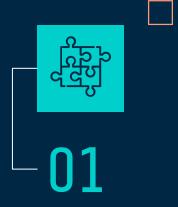
# Time Series Forecasting Library JT, Cameron, River, Noah,

Nick

## In this presentation...



**DEMO** 

A example use case for the library

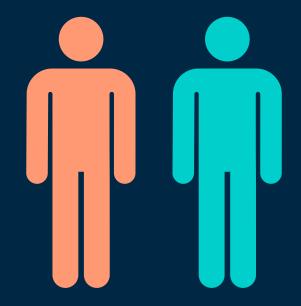




THE TEAM

How team work made the dream work

# Flexibility, ease of use, and accessibility



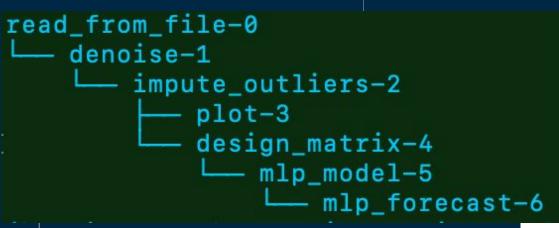
# Demo

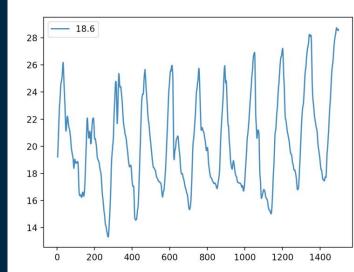
Time series forecasting made simple



```
tree = TS_Tree()
    tree.replace_node("read_from_file", 0)
    tree.add_node("denoise", 0, increment=5.0)
    tree.add_node("impute_outliers", 1)
    tree.add_node("plot", 2)
    tree.add_node("design_matrix", 2, data_start=20.0, data_end=10.0)
    tree.add_node("mlp_model", 4, layers=[100,])
    tree.add_node("mlp_forecast", 5, input_filename="../timeSeriesData/TimeSeriesData1/1_temperature_test.csv")
    tree.print_tree()
    cleaned_ts = tree.execute_path("timeSeriesData/TimeSeriesData1/1_temperature_train.csv", 3)
    modeled = tree.execute_path(None, 6)
    save tree(tree, "example tree.txt")
main()
```

from tree import \*





# Implementation

Our approach



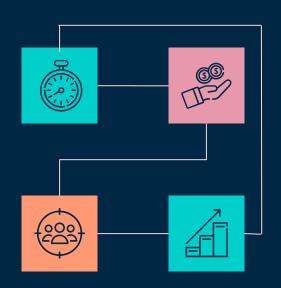
## Incremental development, Testing

#### Requirements

Starting with what is best for the user

#### **Implementation**

Modular Approach



#### **Analysis**

Planning the structures to be used

#### Testing

Nose testing framework

## Components: A Modular Approach



File\_IO

Pandas read in and write out

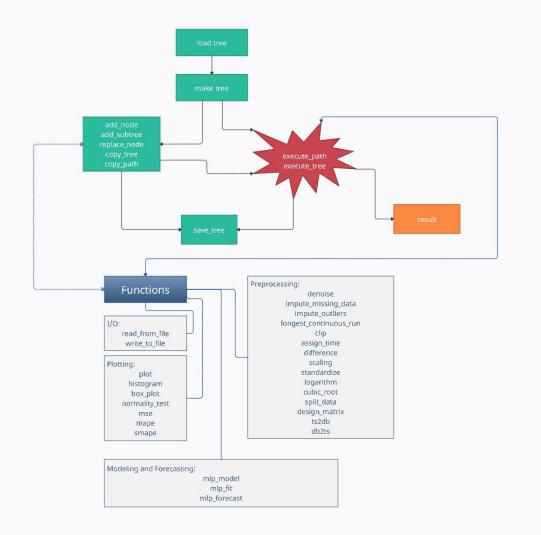






Modeling and forecasting

Straightforward modeling



# Our Team

Time series forecasting made simple



### OUR PROCESS

