# PROJECT REPORT BURN SMART

(Predicting Calorie Burn During Workouts)

UNDER GUIDANCE OF

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### PROJECT OVERVIEW

Introduction

Our project aims to develop machine learning models that can accurately predict calorie burn during workouts.

By taking into account various factors such as heart rate, duration, body temperature, and individual characteristics like age, weight, height, and gender, we hope to provide users with a more personalized and accurate estimate of their calorie burn.



**Project** 

Project Goals



Develop

Develop machine learning models that accurately predict calorie burn during workouts.



Incorporate

Incorporate factors such as heart rate, duration, body temperature, and individual characteristics like age, weight, height, and gender into the models. Compare the performance of different machine learning models.



**Optimize** 

Optimize the best performing model to improve accuracy. Present the results and discuss future work.

## FACTORS AFFECTING

#### Heart Rate

The higher the heart rate during a workout, the more calories burned.

#### Duration

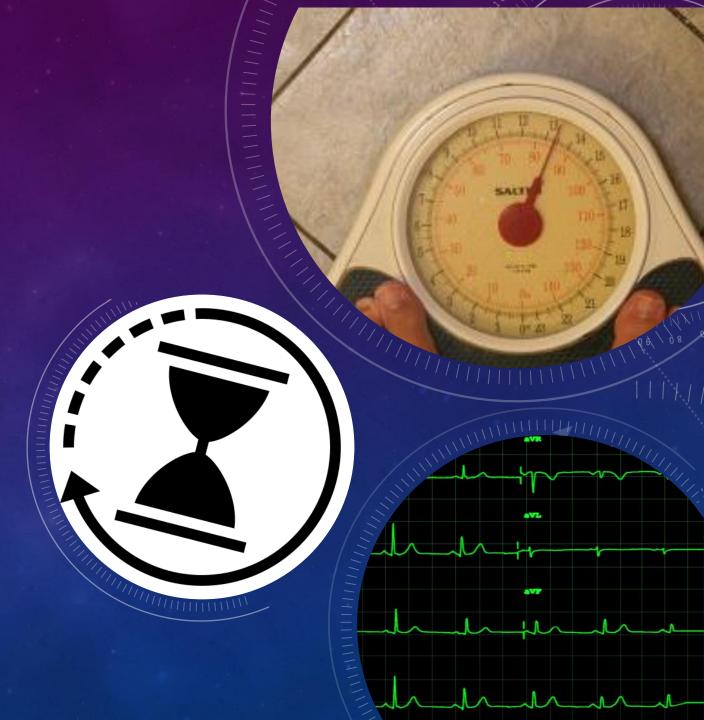
The longer the duration of a workout, the more calories burned.

### Body Temperature

Higher body temperature may increase calorie burn during a workout.

### Individual Characteristics

Age, weight, height, and gender can all impact calorie burn during a workout.



## Predicting Calorie Burn: A Machine Learning Approach



Machine learning Models: Utilizing machine learning algorithms to predict calorie burn.



Data Collection and Analysis: Collecting and analyzing data on physical activity and calorie burn.



Model Training and Evaluation



Training and evaluating the machine learning models for calorie burn prediction



Wearable Devices

Utilizing wearable devices to track physical activity and collect data.

**Mobile Applications** 

Developing mobile apps to record and analyze physical activity.

Sensor Technology

Exploring sensor technology for accurate measurement of physical activity.



## FEATURE SELECTION AND DATA PREPROCESSING

• Feature Importance

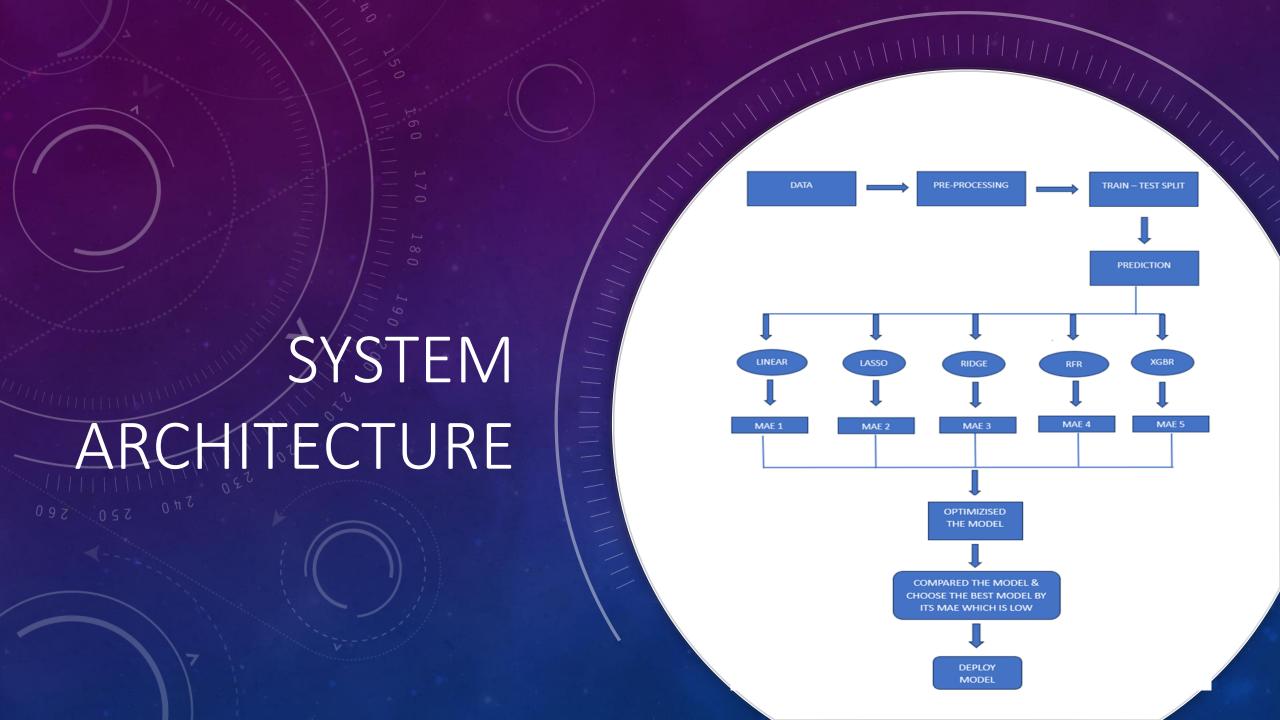
Identifying the most influential features for calorie burn prediction.

Handling Missing Data

Dealing with missing data through imputation or elimination.

• Normalization

Applying normalization techniques to preprocess the data.





## Machine Learning Models for Predicting Calorie Burn:

### **Regression Models:**

Using regression models to predict calorie burn based on various factors.

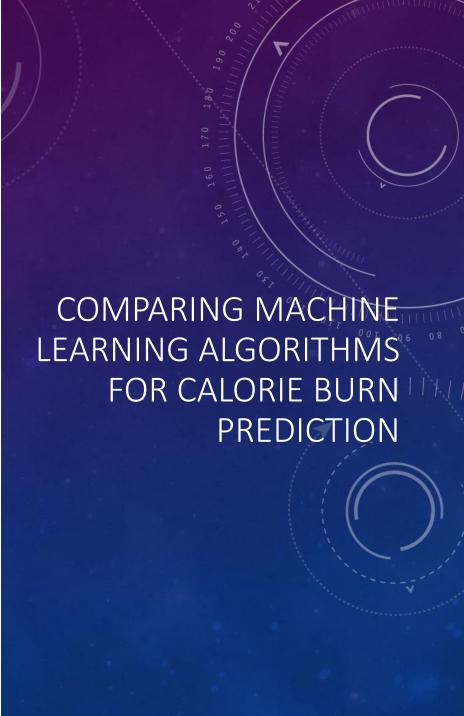
- LINEAR REGRESSION
- LASSO REGRESSION
- RIDGE REGRESSION

### **Ensemble Models:**

Employing ensemble models to improve the accuracy of calorie burn prediction.

- RANDOM FOREST REGRESSOR
- XGB REGRESSOR

```
LinearRegression() :
Training Error: 8.332985229896742
Validation Error: 8.385188053147179
XGBRegressor(base_score=None, booster=None, callbacks=None,
            colsample_bylevel=None, colsample_bynode=None,
            colsample bytree=None, device=None, early stopping rounds=None,
            enable categorical=False, eval metric=None, feature types=None,
            gamma=None, grow policy=None, importance type=None,
            interaction_constraints=None, learning_rate=None, max_bin=None,
            max cat threshold=None, max cat to onehot=None,
            max_delta_step=None, max_depth=None, max_leaves=None,
            min child weight=None, missing=nan, monotone constraints=None,
            multi_strategy=None, n_estimators=None, n_jobs=None,
            num_parallel_tree=None, random_state=None, ...):
Training Error: 0.9322033420062313
Validation Error: 1.4833678883314132
Lasso():
Training Error: 9.049658999072935
Validation Error: 8.989469141792506
RandomForestRegressor() :
Validation Error: 1.6871166666666664
Ridge() :
Training Error: 8.332695704014442
Validation Error: 8.38482196600372
```



## MODEL OPTIMIZATION

HYPERPARAMATER TUNING



Optimizing model performance by tuning hyperparameters.

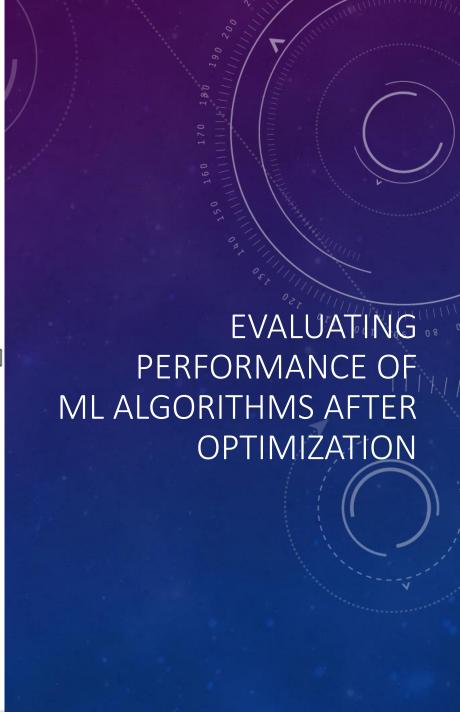


**VALIDATE MODEL** 



Applying validation to assess model performance and generalization.

```
LinearRegression():
Training Error: 8.306790197742497
Validation Error: 8.441513553849703
Ridge(alpha=10):
Training Error: 8.304994376664778
Validation Error: 8.438993797246134
Lasso(alpha=0.05):
Training Error: 8.302465721352533
Validation Error: 8.442647976630964
RandomForestRegressor() :
Training Error: 0.6698925
Validation Error: 1.7103733333333335
/usr/local/lib/python3.10/dist-packages/xgboost/core.py:160: UserWarning: [16:26:49]
 warnings.warn(smsg, UserWarning)
XGBRegressor(base score=None, booster=None, callbacks=None,
            colsample bylevel=None, colsample bynode=None,
            colsample_bytree=0.7, device=None, early_stopping_rounds=None,
            enable_categorical=False, eval_metric=None, feature_types=None,
            gamma=None, grow_policy=None, importance_type=None,
            interaction constraints=None, learning rate=0.07, max bin=None,
            max cat threshold=None, max cat to onehot=None,
            max_delta_step=None, max_depth=5, max_leaves=None,
            min_child_weight=4, missing=nan, monotone_constraints=None,
            multi strategy=None, n estimators=500, n jobs=None, nthread=4,
            num parallel tree=None, ...):
Training Error: 0.7825378867114584
Validation Error: 1.005842310667038
```



### BEST ALGORITHM



XGB REGRESSOR



Achieving the lowest MAE and highest accuracy for calorie burn prediction.



**Model Deployment** 



Implementing the best performing algorithm in practical applications.



After testing multiple machine learning models, we found that the XGB REGRESSOR performed the best in predicting calorie burn during workouts. This model had an MAE of 1.0005, indicating a strong correlation between the predicted and actual calorie burn values.

## **Calories burn prediction**

Enter Your Gender as (Female:0, Male:1) 0 Enter Your Age: 15 Enter Your Height (in cm): 100 Enter Your Weight (in kg): 40 Enter Your duration (in mins): Enter Your Heart Rate from 60 to 130: 60 Enter Your Body temp from 36 to 42 (in celsius): 36

Calories burned

WEB APPLICATION

Developing a user-friendly web interface to access predictive models.

### CONCLUSION



We have proposed a system for predicting the number of calories burned during physical activity that takes into account all of the relevant variables, including age, gender, height, weight, body temperature, heart rate, and the length of the activity. We have compared the accuracy of various regression models, including Random Forest, Ridge Regression, Lasso Regression, and XGBoost. Our system achieves the highest accuracy by using a combination of these models.



In addition to being accurate, our system is also user-friendly. We have developed a webpage for our system that allows users to easily make predictions, receive recommendations, and personalize their experience. The webpage also includes anomaly detection capabilities to identify unusual activity patterns.

## THANK YOU

Take control of your workout and optimize your burning calories!