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My Machine is an Honor Student at Cisco!

Quality of machine learning experience

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DEVLIT-4002

cisco Live!
Barcelona | January 27-31, 2020

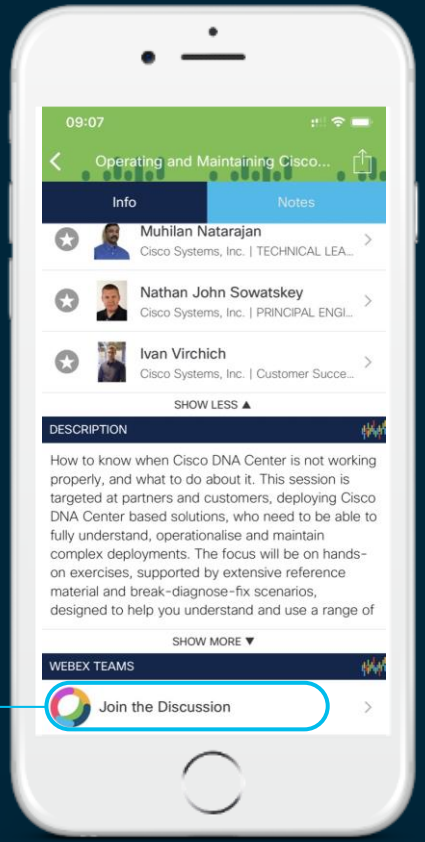
Cisco Webex Teams

Questions?

Use Cisco Webex Teams to chat with the speaker after the session

How

1. Find this session in the Cisco Events Mobile App
2. Click “Join the Discussion”
3. Install Webex Teams or go directly to the team space
4. Enter messages/questions in the team space

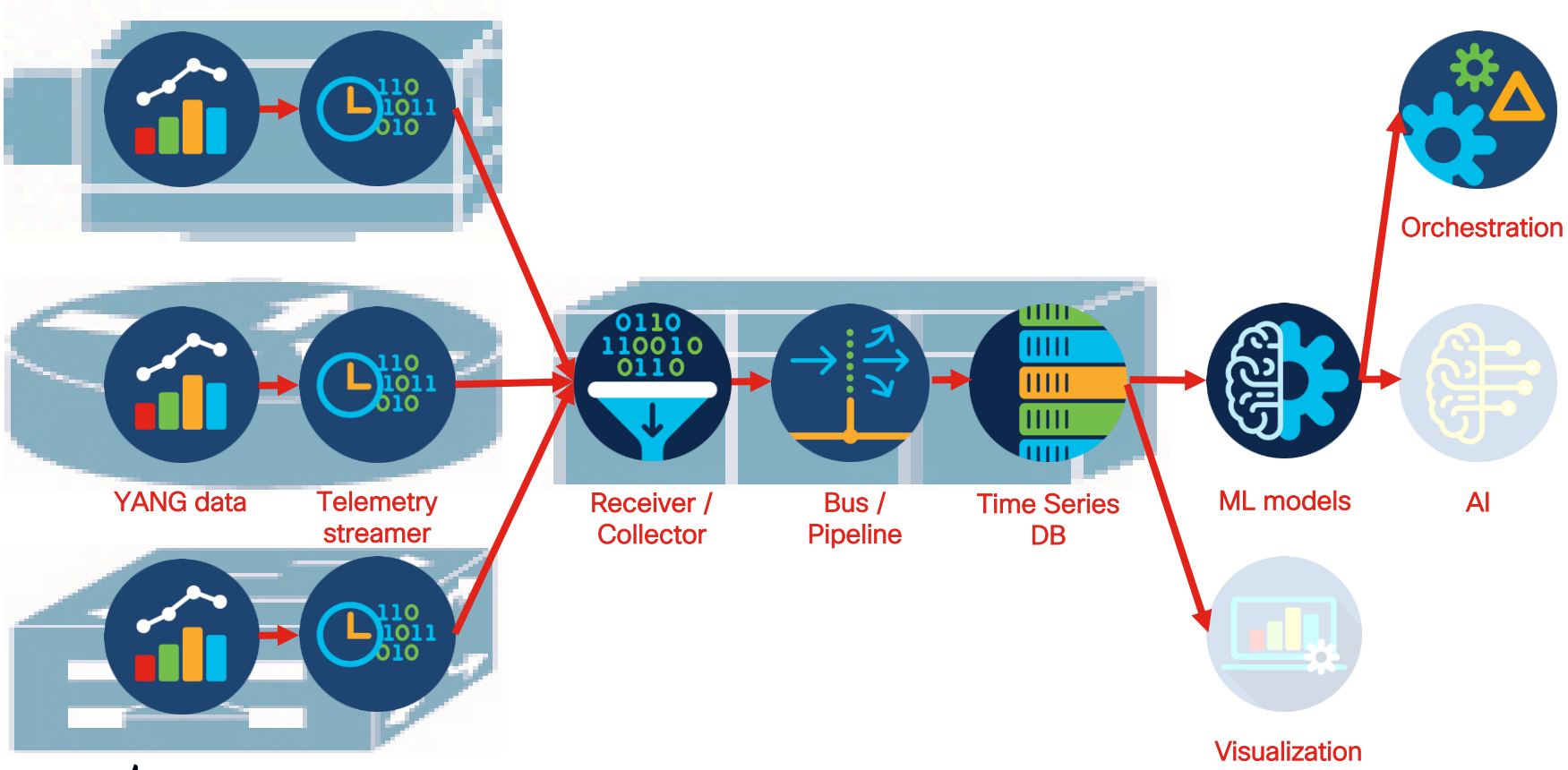


Agenda

- Data analysis and quality
- The process
- Data conditioning
- Visualization
- Quality assurance

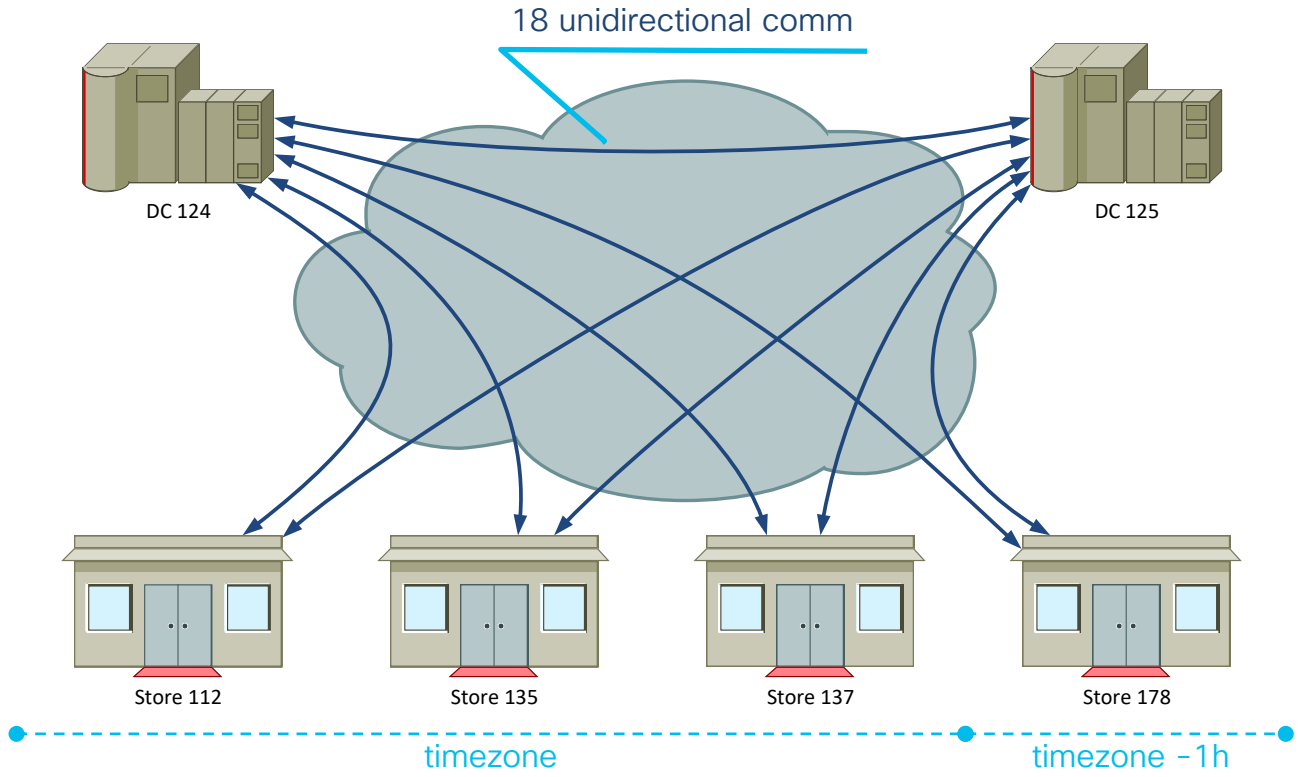
Initiate ML model, initial training...

Production Line

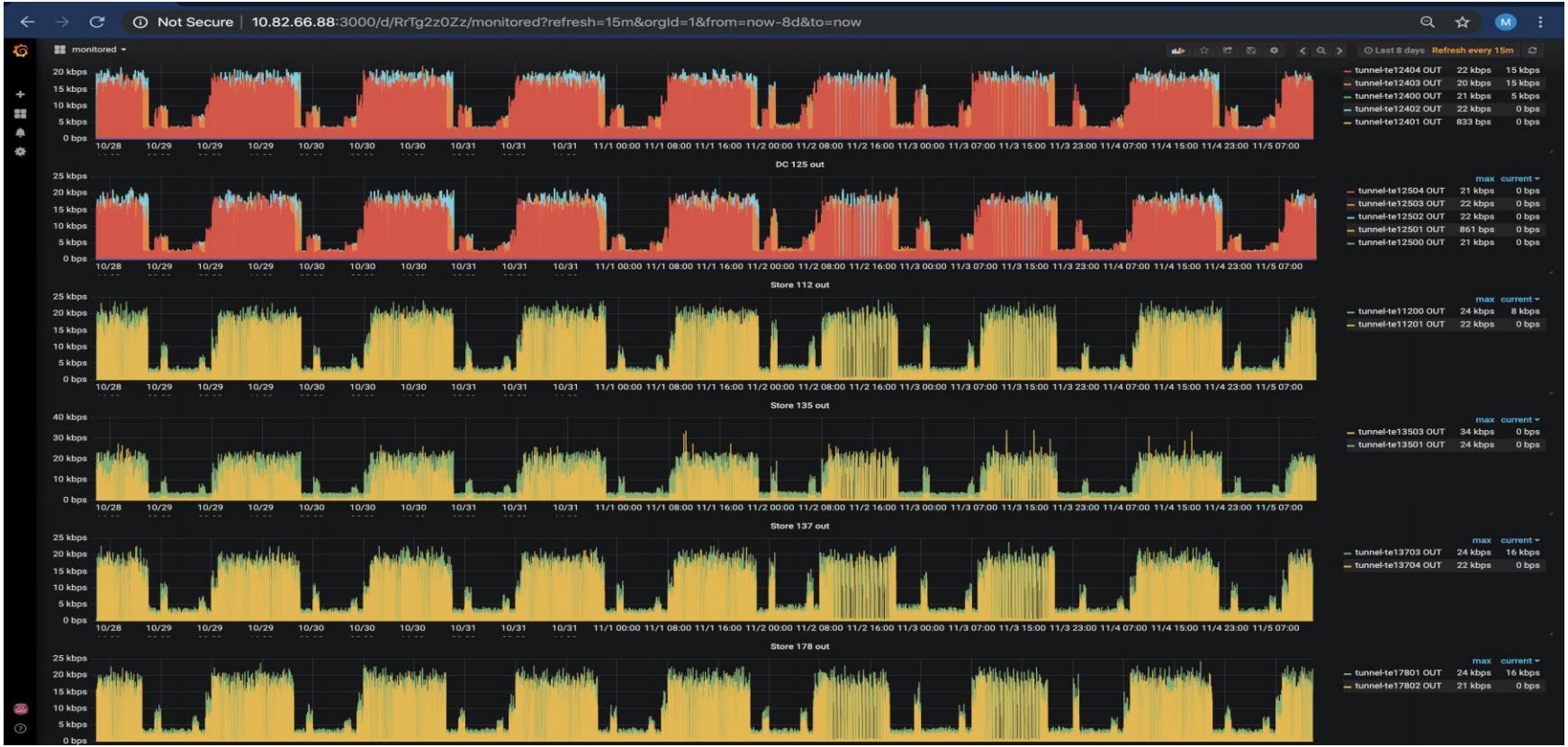


Data Source

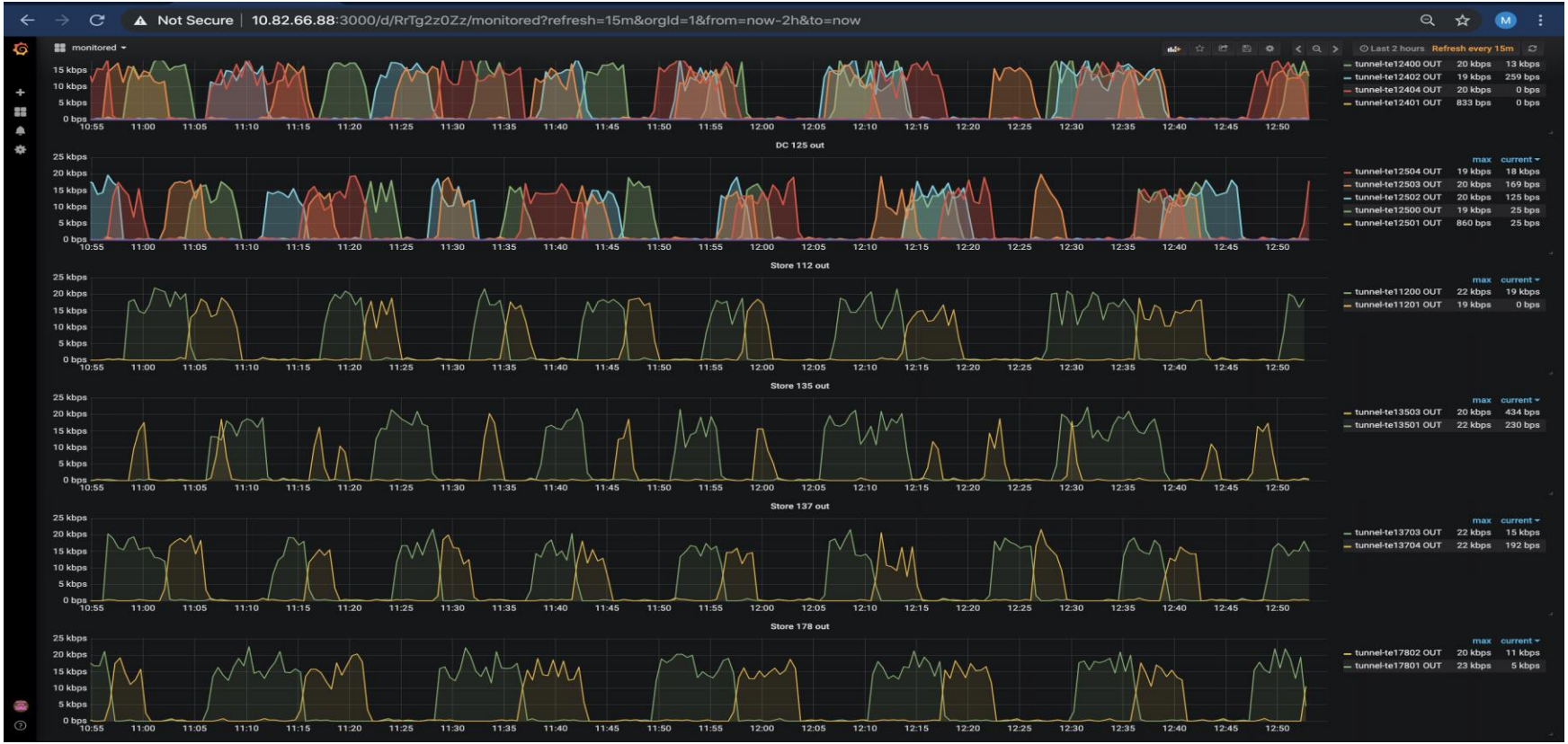
Business app data



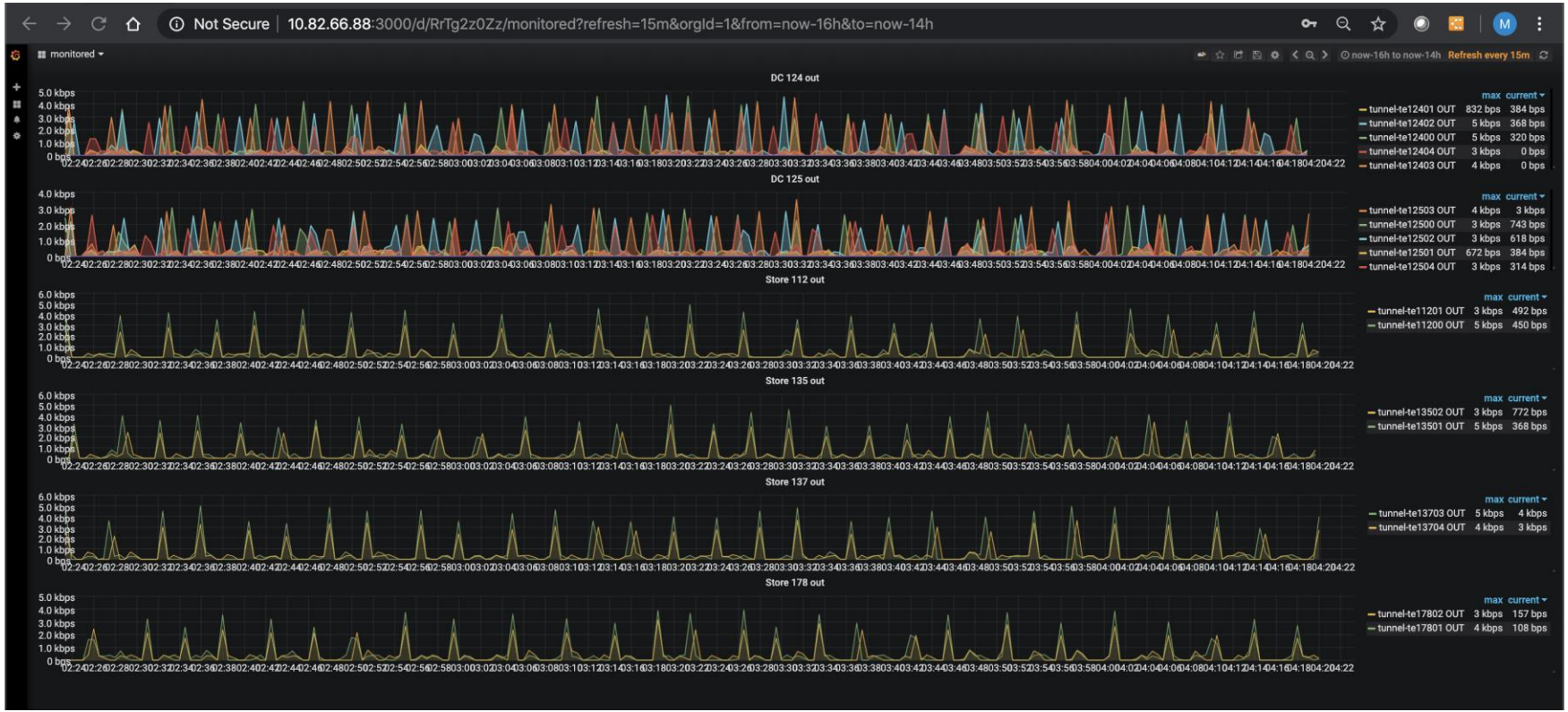
Visualized – 8days



Visualized – 2 business hours

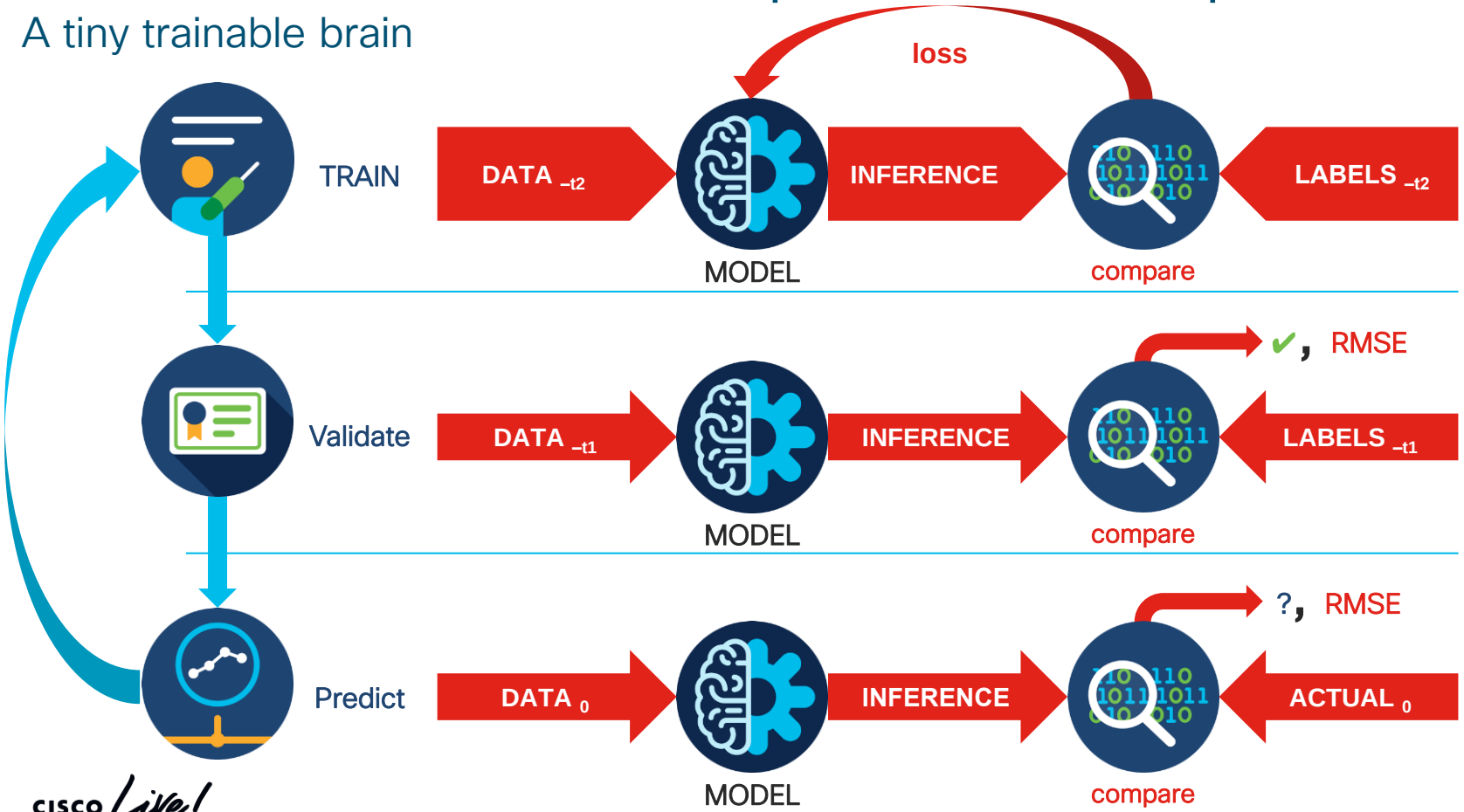


Visualized – 2 off hours



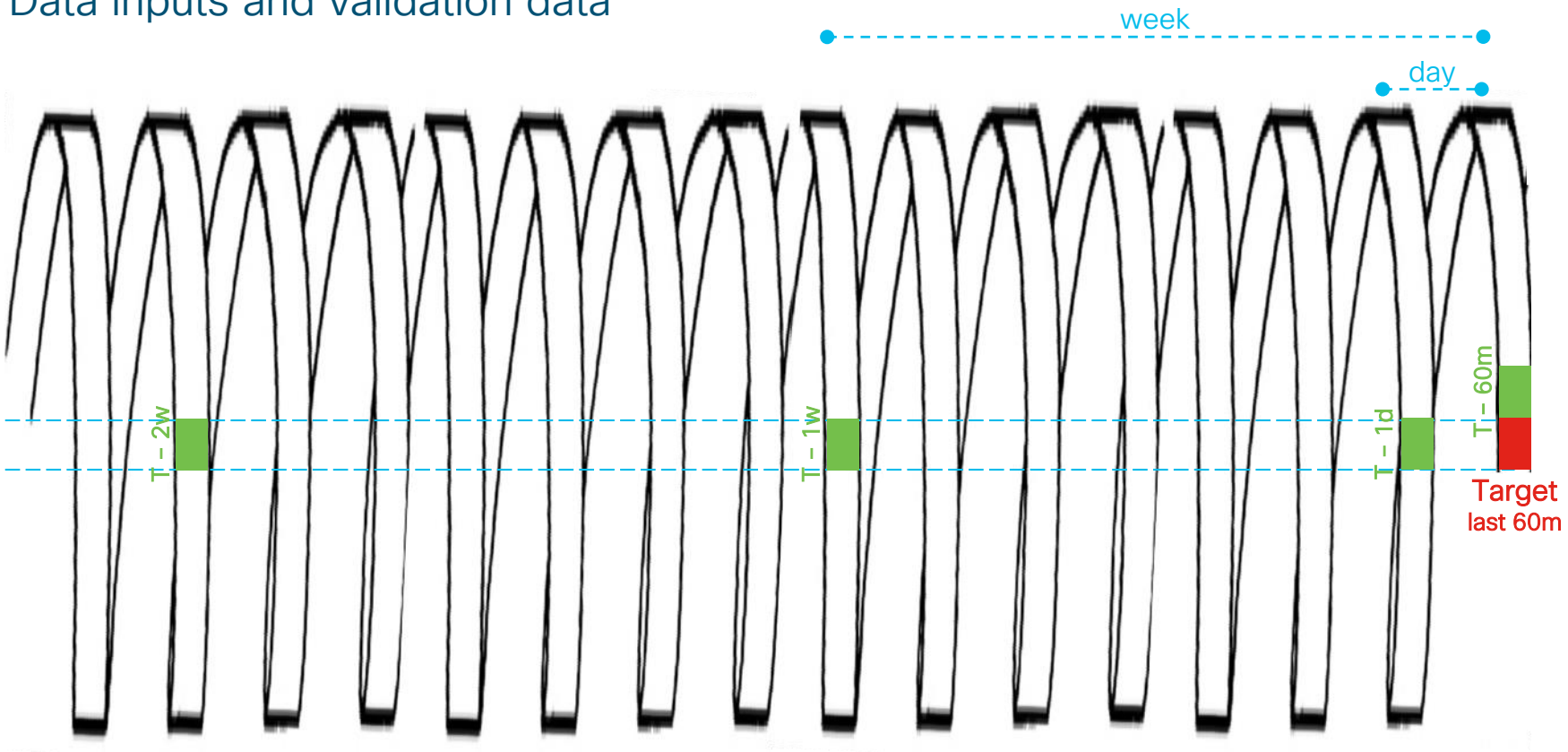
What's an ML Model? Supervised example

A tiny trainable brain



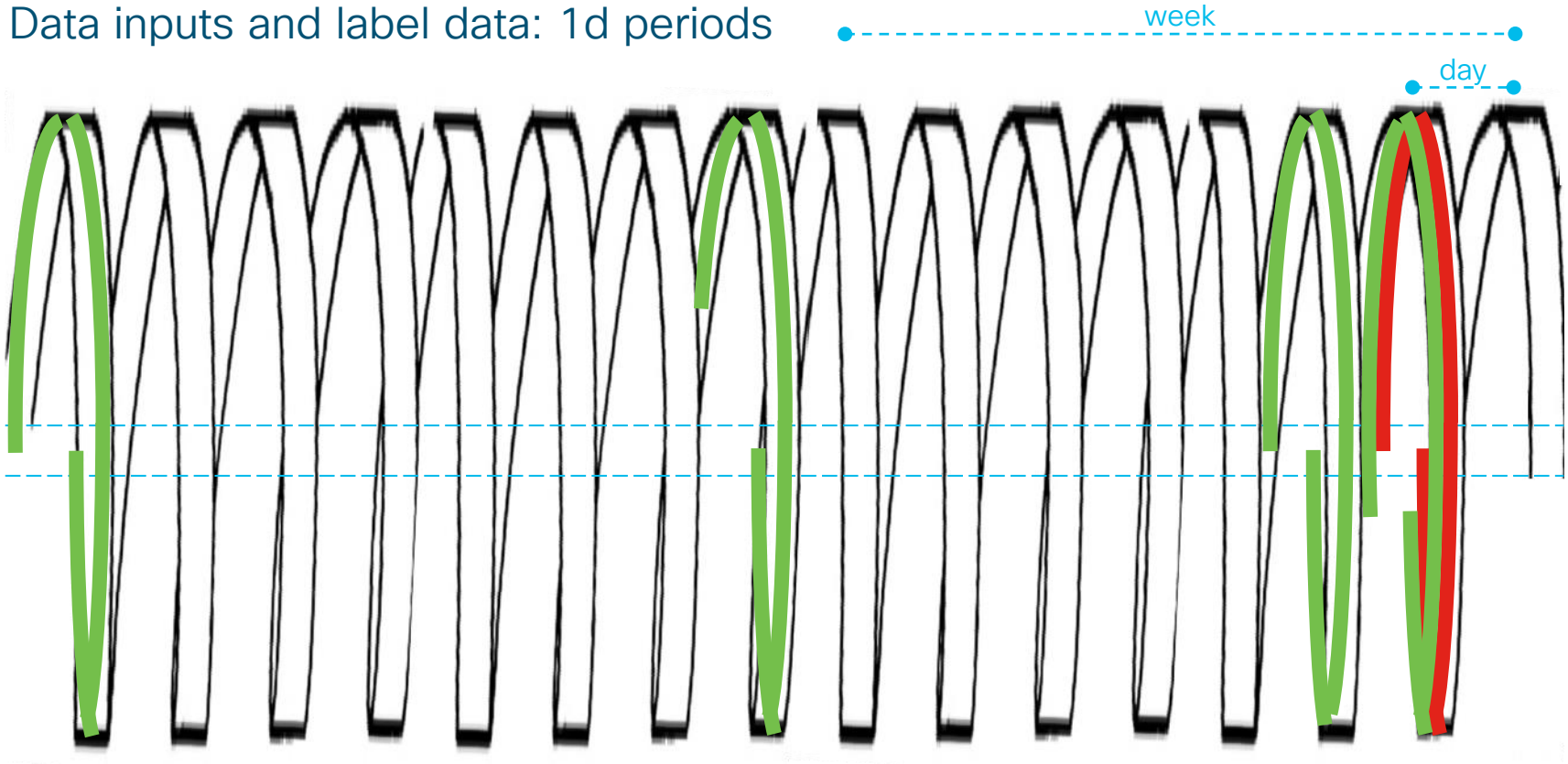
Basis for Inference

Data inputs and validation data



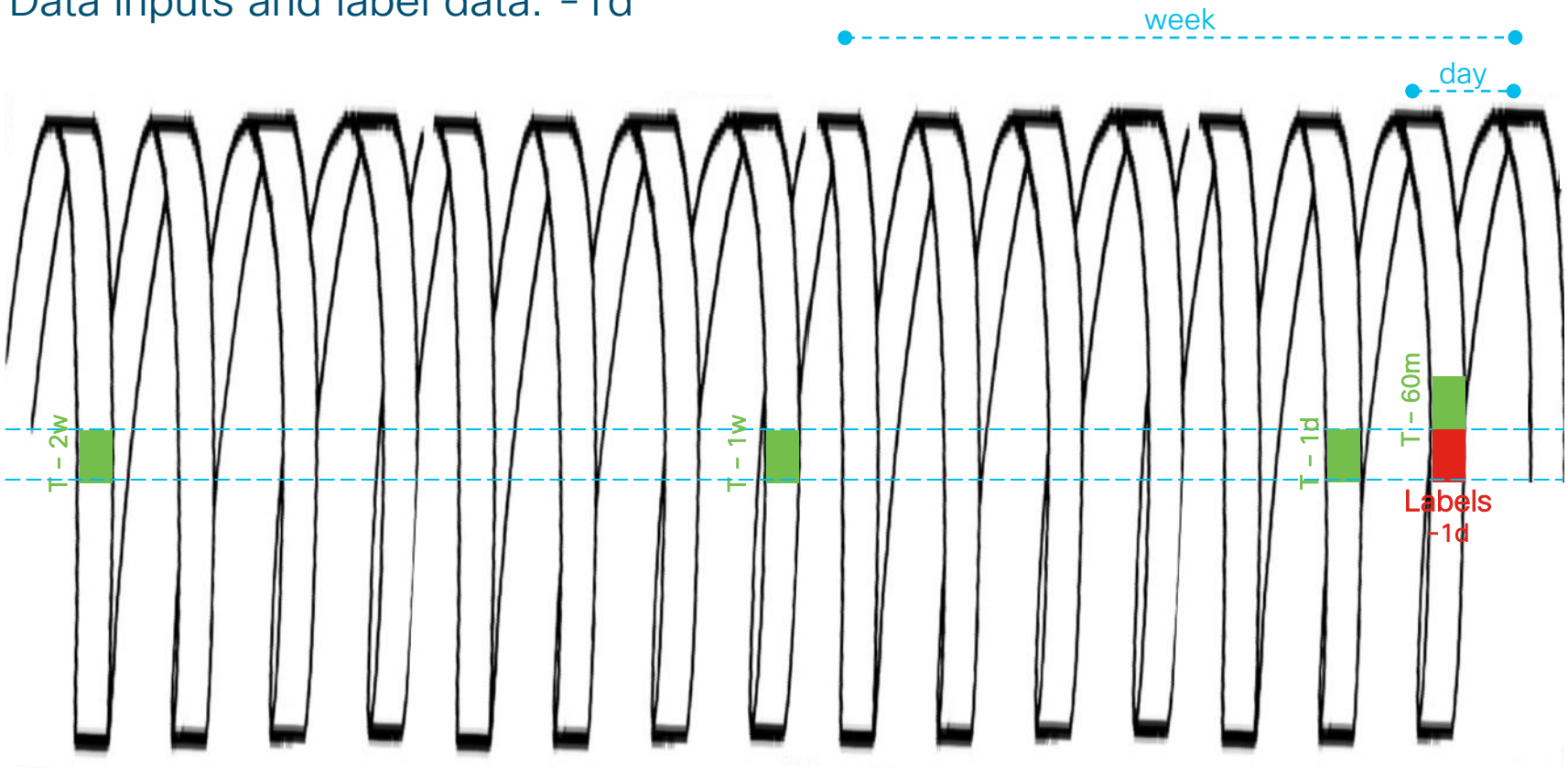
Training: Larger Datasets

Data inputs and label data: 1d periods



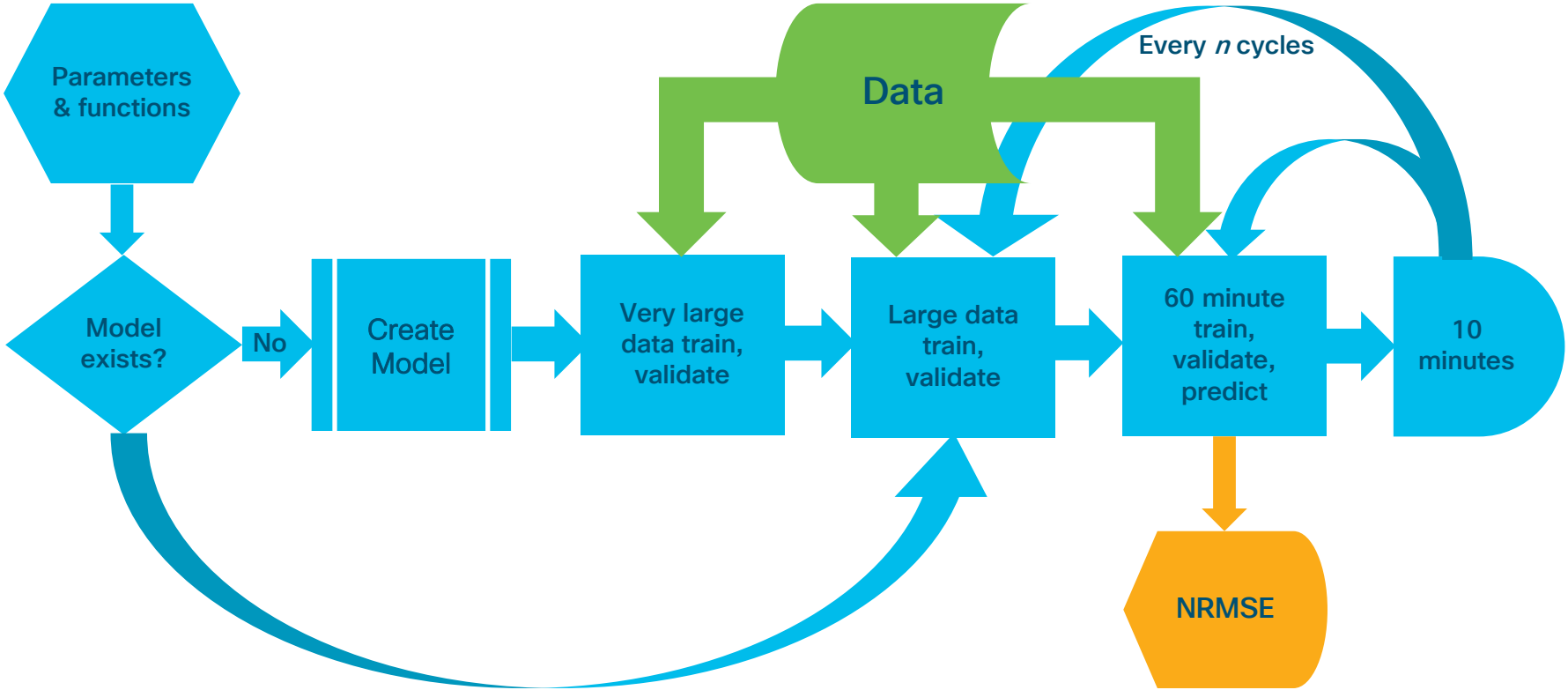
Training

Data inputs and label data: -1d

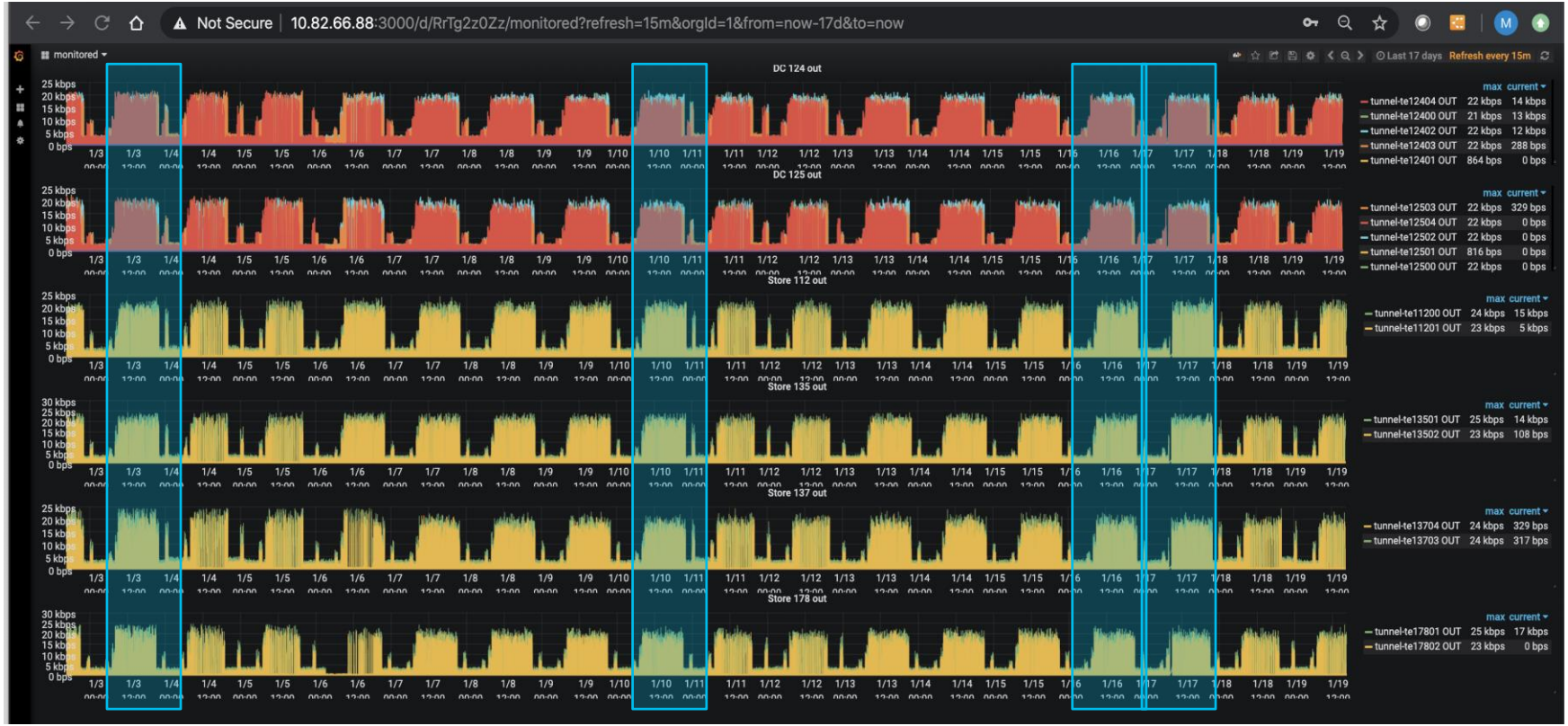


Training: Initial, and Periodic; Predict

“Monitor” flowchart



The Data in 4x 24-hour DataFrame, [4x18]x2880



4 24-hour_period x 18 path x 2880 30-second_field = 72 column x 2880 row DataFrame

DataFrame

```
def read_train_long(record_count, label_prefix, verbose=True):
.
        print('\ntraining long data')
        print(train.describe())
```

```
training long data
       d_te11200_previous  d_te11200_1d  d_te11200_1w  d_te11200_2w  \
count        2.880000e+03  2.880000e+03  2.880000e+03  2.880000e+03
mean         1.093179e+07  1.210950e+07  1.215980e+07  2.038959e+07
std          1.072837e+07  1.140074e+07  1.141777e+07  1.141292e+07
min          0.000000e+00  1.365000e+03  1.087200e+04  2.841000e+03
25%          1.663094e+06  1.665956e+06  1.689180e+06  1.086874e+07
50%          5.952412e+06  8.211694e+06  8.303540e+06  2.399129e+07
75%          1.959832e+07  2.178496e+07  2.212936e+07  3.107957e+07
max          3.269652e+07  3.305397e+07  3.308679e+07  3.262037e+07

       d_te11201_previous  d_te11201_1d  d_te11201_1w  d_te11201_2w  \
count        2.880000e+03  2.880000e+03  2.880000e+03  2.880000e+03
mean         7.527995e+06  8.306467e+06  8.337991e+06  1.384803e+07
std          7.289889e+06  7.747967e+06  7.757351e+06  7.804774e+06
min          0.000000e+00  1.380000e+03  8.182000e+03  1.001000e+03
```

Column “label”

df is 72 x 2880

Math description of column

DataFrame: same format, smaller size

```
validation data
       d_te11200_previous  d_te11200_1d  d_te11200_1w  d_te11200_2w  \
count        1.200000e+02  1.200000e+02  1.200000e+02  1.200000e+02
mean         1.358485e+06  9.795289e+05  1.179847e+06  1.246457e+06
std          8.000545e+05  6.292864e+05  6.797141e+05  7.001280e+05
min          1.380000e+03  7.731200e+04  9.600000e+01  1.928000e+03
25%          4.905140e+05  3.478915e+05  6.696790e+05  8.413410e+05
50%          1.362629e+06  1.087073e+06  1.240100e+06  1.377682e+06
75%          2.190353e+06  1.559045e+06  1.827315e+06  1.777683e+06
max          2.840542e+06  2.067730e+06  2.336623e+06  2.262407e+06


       d_te11201_previous  d_te11201_1d  d_te11201_1w  d_te11201_2w  \
count        1.200000e+02  1.200000e+02  1.200000e+02  1.200000e+02
mean         7.821397e+05  1.026548e+06  7.012429e+05  7.430594e+05
std          5.078806e+05  4.558566e+05  4.654541e+05  5.078973e+05
min          0.000000e+00  0.000000e+00  0.000000e+00  1.350000e+03
25%          3.321160e+05  5.965380e+05  4.511570e+05  3.695710e+05
50%          9.229315e+05  9.878900e+05  8.389290e+05  7.320545e+05
75%          1.225480e+06  1.511388e+06  1.236286e+06  1.205447e+06
max          1.499793e+06  1.857222e+06  1.556056e+06  1.526029e+06
.
```

df is 72 x 120

Fetching, Formatting, Conditioning the Data

```
def read_data(field_key, measurement_name, condition1, condition2, condition3, limit,
label):
  query_db = str('SELECT "%s" FROM "%s" WHERE %s AND %s AND %s LIMIT %d ' % (field_key,
measurement_name, condition1, condition2, condition3, limit+1))
  data_db = client.query(query_db)
  print('\ndata_db:\n', data_db)
  data_df = pd.DataFrame(data_db[str(measurement_name)])
  print('\ndata_df:\n', data_df)
  print('\ndata_df description:\n', data_df.describe())
  data_df.columns = [label]
  data_df.reset_index(drop=True, inplace=True)
  data_df.fillna(method='ffill', inplace=True)
  data_df.fillna(method='bfill', inplace=True)
  data_df -= data_df.min()
  data_df.drop(data_df.index[0], inplace=True)
  print('\ndata_df:\n', data_df)
  print('\ndata_df description:\n', data_df.describe())
  sys.exit()
  # data_df = data_df.sub(data_df.shift(fill_value=0))
  # print('\n', query_db, '\n', data_df.describe())
  return data_df
```

Raw Data

```
data_db:
 defaultdict(<class 'list'>, {'Cisco-IOS-XR-infra-statsd-oper:infra-
statistics/interfaces/interface/latest/generic-
counters':                                        bytes-sent
2020-01-03 06:34:04.032000+00:00  2269840208
2020-01-03 06:34:34.031000+00:00  2269840208
2020-01-03 06:35:04.027000+00:00  2269840208
2020-01-03 06:35:34.031000+00:00  2269840208
2020-01-03 06:36:04.032000+00:00  2269840820
2020-01-03 06:36:34.041000+00:00  2269842350
2020-01-03 06:37:04.032000+00:00  2269842554
2020-01-03 06:37:34.036000+00:00  2269842554
2020-01-03 06:38:04.032000+00:00  2269842554
2020-01-03 06:38:34.033000+00:00  2269842554
2020-01-03 06:39:04.090000+00:00  2269842554
2020-01-03 06:39:34.035000+00:00  2269853583
2020-01-03 06:40:04.036000+00:00  2269854129
2020-01-03 06:40:34.033000+00:00  2269854129
2020-01-03 06:41:04.035000+00:00  2269854228
2020-01-03 06:41:34.033000+00:00  2269855713
2020-01-03 06:42:04.035000+00:00  2269856604
.
```

In Single Column DataFrame

```
data_df:
                                  bytes-sent
2020-01-03 06:34:04.032000+00:00  2269840208
2020-01-03 06:34:34.031000+00:00  2269840208
2020-01-03 06:35:04.027000+00:00  2269840208
2020-01-03 06:35:34.031000+00:00  2269840208
2020-01-03 06:36:04.032000+00:00  2269840820
2020-01-03 06:36:34.041000+00:00  2269842350
2020-01-03 06:37:04.032000+00:00  2269842554
2020-01-03 06:37:34.036000+00:00  2269842554
2020-01-03 06:38:04.032000+00:00  2269842554
2020-01-03 06:38:34.033000+00:00  2269842554
2020-01-03 06:39:04.090000+00:00  2269842554
2020-01-03 06:39:34.035000+00:00  2269853583
2020-01-03 06:40:04.036000+00:00  2269854129
2020-01-03 06:40:34.033000+00:00  2269854129
2020-01-03 06:41:04.035000+00:00  2269854228
2020-01-03 06:41:34.033000+00:00  2269855713
2020-01-03 06:42:04.035000+00:00  2269856604
2020-01-03 06:42:34.034000+00:00  2269866268
2020-01-03 06:43:04.034000+00:00  2269866916
.
```

```
data_df description:
          bytes-sent
count  2.881000e+03
mean   2.284429e+09
std    1.233722e+07
min    2.269840e+09
25%    2.271348e+09
50%    2.283434e+09
75%    2.296978e+09
max    2.303451e+09
```

Column Label Changed, Baselined, Conditioned

```
data_df:
       d_te11200_previous
1                       0
2                       0
3                       0
4                     612
5                    2142
6                    2346
7                    2346
8                    2346
9                    2346
10                   2346
11                  13375
12                  13921
13                  13921
14                  14020
15                  15505
16                  16396
17                  26060
18                  26708
19                  28328
.
```

```
data_df description:
       d_te11200_previous
count        2.880000e+03
mean         1.459358e+07
std          1.233636e+07
min          0.000000e+00
25%          1.507463e+06
50%          1.362955e+07
75%          2.713798e+07
max          3.361109e+07
```

Constructing Multi-Column DataFrame

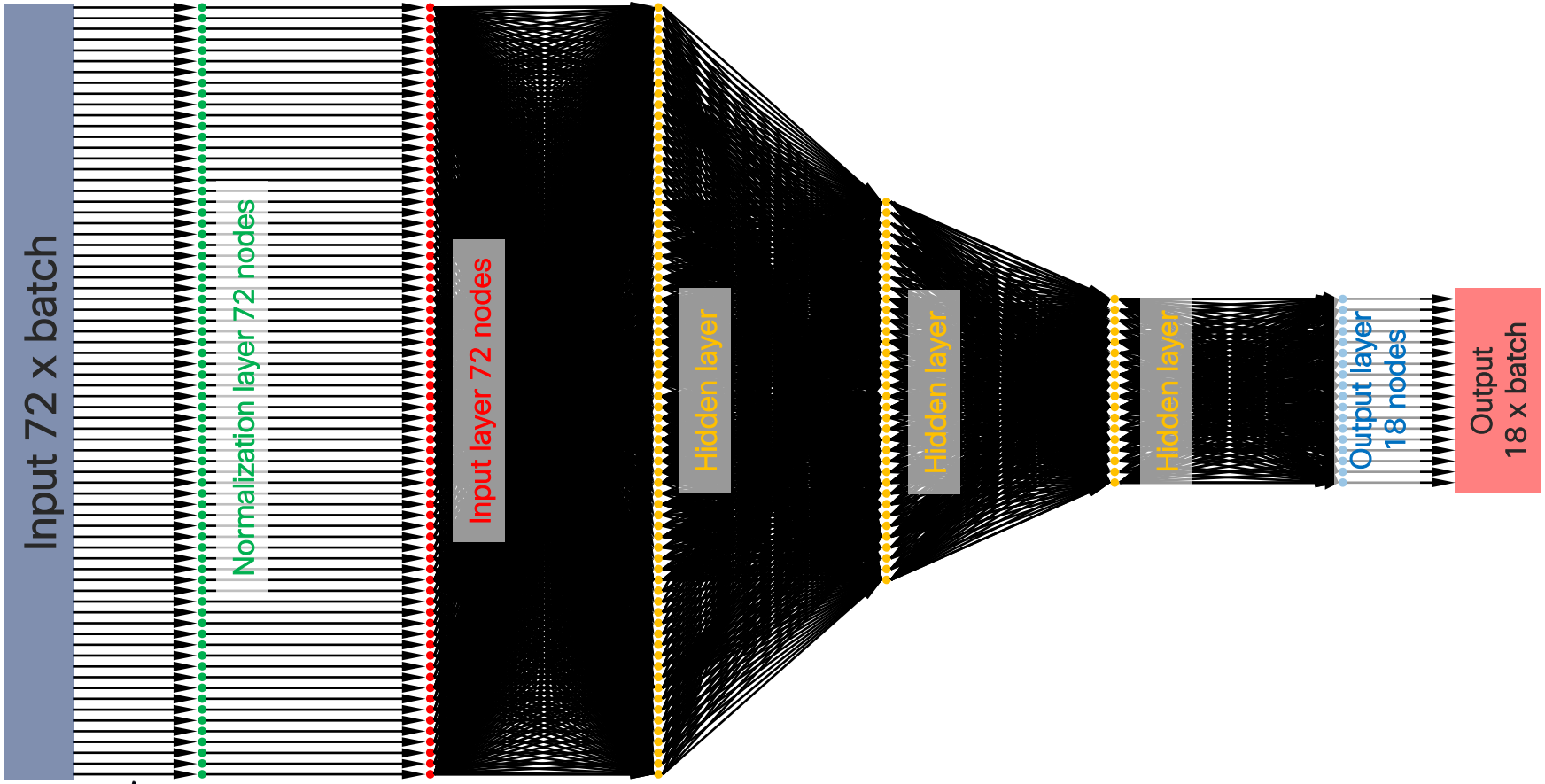
```
def read_train(record_count, label_prefix, verbose=True):
    for interface in tunnel_ifs:
        query_if = str('("interface-name" = \'%s\')' % (interface))
        label = str(label_prefix + interface[-7:] + "_previous")
        read_if = read_data('bytes-sent', 'Cisco-IOS-XR-infra-statsd-oper:infra-
statistics/interfaces/interface/latest/generic-counters', query_if, 'time >= now() - {} - 2h -
1m'.format(previous), 'time <= now()', record_count, label)
        if interface == tunnel_ifs[0]:
            train = read_if
        else:
            train = pd.concat([train, read_if], axis=1, sort=False)
        label = str(label_prefix + interface[-7:] + "_1d")
        read_if = read_data('bytes-sent', 'Cisco-IOS-XR-infra-statsd-oper:infra-
statistics/interfaces/interface/latest/generic-counters', query_if, 'time >= now() - {} - 1d - 1h -
1m'.format(previous), 'time <= now()', record_count, label)
        train = pd.concat([train, read_if], axis=1, sort=False)
.
        label = str(label_prefix + interface[-7:] + "_2w")
        read_if = read_data('bytes-sent', 'Cisco-IOS-XR-infra-statsd-oper:infra-
statistics/interfaces/interface/latest/generic-counters', query_if, 'time >= now() - 3w - {} - 1h -
1m'.format(previous), 'time <= now()', record_count, label)
        train = pd.concat([train, read_if], axis=1, sort=False)

    train.fillna(method='ffill', inplace=True)
    return train
```

Unique column label

Concatenate columns

The Model



The Neural Network, and Sample Cycle

```
.
  my_optimizer = tf.train.AdamOptimizer(learning_rate=learning_rate)
  my_optimizer = tf.contrib.estimator.clip_gradients_by_norm(my_optimizer, 5.0)
  dnn_regressor = tf.estimator.DNNRegressor(
      feature_columns=construct_feature_columns(training_examples),
      hidden_units=hidden_units,
      optimizer=my_optimizer,
      model_dir= model_directory,
      label_dimension= len(tunnel_ifs) + len(physical_ifs)
.
hidden_units = [72, 36, 18]      # probably an overkill for our small scale
.
    cycle += 1
    print('cycle number ', cycle)
    dnn_regressor = train_nn_regression_model(
        learning_rate = 0.0003,
        steps = 1000,
        batch_size = 120,
        hidden_units = hidden_units,
        training_examples = read_train(120, 'd_'),
        training_targets = read_train_target(120, 'l_'),
        validation_examples = read_validate(120, 'd_'),
        validation_targets = read_last_target(120, 'v_'),
        prediction = True
.
```

Set once, for a new model

Can change every call

+ Normalization Layer

```
def read_train_long(record_count, label_prefix, verbose=True):
    global feature_mean
    global feature_std
    global feature_max
.
    if feature_mean == 0:
        feature_mean = train.mean().mean()
        print('feature mean: ', feature_mean)
    if feature_std == 0:
        feature_std = train.std().mean()
        print('feature std: ', feature_std)
    if feature_max == 0:
        feature_max = train.max().mean() / 24 # The mean max per 1 hour
        print('feature max: ', feature_max)
.
def construct_feature_columns(input_features):
.
  # epsilon = 0.000001
.
  # choose best normalization of input data
.
  return set([tf.feature_column.numeric_column(my_feature, normalizer_fn=lambda val: (val) /
(feature_max))
              for my_feature in input_features])
```

The data - inputs

```
feature mean:  7692395.787210648
feature std:  7084487.678199986
feature max:  899857.4594907407

training long data
       d_te11200_previous  d_te11200_1d  d_te11200_1w  d_te11200_2w  \
count        2.880000e+03  2.880000e+03  2.880000e+03  2.880000e+03
mean         1.460392e+07  1.148071e+07  1.377143e+07  1.286884e+07
std          1.208010e+07  1.079281e+07  1.288531e+07  1.193284e+07
min          5.194400e+04  8.214400e+04  7.731200e+04  1.128000e+03
25%          5.464170e+06  2.660394e+06  3.144108e+06  2.952690e+06
50%          7.655800e+06  5.238823e+06  6.525733e+06  6.685706e+06
75%          2.390421e+07  2.051300e+07  2.478536e+07  2.295742e+07
max          4.010818e+07  3.499195e+07  3.808687e+07  3.528133e+07

       d_te11201_previous  d_te11201_1d  d_te11201_1w  d_te11201_2w  \
count        2.880000e+03  2.880000e+03  2.880000e+03  2.880000e+03
mean         9.872998e+06  7.983878e+06  9.340476e+06  8.721320e+06
std          8.096162e+06  7.326994e+06  8.648336e+06  8.025967e+06
min          0.000000e+00  1.080000e+03  0.000000e+00  0.000000e+00
25%          3.781685e+06  2.003244e+06  2.256103e+06  2.102746e+06
50%          5.334513e+06  3.862060e+06  4.561317e+06  4.507384e+06
```

Data indicators,
Useful for normalization

Column key

Record count

Min value

Max value

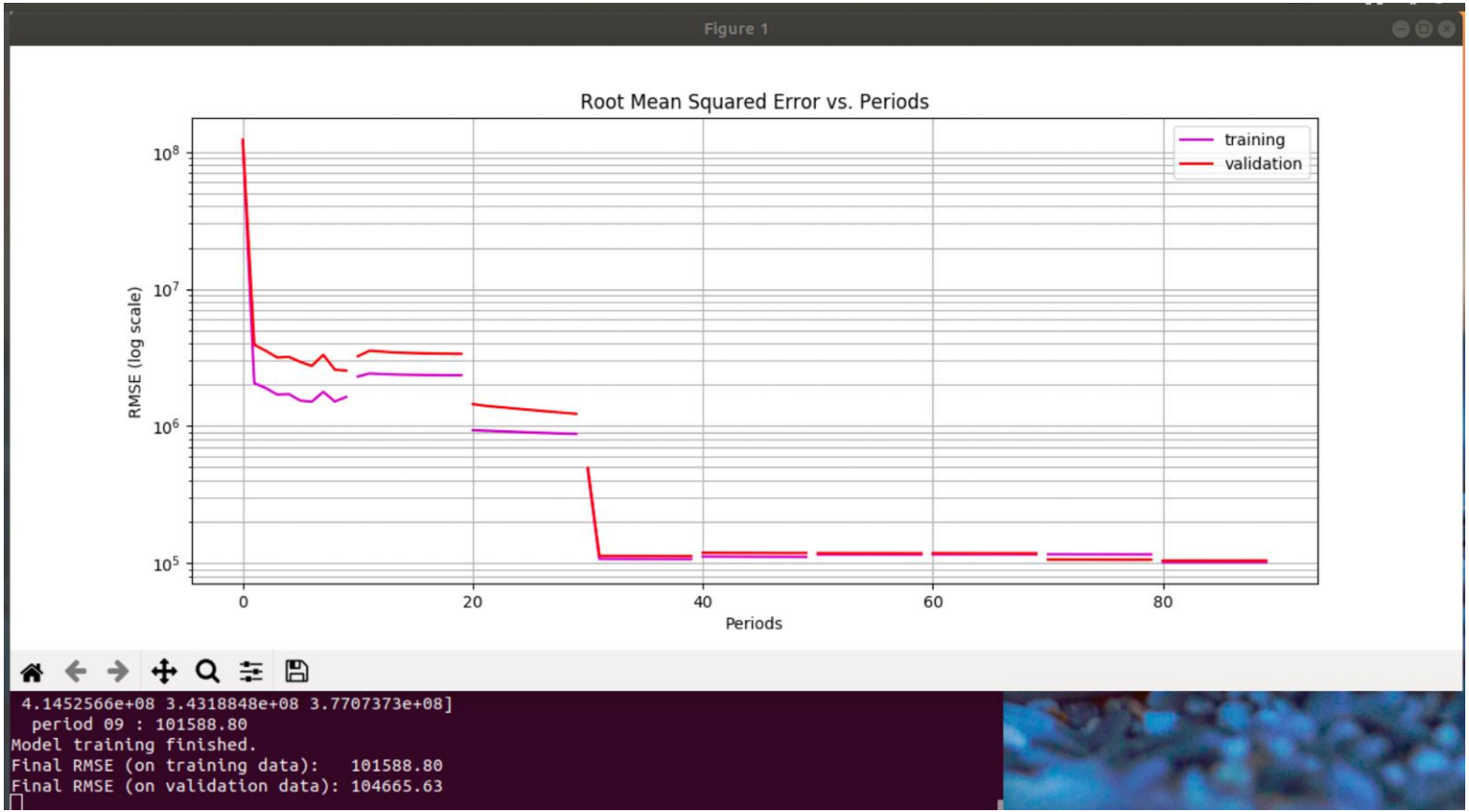
Check the Data: Model Output

```
.
training_predictions baoundaries
[ -9676.203    43134.777    30126.502    49435.23    -32687.848
  53458.71    -41498.527     2386.6956   -7947.478      841.848
  25334.666   -29982.818   -35702.715    39189.33       822.5124
  43476.05     48315.46       459.85422]
[2567748.5   1706653.8   1980904.5   2473544.2   2354506.8   2123372.
 2704984.5   1812864.2   2203519.      67233.766 1694452.8   1959683.9
 2268782.5   1549306.6     66137.98  2231304.2   1900274.9   1607212.8  ]
validation_predictions boundaries
[  9514.773    37849.348     3103.1733   43236.957    -5364.3784
  22101.646   -17330.65     16654.14      8049.155      765.1315
   3570.572    -6731.2876  -15120.497    33285.02       766.89764
  37848.816    19647.463    13101.8    ]
[2846600.   2264547.2  3210291.   2034950.   3650229.2  1717540.4
 1242460.5  1751227.8  2441805.8    68177.71 2717810.5  3045913.5
 1010453.4  2042300.1    66662.13 1811712.9  1519572.6  1552536.8 ]
  period 08 : 55037.89
.
```

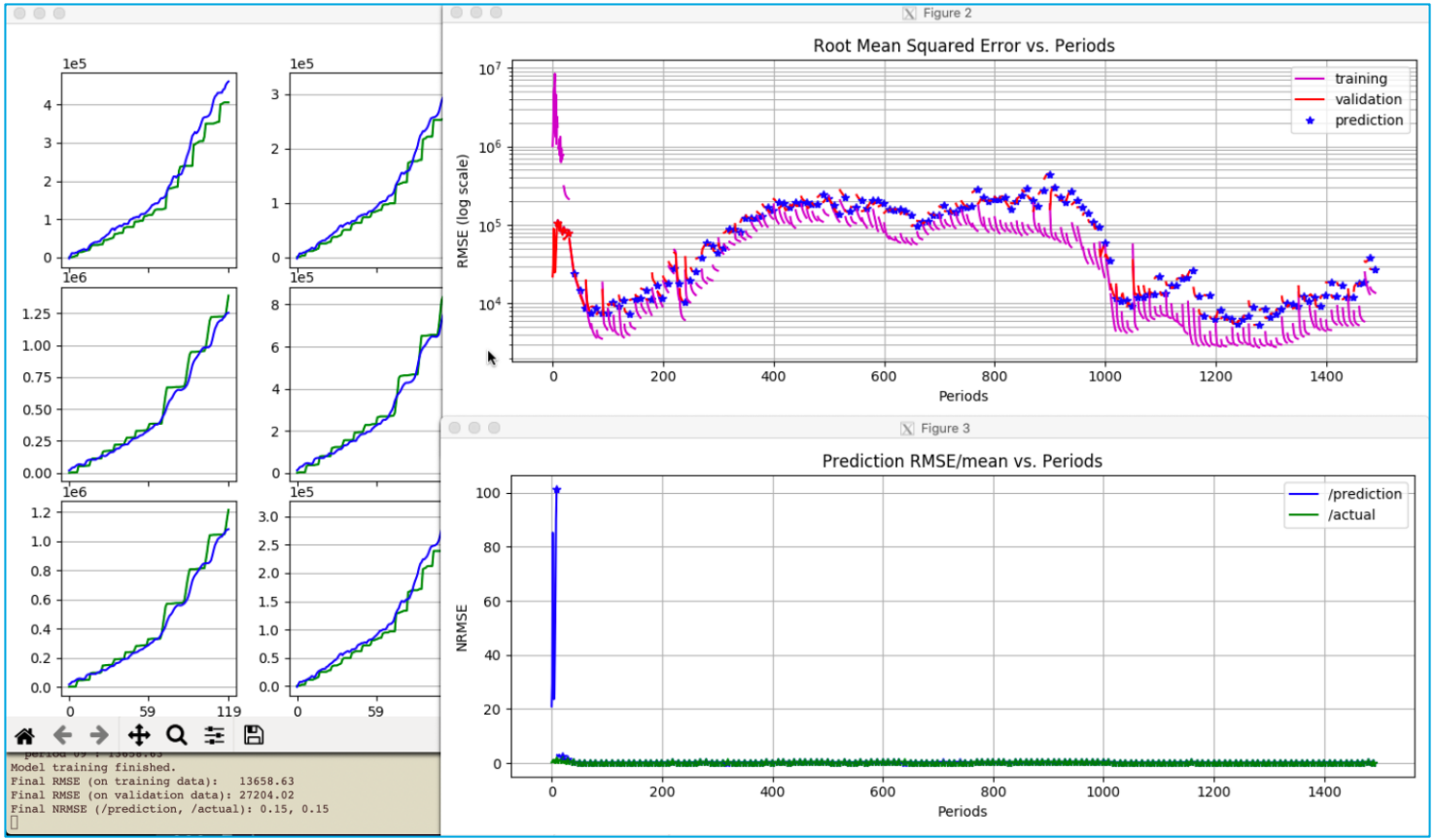
The Training

```
.
 period 08 : 3582.59
training_predictions baoundaries
[ 7208.3516   3251.541    7443.031    4104.243    4939.107    4307.9517
 -4889.6     -2828.935    5941.4014    726.0451   7325.6025   4610.5356
 -3598.1191   3449.1812    635.07806  4949.       4667.8926  -1990.8337 ]
[228423.4  177642.34 234673.56 181417.75 247542.95 188667.36 242869.14
 185446.6  211784.83  67781.5  213718.23 222849.58 222998.84 172893.55
  67602.49 174859.56 182980.92 178175.17]
validation_predictions boundaries
[ 6515.7734   3770.0752   8726.353    3760.6396  13467.836    3967.8716
 -5080.483   -3837.6772   5243.997     539.8233   8261.776   11545.85
 -3959.705    3852.7988    432.32617  4614.4375   4369.189   -3105.7751 ]
[193788.67  157614.84  150174.69  216238.94  187304.67  213302.8
 288468.47  197036.98  181182.31   51306.69  143677.8   166325.1
 265046.75  152532.97   51454.254 206665.23  205994.08  187957.39 ]
  period 09 : 3548.03
Model training finished.
Final RMSE (on training data):   3548.03
Final RMSE (on validation data): 22812.22
Final NRMSE (/prediction, /actual): 0.25, 0.25
cycle number  788
```

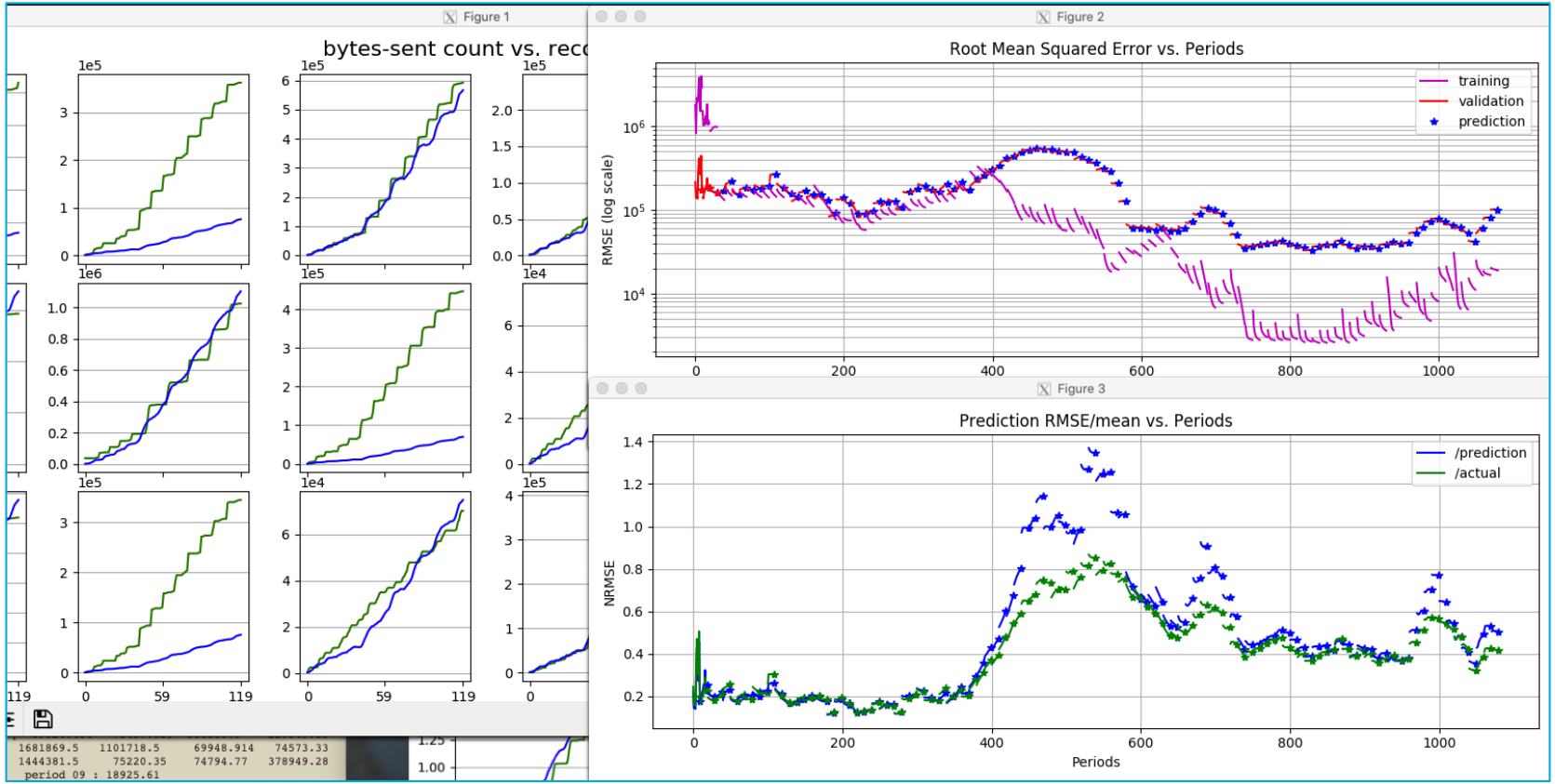
Training Makes Perfect!



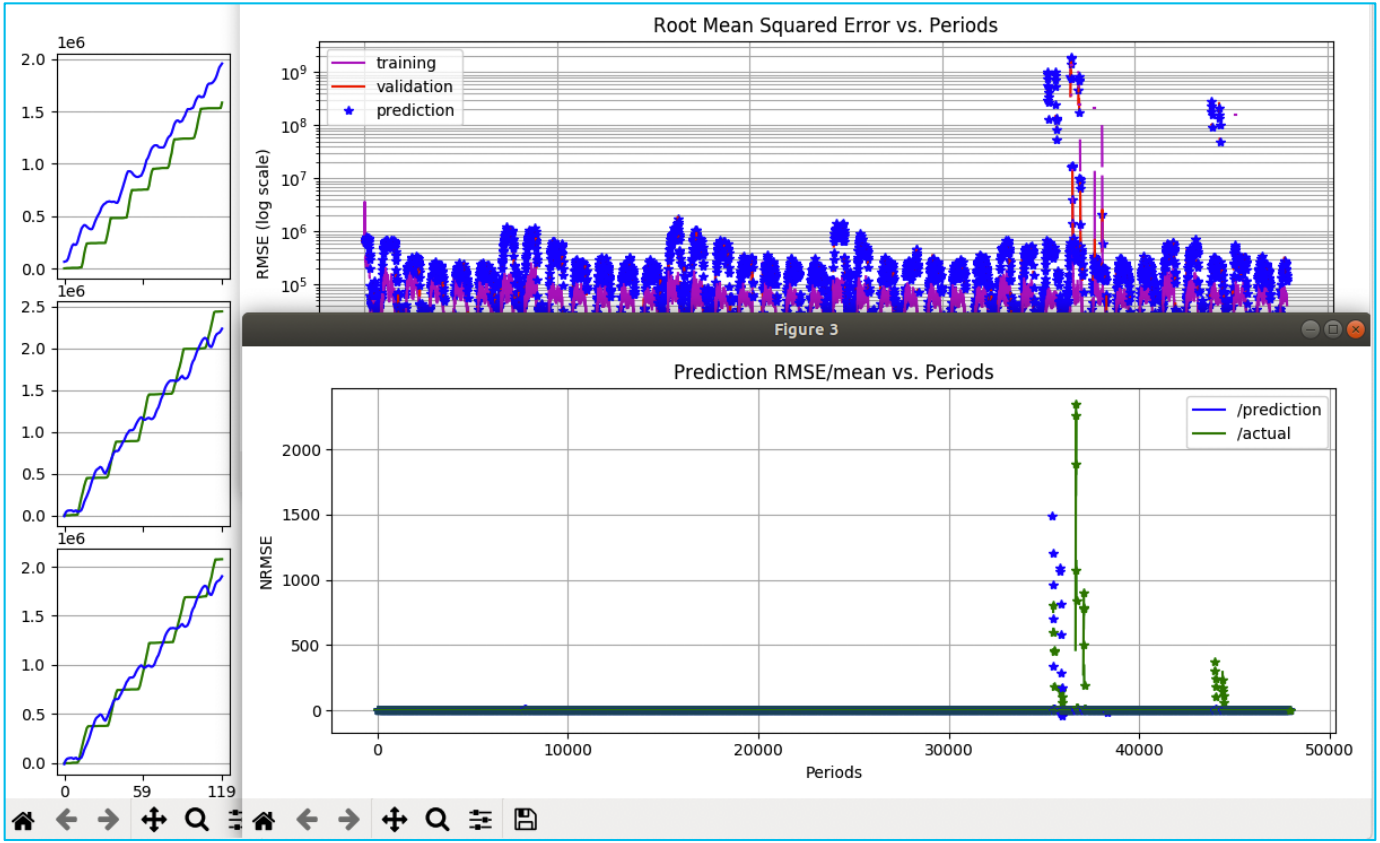
Best Indicator: NRMSE



A Short Term, Small Issue, & the Morning After



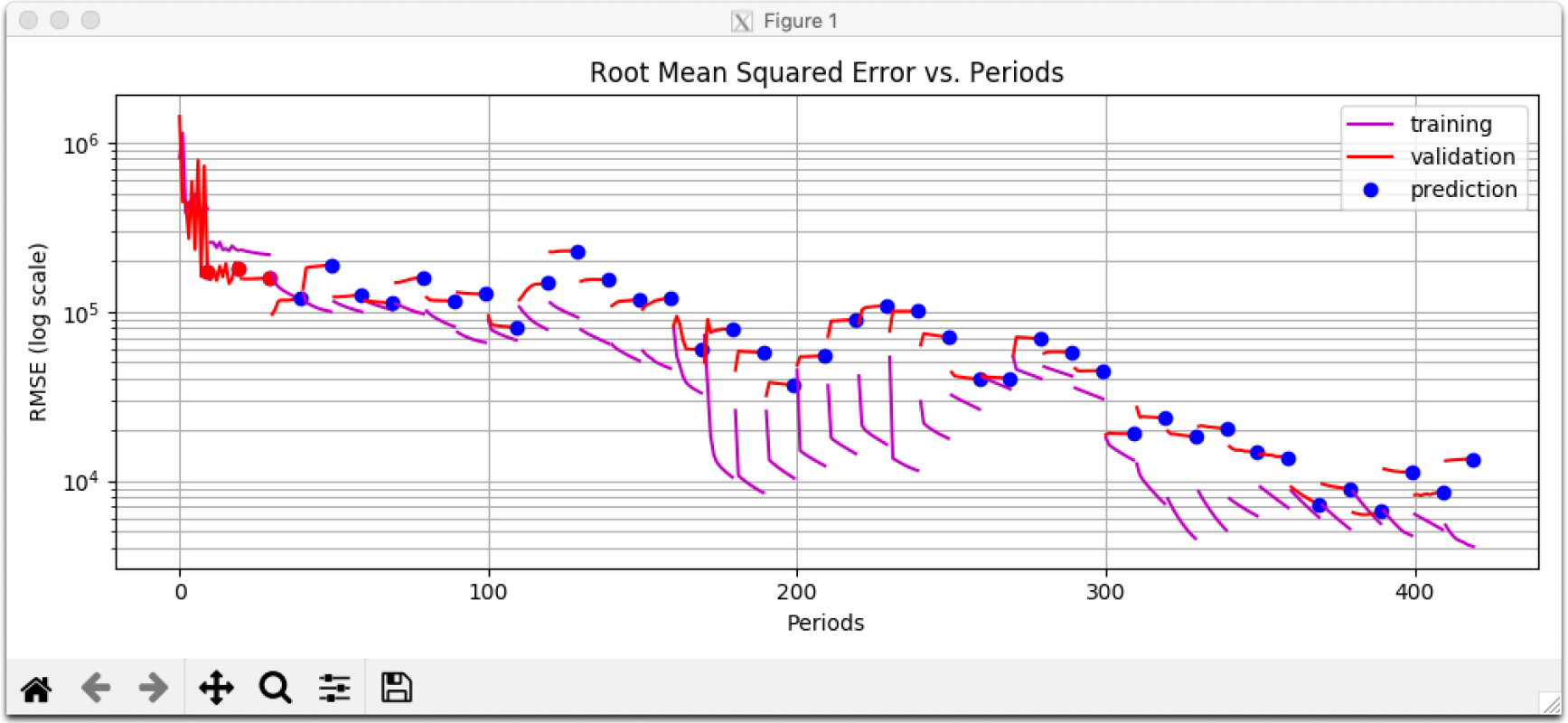
A Severe Issue



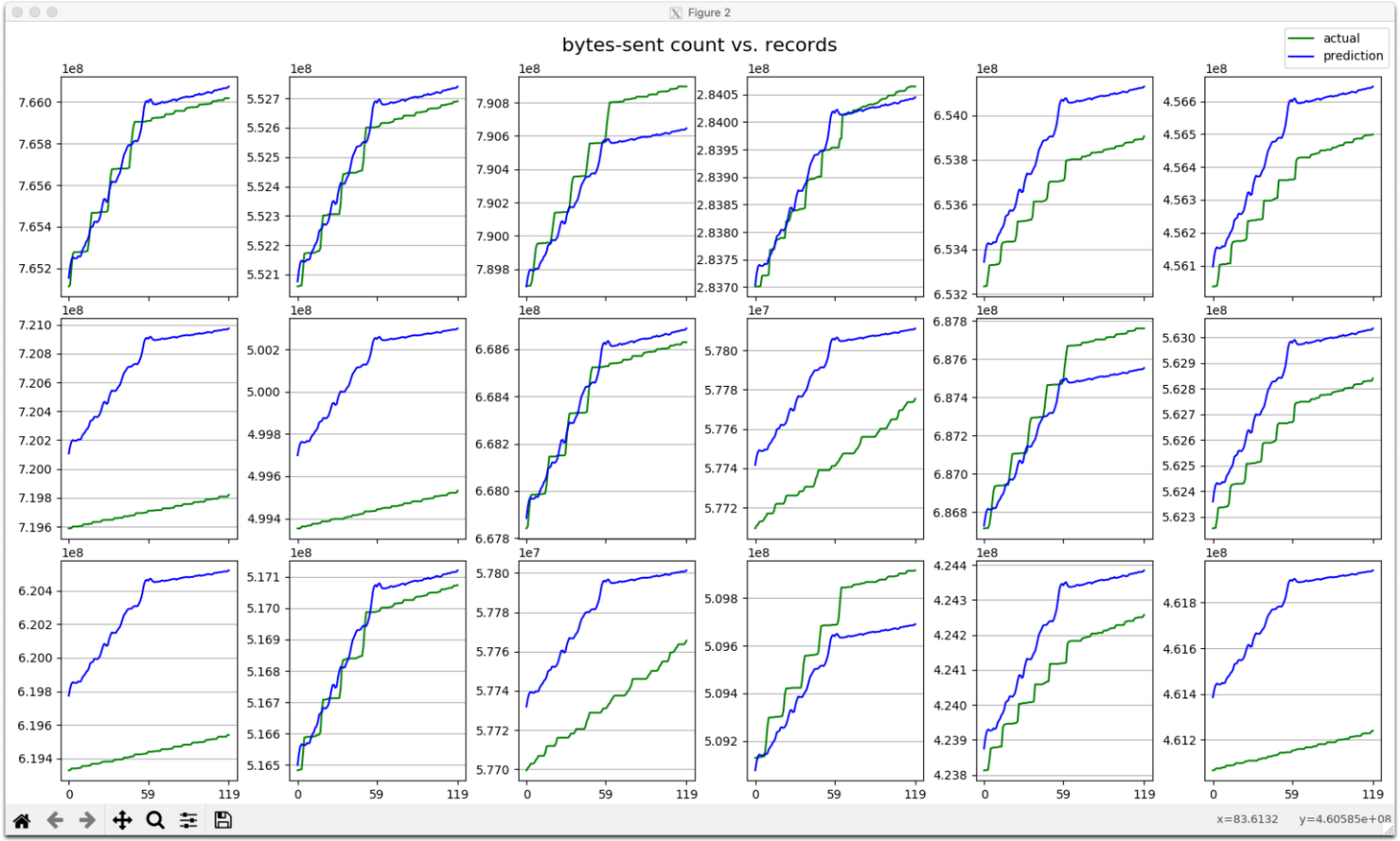
Visualize Progress: RMSE

```
.
 for period in range (0, periods):
    # Train the model, starting from the prior state.
    dnn_regressor.train(
        input_fn=training_input_fn,
        steps=steps_per_period
    )
    # Take a break and compute predictions.
    training_predictions = dnn_regressor.predict(input_fn=predict_training_input_fn)
    training_predictions = np.array([[item['predictions'][i] for i in range(0,
len(tunnel_ifs) + len(physical_ifs))] for item in training_predictions])
.
    validation_predictions =
dnn_regressor.predict(input_fn=predict_validation_input_fn)
    validation_predictions = np.array([[item['predictions'][i]for i in range(0,
len(tunnel_ifs) + len(physical_ifs))] for item in validation_predictions])
.
  if if_plot:
    # RMSE values and graphs
    global x_periods
    x_periods += periods
    plt.ion()
.
```

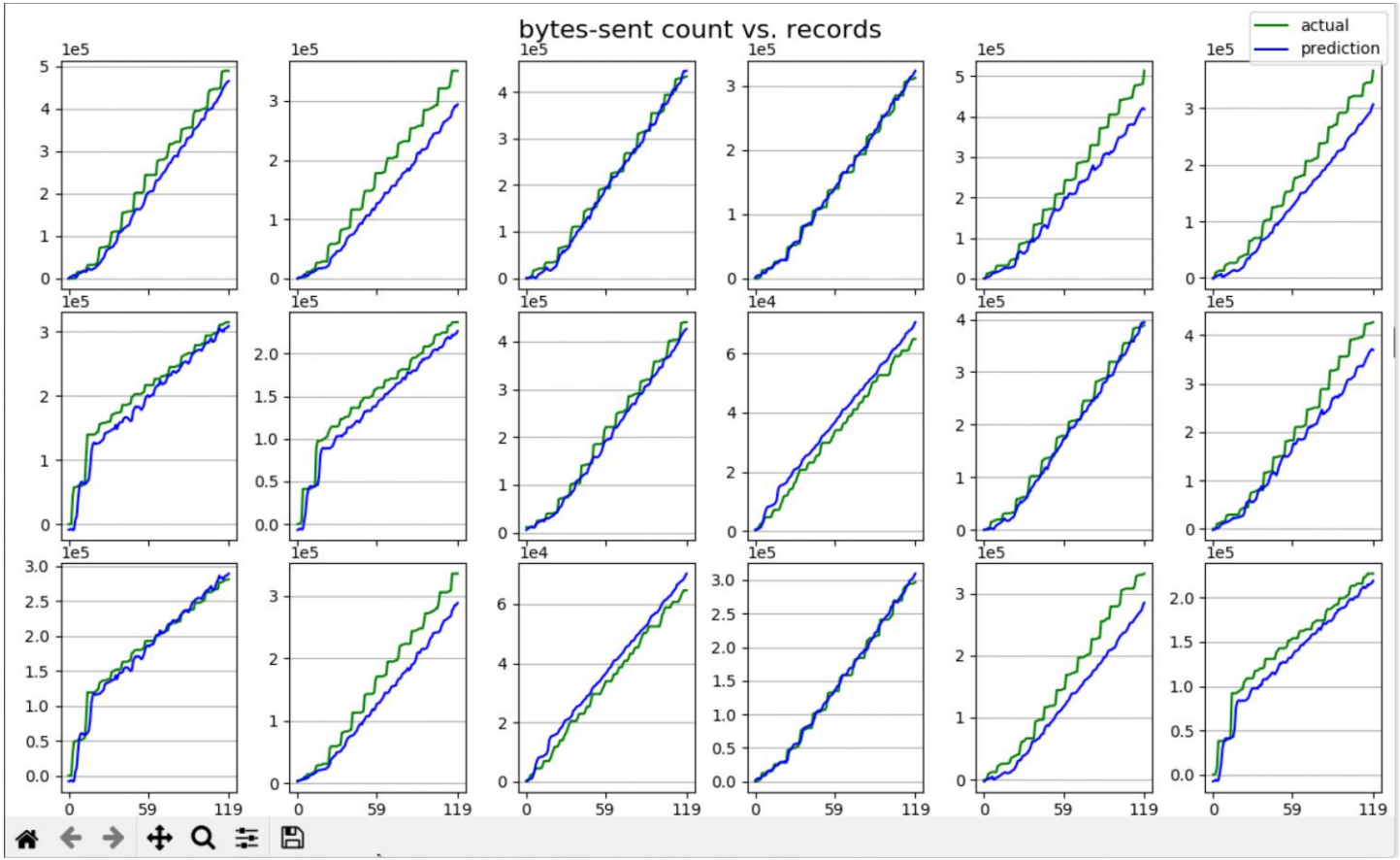
Visualize Progress: RMSE



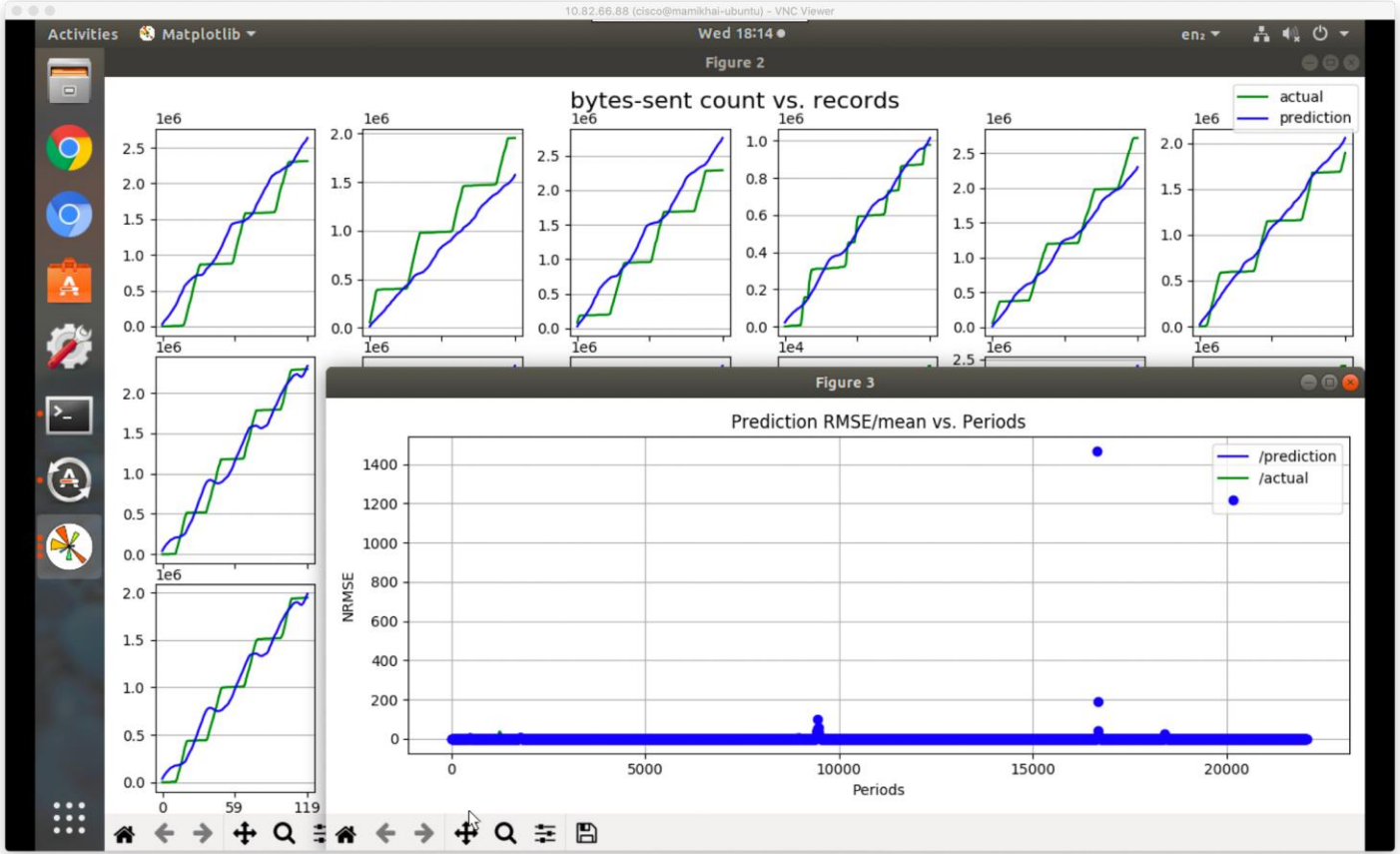
Validate: Visualize Prediction vs. Actual



Validate: Visualize Prediction vs. Actual



Eyes are Open?



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