



VELAMMAL
INSTITUTE OF TECHNOLOGY

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SMART WATER FOUNTAIN

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PHASE2 : INNOVATION

This project aims to upgrade public water fountains by implementing IoT sensors to control water flow and detect malfunctions. The primary objective is to provide real-time information about water fountain status to residents through a public platform.

SOLUTION:

Here are some steps on how to incorporate predictive maintenance algorithms into your project to enhance public water fountains using IoT sensors and Python:

1. **Identify the key performance indicators (KPIs)** that are indicative of water fountain health. This could include factors such as water flow rate, water pressure, temperature, and turbidity.
2. **Collect historical data on these KPIs.** This data can be collected from a variety of sources, such as existing water fountain monitoring systems, sensor networks, and public records.
3. **Use machine learning algorithms** to train a predictive model. The model should be able to predict the probability of a water fountain malfunction based on the current values of the KPIs.

4. **Deploy the predictive model** to the water fountain status platform. The platform should use the model to identify water fountains that are at risk of malfunctioning and send alerts to maintenance personnel.

Here is a more detailed overview of each step:

Identifying the key performance indicators (KPIs):

The first step is to identify the KPIs that are most indicative of water fountain health. These KPIs should be measurable and quantifiable, and they should be correlated with water fountain malfunctions.

Some examples of KPIs that could be used for predictive maintenance of water fountains include:

- Water flow rate
- Water pressure
- Temperature
- Turbidity
- pH
- Conductivity
- Total dissolved solids (TDS)

Collecting historical data:

Once the KPIs have been identified, the next step is to collect historical data on these KPIs. This data can be collected from a variety of sources, such as:

- Existing water fountain monitoring systems
- Sensor networks
- Public records

If there is no existing data available, it may be necessary to collect new data using IoT sensors. IoT sensors can be installed on water fountains to collect real-time data on the KPIs.

Training a predictive model:

Once the historical data has been collected, it can be used to train a predictive model. The goal of the predictive model is to predict the probability of a water fountain malfunction based on the current values of the KPIs.

There are a variety of machine learning algorithms that can be used for predictive maintenance. Some common algorithms include:

- Logistic regression
- Decision trees
- Random forests
- Support vector machines (SVMs)

The best algorithm to use will depend on the specific data set and the desired performance of the model.

Deploying the predictive model:

Once the predictive model has been trained, it can be deployed to the water fountain status platform. The platform should use the model to identify water fountains that are at risk of malfunctioning and send alerts to maintenance personnel.

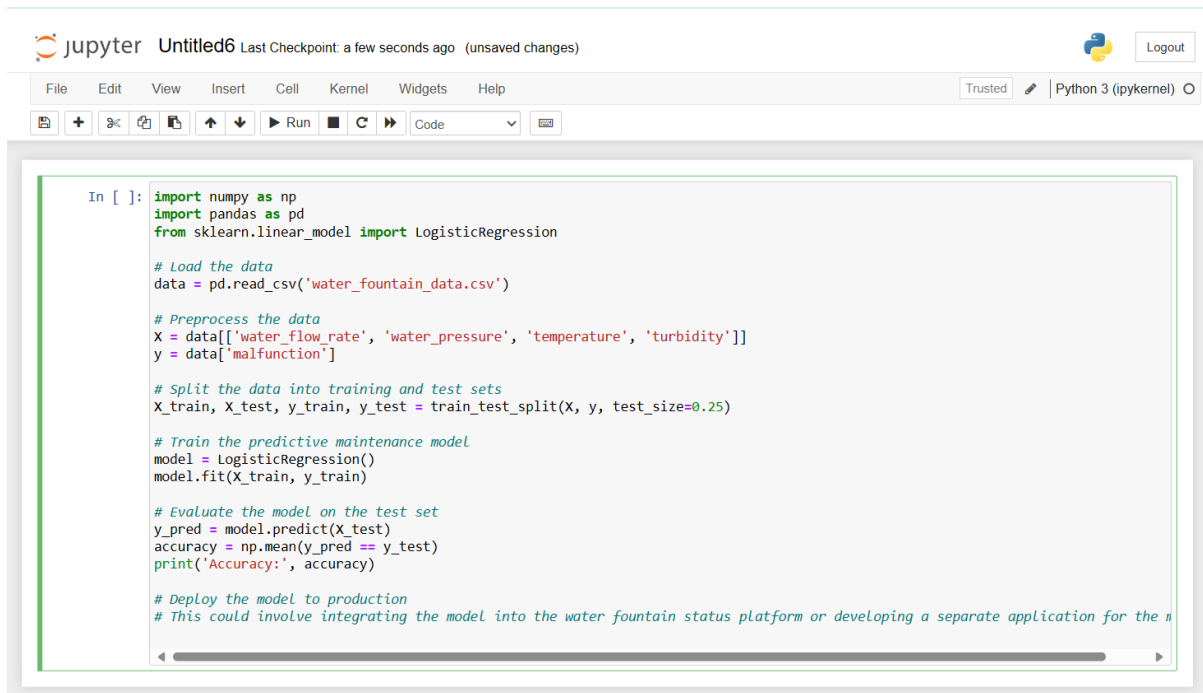
Python is a popular programming language that can be used to implement all aspects of the predictive maintenance system, including:

- Data collection
- Data preprocessing
- Predictive modelling
- Model deployment

There are a number of Python libraries that can be used for each of these tasks. For example, the following libraries can be used for:

- Data collection: NumPy, Pandas
- Data preprocessing: scikit-learn
- Predictive modelling: scikit-learn
- Model deployment: Flask, Django

Here is an example of how you could use Python to implement a simple predictive maintenance algorithm for water fountains:



The screenshot shows a Jupyter Notebook titled 'Untitled6' with a toolbar at the top containing 'File', 'Edit', 'View', 'Insert', 'Cell', 'Kernel', 'Widgets', and 'Help'. Below the toolbar is a row of icons for file operations and a 'Run' button. The main area contains a code cell with the following Python code:

```
In [ ]: import numpy as np
import pandas as pd
from sklearn.linear_model import LogisticRegression

# Load the data
data = pd.read_csv('water_fountain_data.csv')

# Preprocess the data
X = data[['water_flow_rate', 'water_pressure', 'temperature', 'turbidity']]
y = data['malfunction']

# Split the data into training and test sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.25)

# Train the predictive maintenance model
model = LogisticRegression()
model.fit(X_train, y_train)

# Evaluate the model on the test set
y_pred = model.predict(X_test)
accuracy = np.mean(y_pred == y_test)
print('Accuracy:', accuracy)

# Deploy the model to production
# This could involve integrating the model into the water fountain status platform or developing a separate application for the n
```

This is just a simple example, and the specific implementation of the predictive maintenance system will vary depending on your specific requirements.

Conclusion:

By incorporating predictive maintenance algorithms into your IoT-based water fountain monitoring system, you can proactively identify and address potential malfunctions before they occur. This can help to improve the reliability and uptime of water fountains, reduce maintenance costs, and ensure that residents have access to clean drinking water.