

Predictors for Trickling in WSN

INE5424 Operating Systems II

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Outline

Motivation & Objective

Project Planning

- Project Plan

- Changes & Issues

- Design

Development

- Linear Regression Predictor with Gradient Descent

- MLP Predictor

- Comparison

Conclusion & Future Work

Demo

Implementation

- Predictive Smart Data

- Predictors

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Motivation &
Objective

Project Planning

Project Plan
Changes & Issues
Design

Development

Linear Regression
Predictor with
Gradient Descent
MLP Predictor
Comparison

Conclusion &
Future Work

Demo

Implementation

Predictive Smart
Data
Predictors

Motivation & Objective

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Motivation

- ▶ IoT environment with numerous devices
- ▶ Power consumption is an issue (devices' batteries)
- ▶ Communication between sensors consumes a lot of power

Objective

- ▶ Improvement of energy efficiency by reducing communication
- ▶ Development of **SmartData** predictors

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Project Planning

Project Plan
Changes & Issues
Design

Development

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Comparison

Conclusion &
Future Work

Demo

Implementation

Predictive Smart
Data
Predictors

Motivation & Objective

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Objective

Project Planning

Project Plan
Changes & Issues
Design

Development

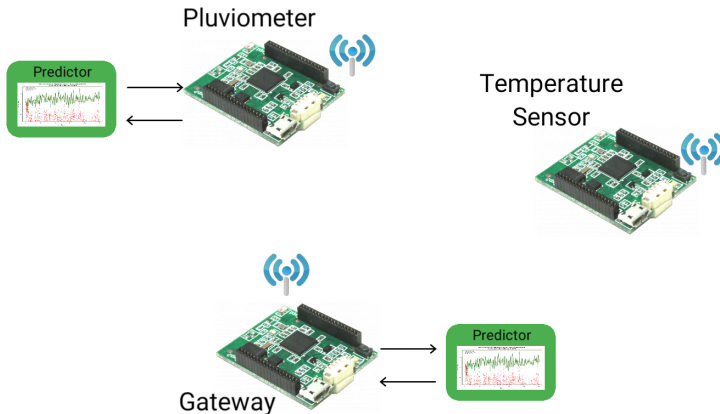
Linear Regression
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Gradient Descent
MLP Predictor
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Conclusion &
Future Work

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Predictive Smart
Data
Predictors



Project Plan

Tasks

1. Study system restrictions and limitations of EPOSMote III and elaborate implementation strategies, such as design patterns ✓
2. Code and test the chosen prediction algorithms in C++ for a general purpose operating system ✓
3. Code and test the linear regression with gradient descent predictor on EPOS ✓
4. Code and test the Elman neural network on EPOS¹
5. Deploy the final product using the UFSC IoT Gateway¹

¹See project changes

Changes & Issues

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Changes

- ▶ Elman Neural Network to Multilayer Perceptron Network
- ▶ Deploy the final product using the UFSC IoT Gateway to test with data from UFSC IoT

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Objective

Project Planning

Project Plan
Changes & Issues
Design

Development

Linear Regression
Predictor with
Gradient Descent
MLP Predictor
Comparison

Conclusion &
Future Work

Demo

Implementation

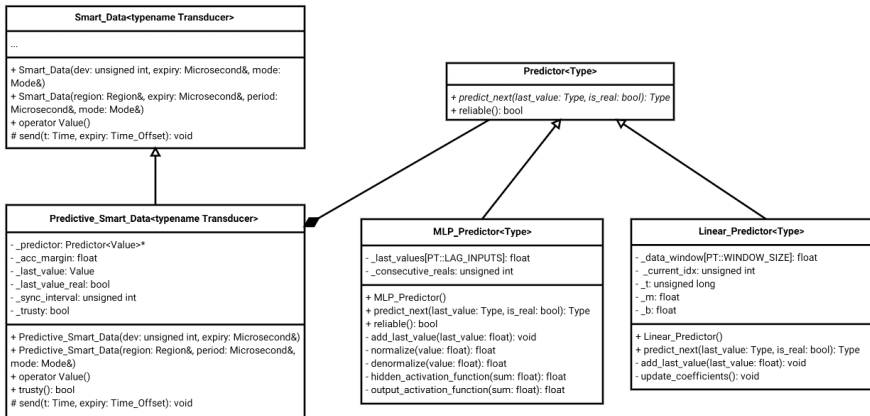
Predictive Smart
Data
Predictors

Issues

- ▶ Synchronization of predictors
- ▶ Implementation of exponential function

Project Design

Class Diagram



Project Design

Design Patterns

Design Patterns

- ▶ Bridge
- ▶ Composition

Programming Techniques

- ▶ Generic programming
- ▶ Static metaprogramming

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Project Planning

Project Plan
Changes & Issues
Design

Development

Linear Regression
Predictor with
Gradient Descent
MLP Predictor
Comparison

Conclusion &
Future Work

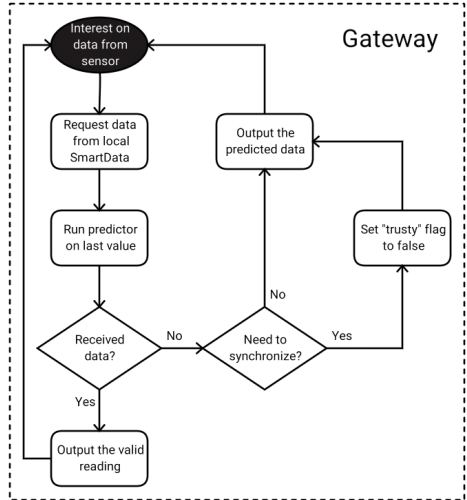
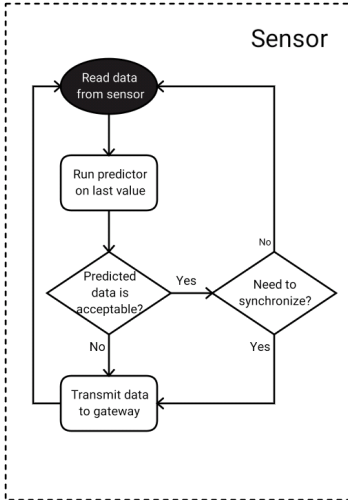
Demo

Implementation

Predictive Smart
Data
Predictors

Project Design

Sensor and Gateway Operation



Linear Regression Predictor with Gradient Descent

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Motivation &
Objective

Project Planning

Project Plan
Changes & Issues
Design

Development

**Linear Regression
Predictor with
Gradient Descent**
MLP Predictor
Comparison

Conclusion &
Future Work

Demo

Implementation

Predictive Smart
Data
Predictors

- ▶ Performance tests and simulation in C++
- ▶ Static parameterization via Traits.h on EPOS
- ▶ Generic type prediction

Linear Regression Predictor

Performance Tests

**Prediction of Hourly Air Temperature at the San Francisco International Airport
from January 1st, 2015 to December 31st, 2016**



MLP Predictor

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Motivation &
Objective

Project Planning

Project Plan
Changes & Issues
Design

Development

Linear Regression
Predictor with
Gradient Descent
MLP Predictor
Comparison

Conclusion &
Future Work

Demo

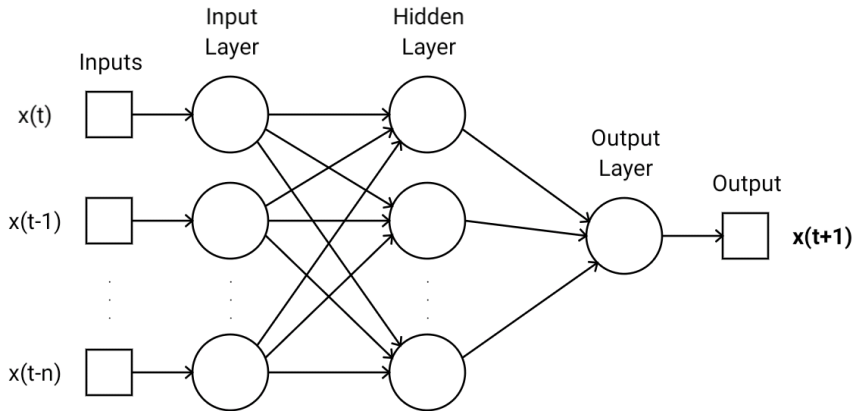
Implementation

Predictive Smart
Data
Predictors

- ▶ Prediction of time series $x(t)$ based on n last values of $x(t)$.
- ▶ Requires historical data of the series to be predicted.
- ▶ Neural network training on a third-party software².

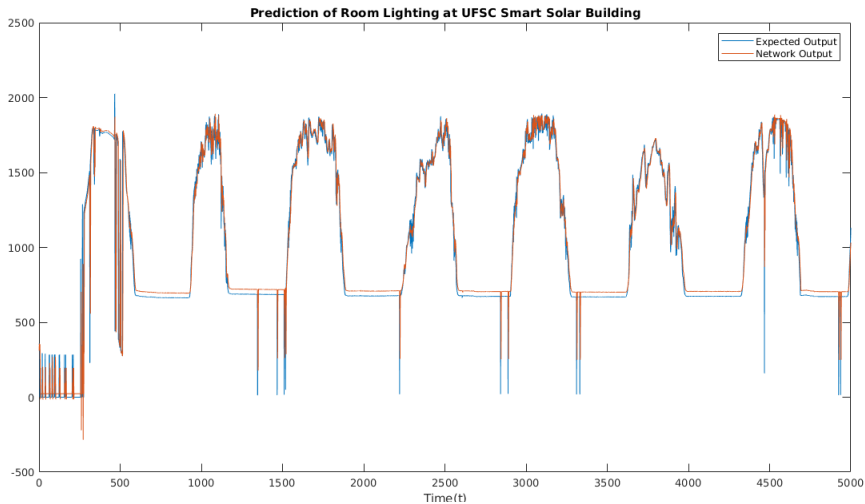
²MATLAB - MathWorks

Multilayer Perceptron Network



MLP Predictor

Performance Tests



Linear Regression vs. MLP Predictor

Linear Regression Predictor

**Prediction of Hourly Air Temperature at the San Francisco International Airport
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Linear Regression vs. MLP Predictor

MLP Predictor

Prediction of Hourly Air Temperature at the San Francisco International Airport
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Linear Regression vs. MLP Predictor

Overall Comparison

Feature	Linear Predictor	MLP Predictor
Type	Dynamic	Static
Online learning	✓	
Needs series historical data		✓
Resilient to desynchronization		✓
No preprocessing needed	✓	

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Objective

Project Planning

Project Plan
Changes & Issues
Design

Development

Linear Regression
Predictor with
Gradient Descent
MLP Predictor
Comparison

Conclusion &
Future Work

Demo

Implementation

Predictive Smart
Data
Predictors

Conclusion & Future Work

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Motivation &
Objective

Project Planning

Project Plan
Changes & Issues
Design

Development

Linear Regression
Predictor with
Gradient Descent
MLP Predictor
Comparison

Conclusion &
Future Work

Demo

Implementation

Predictive Smart
Data
Predictors

- ▶ Project planning requirements fulfilled
- ▶ Achieved satisfying results
- ▶ Future Work
 - ▶ Develop other types of predictors
 - ▶ Expand neural network implementation

Project Demo

Project demonstration

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Project Planning

Project Plan
Changes & Issues
Design

Development

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Predictor with
Gradient Descent
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Comparison

Conclusion &
Future Work

Demo

Implementation

Predictive Smart
Data
Predictors

Predictive_Smart_Data

Traits

```
template<> template <typename S> struct Traits<
    Predictive_Smart_Data<S>>: public Traits<
    Smart_Data<S>>
{
    enum {LINEAR, MLP};

    static const bool debugged = true;
    static const unsigned int ACC_MARGIN = 8;
    static const unsigned int PREDICTOR = LINEAR;
    static const unsigned int SYNC_INTERVAL = 2;
};
```

Predictive_Smart_Data

Predictor Selection

```
private:
  typedef Traits<Predictive_Smart_Data<Transducer>> PT;
  typedef typename IF<(PT::PREDICTOR == PT::LINEAR),
    Linear_Predictor<Value>,
    MLP_Predictor<Value>>::Result P_Type;
```

Predictive_Smart_Data

send()

```
void send(const Time t, Time_Offset expiry) {
    Value predicted = _predictor->predict_next(
        _last_value, _last_value_real);
    Value real = Smart_Data<Transducer>::_value;

    bool acceptable_margins, check_sync = ...

    if(acceptable_margins && check_sync) {
        _last_value = predicted;
        _last_value_real = false;
        _sync_interval--;
    } else {
        _last_value = real;
        _last_value_real = true;
        Smart_Data<Transducer>::send(t, expiry);
        _sync_interval = PT::SYNC_INTERVAL;
    }
}
```

Predictive_Smart_Data

operator Value()

```
operator Value() {
    Value predicted;
    bool device_remote, sensor_side = ...
    if(device_remote)
        predicted = _predictor->predict_next(
            _last_value, _last_value_real);
    if(Smart_Data<Transducer>::expired()) {
        if(sensor_side) {
            Transducer::sense(
                Smart_Data<Transducer>::_device, this);
            Smart_Data<Transducer>::_time = TSTP::now();
        } else {
            _last_value = predicted;
            _last_value_real = false;
            if(PT::SYNC_INTERVAL && !_sync_interval)
                _trusty = false;
            if(_sync_interval > 0)
                _sync_interval--;
        }
    }
    ...
}
```

Predictive_Smart_Data

operator Value()

```
...
else {
    _last_value = Smart_Data<Transducer>::_value;
    _last_value_real = true;
    if(device_remote)
        _sync_interval = PT::SYNC_INTERVAL;
}

if(!_trusty)
    _trusty = _predictor->reliable();
return _last_value;
}
```


Predictor

Predictor Interface

```
template<typename Type>
class Predictor
{
public:
    virtual Type predict_next(
        Type last_value, bool is_real = false) = 0;
    virtual bool reliable() { return false; };
};
```

```
template<typename Type>
class Linear_Predictor: public Predictor<Type>
{
    ...
};
```

```
template<typename Type>
class MLP_Predictor: public Predictor<Type>
{
    ...
};
```

Linear_Predictor

Traits

```
template <typename S> struct Traits<Linear_Predictor<S
    >>: public Traits<void>
{
    static const bool debugged = true;
    static const unsigned int WINDOW_SIZE = 30;
    static const float LRATE;
    static const unsigned short GD_ITERATIONS = 200;
    static const unsigned short M = 0;
    static const unsigned short B = 10;
};
template <typename S> const float Traits<
    Linear_Predictor<S>>::LRATE = 0.000000001f;
```

MLP_Predictor

Traits

```
template <typename S> struct Traits<MLP_Predictor<S>>:
    public Traits<void>
{
    static const bool debugged = true;
    static const unsigned int HIDDEN_UNITS = 5;
    static const unsigned int LAG_INPUTS = 3;
    static const float HIDDEN_WEIGHTS[HIDDEN_UNITS*
        LAG_INPUTS];
    static const float HIDDEN_BIASES[HIDDEN_UNITS];
    static const float OUTPUT_WEIGHTS[HIDDEN_UNITS];
    static const float OUTPUT_BIAS;
    static const bool NORMALIZATION = true;
    static const float NORMALIZATION_MIN;
    static const float NORMALIZATION_MAX;
};
```