), numerical_cols)
    ])
```

```
thyroid_model_pipeline = Pipeline(steps=[
    ('preprocessor', preprocessor),
    ('svm', SVC(probability=True, kernel='rbf', C=10, random_state=42))
])
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
```

```
thyroid_model_pipeline.fit(X_train, y_train)
```

```
y_pred = thyroid_model_pipeline.predict(X_test)
prob = thyroid_model_pipeline.predict_proba(X_test)
```

```
f1 = f1_score(y_test, y_pred)
accuracy = accuracy_score(y_test, y_pred)
precision = precision_score(y_test, y_pred)
recall = recall_score(y_test, y_pred)
```

```
print("F1 Score:", f1)
print("Accuracy:", accuracy)
print("Precision:", precision)
print("Recall:", recall)
```

```
def get_user_data(str):
    """
    Arguments
    - str: 'diabetes', 'heartdis', or 'thyroid'
    """
    if str == "diabetes":
        X = df_diabetes.drop('diabetes_type', axis=1)
        y = df_diabetes['diabetes_type']
```

```

elif str == "heartdis":
    X = df_heartdis.drop('heartdis_type', axis=1)
    y = df_heartdis['heartdis_type']
elif str == "thyroid":
    X = df_thyroid.drop('has_thyroid_disease', axis=1)
    y = df_thyroid['has_thyroid_disease']
else:
    raise ValueError("Invalid input: expected 'diabetes', 'heartdis', or 'thyroid'")
_, X_test, _ = train_test_split(X, y, test_size=0.2, random_state=42)
random_selection = X_test.sample(1)
return random_selection

def get_dscore(row):
    classes = diabetes_model.classes_
    prediction = diabetes_model.predict_proba(row)
    probas = { classes[i]: prediction[0][i] for i in range(len(classes)) }
    return max(probas.values())

def get_hscore(row):
    classes = heartdis_model_pipeline.classes_
    prediction = heartdis_model_pipeline.predict_proba(row)
    probas = { classes[i]: prediction[0][i] for i in range(len(classes)) }
    return sum(probas[class_label] * int(class_label) for class_label in probas)

def get_tscore(row):
    classes = thyroid_model_pipeline.classes_
    prediction = thyroid_model_pipeline.predict_proba(row)
    probas = { classes[i]: prediction[0][i] for i in range(len(classes)) }
    wtscore = sum(probas[class_label] * int(class_label) for class_label in probas)
    return wtscore / (sum([ int(class_label) for class_label in probas ]))

def low(x):

```

```
return max(0, min(1, (0.4 - x) / 0.4))
```

```
def moderate(x):
```

```
    return max(0, min((x - 0.3) / 0.4, (0.7 - x) / 0.4))
```

```
def high(x):
```

```
    return max(0, min((x - 0.6) / 0.4, 1))
```

```
def fuzzify(score):
```

```
    return {  
        "low": low(score),  
        "moderate": moderate(score),  
        "high": high(score)  
    }
```

```
def evaluate_rules(d_score, h_score, t_score):
```

```
    # Fuzzify each input
```

```
    d_fuzzy = fuzzify(d_score)
```

```
    h_fuzzy = fuzzify(h_score)
```

```
    t_fuzzy = fuzzify(t_score)
```

```
    # Initialize outputs for risk levels
```

```
    low_risk, moderate_risk, high_risk = 0, 0, 0
```

```
    # Rules
```

```
    # Rule 1: IF d_score is high OR h_score is high THEN risk is high
```

```
    high_risk = max(high_risk, max(d_fuzzy['high'], h_fuzzy['high']))
```

```
    # Rule 2: IF d_score is moderate AND h_score is moderate AND t_score is moderate THEN risk is moderate
```

```
    moderate_risk = max(moderate_risk, min(d_fuzzy['moderate'], h_fuzzy['moderate'],  
t_fuzzy['moderate']))
```

```
    # Rule 3: IF d_score is low AND h_score is low AND t_score is low THEN risk is low
```

```

low_risk = max(low_risk, min(d_fuzzy['low'], h_fuzzy['low'], t_fuzzy['low']))

# Rule 4: IF d_score is high OR t_score is high THEN risk is moderate
moderate_risk = max(moderate_risk, max(d_fuzzy['high'], t_fuzzy['high']))

return low_risk, moderate_risk, high_risk

```

```

def defuzzify(low_risk, moderate_risk, high_risk):
    risk_levels = {
        "low": 0.25,
        "moderate": 0.5,
        "high": 0.75
    }
    numerator = (low_risk * risk_levels["low"] +
                 moderate_risk * risk_levels["moderate"] +
                 high_risk * risk_levels["high"])
    denominator = low_risk + moderate_risk + high_risk
    return numerator / denominator if denominator != 0 else 0

```

```

def calculate_final_risk(d_score, h_score, t_score, threshold=0.5):
    low_risk, moderate_risk, high_risk = evaluate_rules(d_score, h_score, t_score)
    final_risk_score = defuzzify(low_risk, moderate_risk, high_risk)
    risk_status = "In Danger" if final_risk_score > threshold else "Not in Danger"
    return final_risk_score, risk_status

```

```

d_score = get_dscore(get_user_data('diabetes'))
h_score = get_hscore(get_user_data('heartdis'))
t_score = get_tscore(get_user_data('thyroid'))

```

```

print("Diabetes Score:", d_score)
print("Heart Disease Score:", h_score)
print("Thyroid Score:", t_score)

```



```
final_score, status = calculate_final_risk(d_score, h_score, t_score)
print("Final Risk Score:", final_score)
print("Risk Status:", status)
```