



# **Model Optimization and Tuning Phase Template**

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Team ID	739656
Project Title	Zombie detection using the machine learing
Maximum Marks	10 Marks

### **Model Optimization and Tuning Phase**

The Model Optimization and Tuning Phase involves refining neural network models for peak performance. It includes optimized model code, fine-tuning hyperparameters, comparing performance metrics, and justifying the final model selection for enhanced predictive accuracy and efficiency.

**Hyperparameter Tuning Documentation (8 Marks):** 





Model	Tuned Hyperparameters						
	#importing the library for grid search from sklearn.model_selection import GridSearchCV  The 'lr_param_grid' specifies different values for regularization strength (C), solvers (solver), and penalty types (penalty). GridSearchCV (lr_cv) is employed with 5-fold cross-validation (cv=5), evaluating model performance based on accuracy (scoring="accuracy"). The process uses all available CPU cores (n_jobs=-1) for parallel processing and provides verbose output (verbose=True) to track progress.						
Logistic	LOGISTIC REGRESSION HYPER PARAMETER TUNNING						
Regression	<pre>[54] #finding the grid search cv for logistic regression lr=logisticRegression(n_jobs=-1,random_state=0) lr_param_grid={     'c':[0.1,0.5,1,5,10],     'solver':['liblinear','saga'],     'penalty':['l1','l2'] } lr_cv=GridSearchCV(lr,lr_param_grid,cv=5,scoring="accuracy",n_jobs=-1,verbose=T lr_cv.fit(x_train,y_train)</pre> Fitting 5 folds for each of 20 candidates, totalling 100 fits						
	/usr/local/lib/python3.10/dist-packages/sklearn/linear_model/_logistic.py:1211:     warnings.warn(						





The parameter grid (rfc\_param\_grid) for hyperparameter tuning. It specifies different values for the number of trees (n\_estimators), splitting criterion (criterion), maximum depth of trees (max\_depth), and maximum number of features considered for splitting (max\_features). GridSearchCV (rfc\_cv) is employed with 3-fold cross-validation (cv=3), evaluating model performance based on accuracy (scoring="accuracy").

#### Random Forest

```
RANDOM FOREST HYPER PARAMETER TUNNING
[55] #finding the grid search cv for random forest classifier
     rfc=RandomForestClassifier()
     rfc_param_grid={
          'n_estimators':[100,200],
         'criterion':['entropy','gini'],
          'max_depth':[5,10],
         'max_features':['auto','sqrt']
     rfc_cv=GridSearchCV(rfc,rfc_param_grid,cv=3,scoring="accuracy",n_jobs=-1,verbose=3)
     rfc cv.fit(x train,v train)
⊋ Fitting 3 folds for each of 16 candidates, totalling 48 fits
     /usr/local/lib/python3.10/dist-packages/sklearn/ensemble/_forest.py:424: FutureWarning:
      warn(
                  GridSearchCV
      > estimator: RandomForestClassifier
           ▶ RandomForestClassifier
```

The (params) define a grid for hyperparameter tuning of the XGBoost Classifier (XGBClassifier), including min\_child\_weight, gamma, colsample\_bytree, and max\_depth. The XGBClassifier is configured with a learning rate of 0.5, 100 estimators, using a binary logistic regression objective, and utilizing 3 threads for processing. GridSearchCV (xg\_cv) is used with 5-fold cross-validation (cv=5), refitting the best model (refit=True), evaluating based on accuracy (scoring="accuracy")

#### **XGBoost**

```
#finding the grid search cv for xgboost

params={
    'min_child_weight':[10,20],
    'gamma':[1.5,2.0,2.5],
    'colsample_bytree':[0.6,0.8,0.9],
    'max_depth':[4,5,6]
}

xg_XGBClassifier(learning_rate=0.5,n_estimators=100,objective='binary:logistic',nthreads=3)
    xg_cv=GridSearchCv(xg,param_grid=params,cv=5,refit=True,scoring="accuracy",n_jobs=-1,verbose=3)
    xg_cv.fit(x_train,y_train)

Fitting 5 folds for each of 54 candidates, totalling 270 fits
    //usr/local/lib/python3.10/dist-packages/xgboost/core.py:160: UserWarning: [14:07:26] WARNING: /works
Parameters: { "nthreads" } are not used.

warnings.warn(smsg, UserWarning)

GridSearchCV

• estimator: XGBClassifier

• XGBClassifier
```

XGBOOST CLASSIFIER-HYPER PARAMETER TUNNIG





The parameters (params) define a grid for hyperparameter tuning of the Decision Tree Classifier (DecisionTreeClassifier), including max\_depth, min\_samples\_leaf, and criterion ('gini' or 'entropy'). GridSearchCV (dec\_cv) is used with 5-fold cross-validation (cv=5), evaluating model performance based on accuracy (scoring="accuracy") DECISION TREE CLASSIFIER-HYPER PARAMETER TUNNING **Decision Tree** [68] #finding grid search cv for decision tree classifier dec=DecisionTreeClassifier(random\_state=42) params={ 'max\_depth': [2, 3, 5, 10, 20],
 'min\_samples\_leaf': [5, 10, 20, 50, 100],
 'criterion': ['gini', 'entropy'] dec\_cv=GridSearchCV(dec,param\_grid=params,cv=5,n\_jobs=-1,scoring="accuracy",verbose=3) dec cv.fit(x train,y train) Fitting 5 folds for each of 50 candidates, totalling 250 fits GridSearchCV estimator: DecisionTreeClassifier ▶ DecisionTreeClassifier The parameters (params) define a grid for hyperparameter tuning of the Decision Tree Classifier (DecisionTreeClassifier), including max\_depth, min\_samples\_leaf, and criterion ('gini' or 'entropy'). GridSearchCV (dec\_cv) is used with 5-fold cross-validation (cv=5), evaluating model performance based on accuracy (scoring="accuracy") RIDGE-CLASSIFIER-HYPER PARAMETER TUNNING Ridge Classifier #finding the grid search cv for ridge classifier rg=RidgeClassifier(random\_state=42) params={ 'alpha':(np.logspace(-8,8,100)) rg\_cv=GridSearchCV(rg,param\_grid=params,cv=5) rg\_cv.fit(x\_train,y\_train)  $\overline{\Rightarrow}$ GridSearchCV ▶ estimator: RidgeClassifier ▶ RidgeClassifier





The parameters (params) define a grid for hyperparameter tuning of the K-Nearest Neighbors Classifier (KNeighbors Classifier), including n\_neighbors, weights ('uniform' or 'distance'), and metric ('minkowski', 'euclidean', or 'manhattan'). GridSearchCV (knn\_cv) is used with 5-fold cross-validation (cv=5), evaluating model performance based on accuracy (scoring="accuracy") K-NEAREST NEIGHBORS-HYPER PARAMETER TUNNING K- Nearest [69] #finding the grid search cv for k-nearest neighbors knn=KNeighborsClassifier() Neighbors 'n\_neighbors':[3,5,7,9,11], 'weights':['uniform','distance'],
'metric':['minkowski','eculidean','manhattan'] knn\_cv = GridSearchCV(knn, param\_grid=params,cv=5, n\_jobs=-1, verbose=3) knn\_cv.fit(x\_train, y\_train) GridSearchCV estimator: KNeighborsClassifier ▶ KNeighborsClassifier

## Final Model Selection Justification (2 Marks):





Final Model	Reasoning							
	com		nodel is chosen for its robustness in handling and its ability to mitigate overfitting while providing couracy.					
		Name	Accuracy	f1_score	Recall	Precision		
	0	Logistic Regression	67.90	64.68	59.16	71.35		
Random Forest	1	Decision Tree Classifier	73.88	66.60	52.41	91.32		
	2	Random Forest	74.68	66.70	51.03	96.24		
	3	K-Nearest Nieghbors	74.56	71.57	64.44	80.48		
	4	Xgboost	74.18	68.61	56.78	86.67		
	5	Ridge Classifier	68.39	63.91	56.32	73.87		
	Above all the models Random Forest model have the highest accuracy among all the models.							