

# **Initial Model Training Code, Model Validation and Evaluation Report**

**Project Name :** Covid - 19 Infant Growth Analysis and Prediction

## **Data Split**

The dataset was split into **training (75%)** and **testing (25%)** sets

## **Features**


- Numerical features: age\_months, height\_cm, weight\_kg, speech\_score, milestone\_score.
- Categorical feature: period (encoded using LabelEncoder)
- Target labels: Infant developmental outcome (encoded using LabelEncoder)

Three machine learning models were trained:

- TabPFNClassifier
- XGBClassifier

**Initial Model Training Code :**

```
from tabpfnclassifier import TabPFNCClassifier
```

 `tabpfnc = TabPFNCClassifier()  
tabpfnc.fit(x_train, y_train)`



▼ TabPFNCClassifier ⓘ

TabPFNCClassifier()

```
y_pred_tabpfnc = tabpfnc.predict(x_test)
```

```
from sklearn.metrics import accuracy_score  
accuracy = accuracy_score(y_test, y_pred_tabpfnc)  
accuracy
```



1.0

▶ `from sklearn.metrics import confusion_matrix, classification_report`

```
cm_tabpfn = confusion_matrix(y_test, y_pred_tabpfn)
print(cm_tabpfn)
```

```
cs_report_tabpfn = classification_report(y_test, y_pred_tabpfn)
print(cs_report_tabpfn)
```



```
[[49  0  0]
 [ 0 53  0]
 [ 0  0 48]]
```

	precision	recall	f1-score	support
0	1.00	1.00	1.00	49
1	1.00	1.00	1.00	53
2	1.00	1.00	1.00	48
accuracy			1.00	150
macro avg	1.00	1.00	1.00	150
weighted avg	1.00	1.00	1.00	150

```
▶ from xgboost import XGBClassifier
  from sklearn.metrics import accuracy_score

  # Initialize XGBoost model
  xgb = XGBClassifier(
      n_estimators=100,
      max_depth=6,
      learning_rate=0.1,
      random_state=42
  )

  # Train the model
  # Use the encoded training labels (y_train_encoded) from the previous cell
  xgb.fit(x_train, y_train)

  # Make predictions
  y_pred_xgb = xgb.predict(x_test)

  # Evaluate the model
  accuracy_xgb = accuracy_score(y_test, y_pred_xgb)
  print(f"Accuracy: {accuracy_xgb}")
```

```
⇒ Accuracy: 0.9866666666666667
```

```
from sklearn.metrics import confusion_matrix, classification_report
```

```
cm_xgb = confusion_matrix(y_test, y_pred_xgb)
print(cm_xgb)
```

```
cs_report_xgb = classification_report(y_test, y_pred_xgb)
print(cs_report_xgb)
```

```
[[49  0  0]
 [ 2 51  0]
 [ 0  0 48]]
```

	precision	recall	f1-score	support
0	0.96	1.00	0.98	49
1	1.00	0.96	0.98	53
2	1.00	1.00	1.00	48
accuracy			0.99	150
macro avg	0.99	0.99	0.99	150
weighted avg	0.99	0.99	0.99	150

## Model Validation and Evaluation Report :

Model	Summary	Training and validation performance metrics																																			
Model - 1 : TabPFNCClassifier	<pre>from tabpfn import TabPFNCClassifier</pre> <pre>tabpfn = TabPFNCClassifier() tabpfn.fit(x_train, y_train)</pre> <pre>TabPFNCClassifier() TabPFNCClassifier()</pre> <pre>y_pred_tabpfn = tabpfn.predict(x_test)</pre>	<pre>from sklearn.metrics import accuracy_score accuracy = accuracy_score(y_test, y_pred_tabpfn) accuracy</pre> <pre>1.0</pre> <pre>from sklearn.metrics import confusion_matrix, classification_report  cm_tabpfn = confusion_matrix(y_test, y_pred_tabpfn) print(cm_tabpfn)  cs_report_tabpfn = classification_report(y_test, y_pred_tabpfn) print(cs_report_tabpfn)</pre> <pre>[[49  0  0]  [ 0 53  0]  [ 0  0 48]]</pre> <table><thead><tr><th></th><th>precision</th><th>recall</th><th>f1-score</th><th>support</th></tr></thead><tbody><tr><td>0</td><td>1.00</td><td>1.00</td><td>1.00</td><td>49</td></tr><tr><td>1</td><td>1.00</td><td>1.00</td><td>1.00</td><td>53</td></tr><tr><td>2</td><td>1.00</td><td>1.00</td><td>1.00</td><td>48</td></tr><tr><td>accuracy</td><td></td><td></td><td>1.00</td><td>150</td></tr><tr><td>macro avg</td><td>1.00</td><td>1.00</td><td>1.00</td><td>150</td></tr><tr><td>weighted avg</td><td>1.00</td><td>1.00</td><td>1.00</td><td>150</td></tr></tbody></table>		precision	recall	f1-score	support	0	1.00	1.00	1.00	49	1	1.00	1.00	1.00	53	2	1.00	1.00	1.00	48	accuracy			1.00	150	macro avg	1.00	1.00	1.00	150	weighted avg	1.00	1.00	1.00	150
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## Model - 2 : XGBClassifier

```
from xgboost import XGBClassifier
from sklearn.metrics import accuracy_score

# Initialize XGBoost model
xgb = XGBClassifier(
    n_estimators=100,
    max_depth=4,
    learning_rate=0.1,
    random_state=2
)

# Train the model
# Use the encoded training labels (y_train_encoded) from the previous cell
xgb.fit(x_train, y_train)

# Make predictions
y_pred_xgb = xgb.predict(x_test)

# Evaluate the model
accuracy_xgb = accuracy_score(y_test, y_pred_xgb)
print(f"Accuracy: {accuracy_xgb}")
```

```
# Evaluate the model
accuracy_xgb = accuracy_score(y_test, y_pred_xgb)
print(f"Accuracy: {accuracy_xgb}")
```

Accuracy: 0.9866666666666667

```
from sklearn.metrics import confusion_matrix, classification_report

cm_xgb = confusion_matrix(y_test, y_pred_xgb)
print(cm_xgb)

cs_report_xgb = classification_report(y_test, y_pred_xgb)
print(cs_report_xgb)
```

```
[[49  0]
 [ 2 51]]
```

	precision	recall	f1-score	support
0	0.98	1.00	0.98	49
1	1.00	0.06	0.98	53
2	1.00	1.00	1.00	48
accuracy			0.99	150
macro avg	0.99	0.99	0.99	150
weighted avg	0.99	0.99	0.99	150