

Plot the Graph for the total number of issues with different labels created and closed on every date on GitHub for SPM587SP19 issues project

SPM Assignment: 5
Phase 3
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```
In [1]: import os

import cPickle as pickle

import pandas as pd                # panda's nickname is pd
import numpy as np                # numpy as np
from pandas import DataFrame, Series  # for convenience
import matplotlib.pyplot as plt

%matplotlib inline
```

Requirement : Get the total number of issues with different labels for every date and plot them in a stacked chart

```
In [2]: # Read the JSON file into a list of dictionaries

import json
list_of_issues_dict_data = [json.loads(line) for line in open('SPM587SP19issues...)]
```

```
In [3]: # Create the DataFrame object for the list_of_issues_dict_data object

issues_df = DataFrame(list_of_issues_dict_data)
```

In [4]: *# Sanity test: print first 10 rows in our DataFrame*

```
issues_df
```

Out[4]:

	Author	State	closed_at	created_at	issue_number	labels
0	DSP19SCM782	open	None	2019-04-21	491	[Category:Bug, DetectionPhase:Field, Originati...
1	DSP19SCM782	open	None	2019-04-21	490	[Category:Bug, DetectionPhase:Design, Originat...
2	DSP19SCM781	open	None	2019-04-21	489	[Category:Enhancement, DetectionPhase:Coding, ...
3	DSP19SCM781	open	None	2019-04-21	488	[Category:Bug, DetectionPhase:Testing, Origina...
4	DSP19SCM781	open	None	2019-04-21	487	[Category:Inquiry, DetectionPhase:Design, Orig...
5	SSP19SCM781	open	None	2019-04-20	486	[Category:Enhancement, DetectionPhase:Design, ...
6	SSP19SCM781	open	None	2019-04-20	485	[Category:Enhancement, DetectionPhase:Field, O...
7	SSP19SCM781	open	None	2019-04-20	484	[Category:Inquiry, DetectionPhase:Field, Origi...
8	SSP19SCM782	open	None	2019-04-20	483	[Category:Inquiry, DetectionPhase:Field, Origi...
9	SSP19SCM782	open	None	2019-04-20	482	[Category:Bug, DetectionPhase:Field, Originati...
10	JSP19SCM03G	closed	2019-04-20	2019-04-20	480	[Category:Enhancement, DetectionPhase:Testing,...
11	ZSP19SCM65P	open	None	2019-04-20	479	[Address:McDonalds 2525 S King Dr Chicago 6061...
12	YSP19SCM26X	closed	2019-04-19	2019-04-18	477	[Category:Inquiry, DetectionPhase:Field, Origi...
13	RSP19SCM781	open	None	2019-04-18	476	[Category:Bug, DetectionPhase:Field, Originati...
14	RSP19SCM781	open	None	2019-04-18	475	[Category:Inquiry, DetectionPhase:Design, Orig...
15	TSP19SCM782	open	None	2019-04-18	474	[Category:Enhancement, DetectionPhase:Design, ...
16	TSP19SCM782	open	None	2019-04-18	473	[Category:Bug, DetectionPhase:Design, Originat...
17	TSP19SCM782	open	None	2019-04-18	472	[Category:Enhancement, DetectionPhase:Testing,...
18	DSP19SCM782	open	None	2019-04-18	471	[Category:Inquiry, DetectionPhase:Design, Orig...
19	DSP19SCM782	open	None	2019-04-18	470	[Category:Bug, DetectionPhase:Testing, Origina...

	Author	State	closed_at	created_at	issue_number	labels
20	TSP19SCM781	open	None	2019-04-18	469	[Category:Enhancement, DetectionPhase:Field, O...
21	TSP19SCM781	open	None	2019-04-18	468	[Category:Bug, DetectionPhase:Field, Originati...
22	TSP19SCM781	open	None	2019-04-18	467	[Category:Inquiry, DetectionPhase:Field, Oriti...
23	YSP19SCM26X	closed	2019-04-18	2019-04-18	465	[Address:2525 S Martin Luther King Dr Chicago ...
24	SSP19SCM782	open	None	2019-04-17	464	[Category:Bug, DetectionPhase:Coding, Originat...
25	SSP19SCM782	open	None	2019-04-17	463	[Category:Inquiry, DetectionPhase:Testing, Ori...
26	SSP19SCM781	open	None	2019-04-17	462	[Category:Enhancement, DetectionPhase:Design, ...
27	SSP19SCM781	open	None	2019-04-17	461	[Category:Inquiry, DetectionPhase:Field, Origi...
28	SSP19SCM781	closed	2019-04-21	2019-04-17	460	[Category:Enhancement, DetectionPhase:Design, ...
29	DSP19SCM782	closed	2019-04-21	2019-04-17	459	[Category:Enhancement, DetectionPhase:Field, O...
...
326	SPM587SP19	closed	2019-04-03	2019-03-31	31	[Category:Bug, DetectionPhase:Design, Originat...
327	SPM587SP19	closed	2019-04-02	2019-03-31	30	[Category:Inquiry, DetectionPhase:Coding, Orit...
328	SPM587SP19	closed	2019-04-07	2019-03-31	29	[Category:Bug, DetectionPhase:Design, Originat...
329	SPM587SP19	closed	2019-04-02	2019-03-31	28	[Category:Enhancement, DetectionPhase:Design, ...
330	SPM587SP19	closed	2019-04-03	2019-03-31	27	[Category:Inquiry, DetectionPhase:Design, Orit...
331	SPM587SP19	closed	2019-03-31	2019-03-31	26	[Category:Enhancement, DetectionPhase:Coding, ...
332	SPM587SP19	closed	2019-03-31	2019-03-31	25	[Category:Inquiry, DetectionPhase:Design, Orig...
333	SPM587SP19	closed	2019-03-31	2019-03-31	24	[Category:Enhancement, DetectionPhase:Coding, ...
334	SPM587SP19	closed	2019-03-31	2019-03-31	23	[Category:Bug, DetectionPhase:Design, Originat...
335	SPM587SP19	closed	2019-03-31	2019-03-31	22	[Category:Inquiry, DetectionPhase:Coding, Orig...
336	SPM587SP19	closed	2019-03-31	2019-03-31	21	[Category:Bug, DetectionPhase:Design, Originat...
337	SPM587SP19	closed	2019-03-31	2019-03-31	20	[Category:Bug, DetectionPhase:Coding, Oritinat...
338	SPM587SP19	closed	2019-03-31	2019-03-31	19	[Category:Enhancement, DetectionPhase:Design, ...

	Author	State	closed_at	created_at	issue_number	labels
339	SPM587SP19	closed	2019-03-31	2019-03-31	18	[Category:Inquiry, DetectionPhase:Coding, Orig...
340	SPM587SP19	closed	2019-03-31	2019-03-31	17	[Category:Bug, DetectionPhase:Field, Originati...
341	SPM587SP19	closed	2019-03-31	2019-03-31	16	[Category:Inquiry, DetectionPhase:Design, Orig...
342	SPM587SP19	closed	2019-03-31	2019-03-31	15	[Category:Enhancement, DetectionPhase:Coding, ...
343	SPM587SP19	closed	2019-03-31	2019-03-31	14	[Category:Inquiry, DetectionPhase:Design, Orig...
344	SPM587SP19	closed	2019-03-31	2019-03-30	13	[Category:Enhancement, DetectionPhase:Design, ...
345	SPM587SP19	closed	2019-03-31	2019-03-30	12	[Category:Enhancement, DetectionPhase:Coding, ...
346	SPM587SP19	closed	2019-03-31	2019-03-30	11	[Category:Enhancement, DetectionPhase:Design, ...
347	SPM587SP19	closed	2019-03-31	2019-03-30	10	[Category:Bug, DetectionPhase:Coding, Originat...
348	SPM587SP19	closed	2019-03-31	2019-03-30	9	[Category:Enhancement, DetectionPhase:Design, ...
349	SPM587SP19	closed	2019-03-31	2019-03-30	8	[Category:Bug, DetectionPhase:Coding, Originat...
350	SPM587SP19	closed	2019-03-31	2019-03-30	7	[Category:Inquiry, DetectionPhase:Design, Orig...
351	SPM587SP19	closed	2019-03-31	2019-03-30	6	[Category:Inquiry, DetectionPhase:Field, Origi...
352	SPM587SP19	closed	2019-03-31	2019-03-30	5	[Category:Inquiry, DetectionPhase:Field, Origi...
353	SPM587SP19	closed	2019-03-31	2019-03-29	4	[Category:Inquiry, DetectionPhase:Field, Origi...
354	SPM587SP19	closed	2019-03-31	2019-03-29	3	[Category:Inquiry, DetectionPhase:Field, Origi...
355	SPM587SP19	closed	2019-03-31	2019-03-24	1	[Category:Inquiry, DetectionPhase:Field, Origi...

356 rows × 6 columns

In [5]: *# Prepare and Clean the dataframe object*

```
wrangled_issues_df = issues_df[['Author', 'State', 'closed_at', 'created_at', 'issue_number', 'Category', 'DetectionPhase', 'OriginationPhase', 'Priority', 'Status']]
wrangled_issues_df.loc[0:len(wrangled_issues_df), 'OriginationPhase'] = np.NaN
wrangled_issues_df.loc[0:len(wrangled_issues_df), 'DetectionPhase'] = np.NaN
wrangled_issues_df.loc[0:len(wrangled_issues_df), 'Category'] = np.NaN
wrangled_issues_df.loc[0:len(wrangled_issues_df), 'Priority'] = np.NaN
wrangled_issues_df.loc[0:len(wrangled_issues_df), 'Status'] = np.NaN
```

In [6]: wrangled_issues_df

Out[6]:

	Author	State	closed_at	created_at	issue_number	labels	Orig
0	DSP19SCM782	open	None	2019-04-21	491	[Category:Bug, DetectionPhase:Field, Originati...	
1	DSP19SCM782	open	None	2019-04-21	490	[Category:Bug, DetectionPhase:Design, Originat...	
2	DSP19SCM781	open	None	2019-04-21	489	[Category:Enhancement, DetectionPhase:Coding, ...	
3	DSP19SCM781	open	None	2019-04-21	488	[Category:Bug, DetectionPhase:Testing, Origina...	
4	DSP19SCM781	open	None	2019-04-21	487	[Category:Inquiry, DetectionPhase:Design, Orig...	

```
In [7]: for i in range(0, len(wrangled_issues_df)):
        if wrangled_issues_df.iloc[i]['labels']:
            for label in wrangled_issues_df.iloc[i]['labels']:
                label_name= (label.split(':')[0])
                label_value= (label.split(':')[1])
                wrangled_issues_df.loc[i, label_name]=label_value
```

In [8]: wrangled_issues_df

Out[8]:

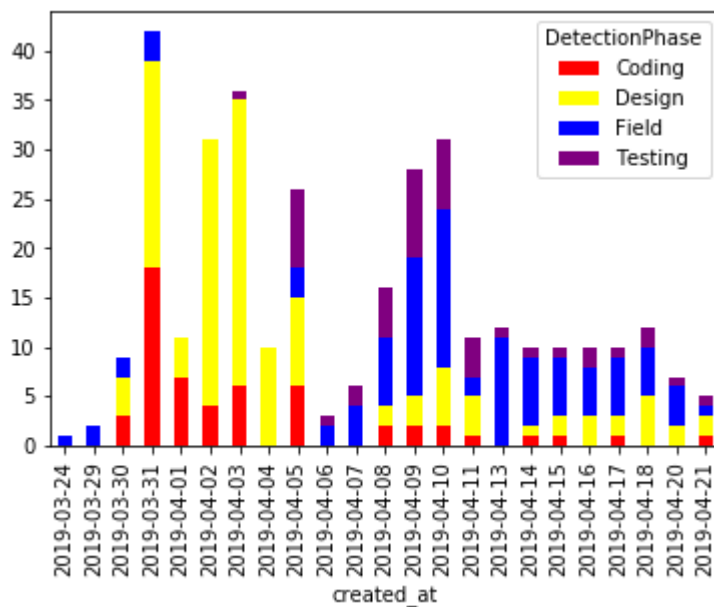
	Author	State	closed_at	created_at	issue_number	labels	Orig
0	DSP19SCM782	open	None	2019-04-21	491	[Category:Bug, DetectionPhase:Field, Originati...	
1	DSP19SCM782	open	None	2019-04-21	490	[Category:Bug, DetectionPhase:Design, Originat...	
2	DSP19SCM781	open	None	2019-04-21	489	[Category:Enhancement, DetectionPhase:Coding, ...	
3	DSP19SCM781	open	None	2019-04-21	488	[Category:Bug, DetectionPhase:Testing, Origina...	
4	DSP19SCM781	open	None	2019-04-21	487	[Category:Inquiry, DetectionPhase:Design, Orig...	[

```
In [9]: wrangled_issues_df['labels'][0]
```

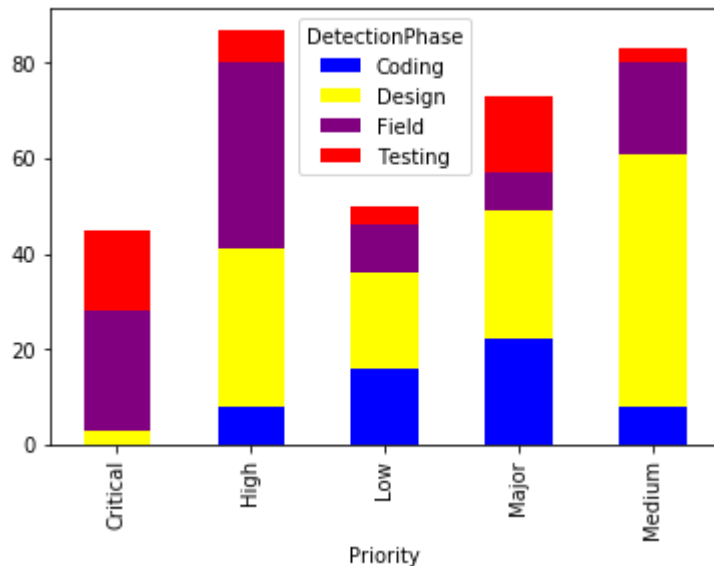
```
Out[9]: ['Category:Bug',
'DetectionPhase:Field',
'OriginationPhase:Field',
'Priority:Major',
'Status:pendingReview']
```

```
In [10]: # Plot in Bar Chart the total number of issues created every day for every DetectionPhase

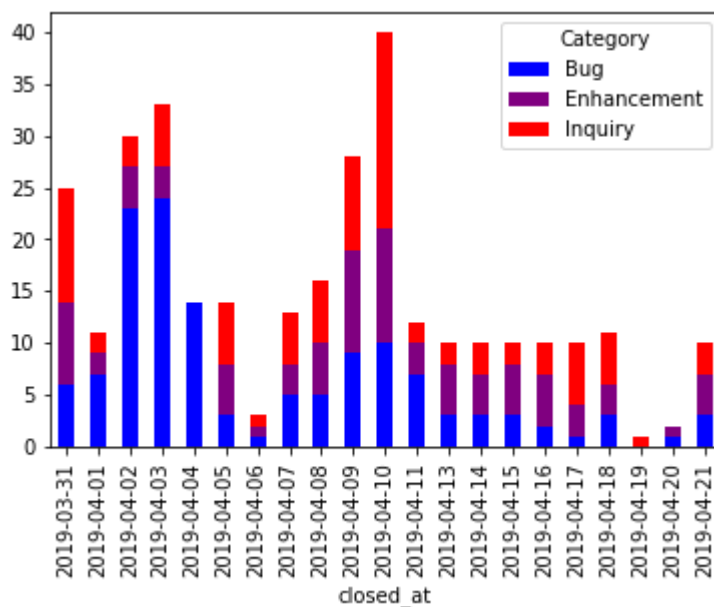
LabelsReviewedByDate = wrangled_issues_df.groupby(['created_at', 'DetectionPhase'])
dateLabelsFig = LabelsReviewedByDate.unstack().plot(kind='bar', stacked=True, color=
```



```
In [11]: # Plot in Bar Chart the total number of issues created for every Phase based on t
LabelsReviewedByDate = wrangled_issues_df.groupby(['Priority','DetectionPhase'])
dateLabelsFig = LabelsReviewedByDate.unstack().plot(kind='bar',stacked=True, co
```



```
In [12]: # Plot in Bar Chart the total number of issues closed every day for every Category
LabelsReviewedByDate = wrangled_issues_df.groupby(['closed_at','Category']).close
dateLabelsFig = LabelsReviewedByDate.unstack().plot(kind='bar',stacked=True, co
```

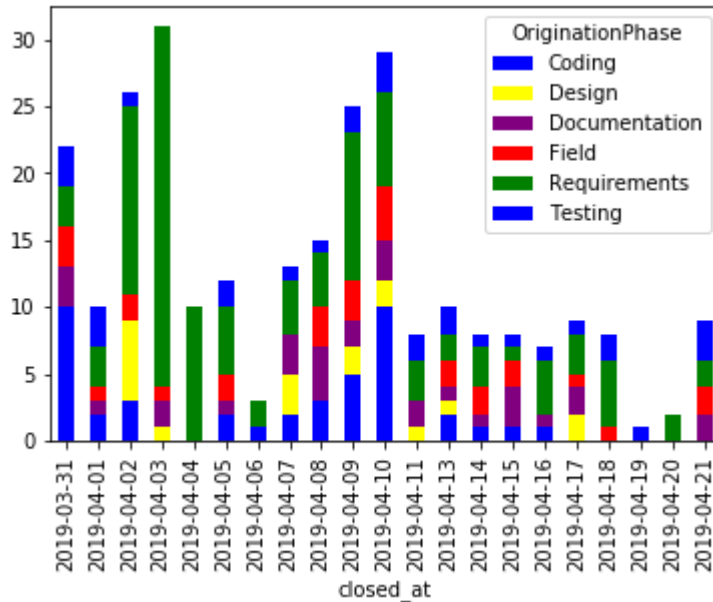


Requirement #1: Plot in Bar Chart the total

number of issues closed every day for every Origination Phase

```
In [13]: # Requirement #1:
# Plot in Bar Chart the total number of issues closed every day for every day for

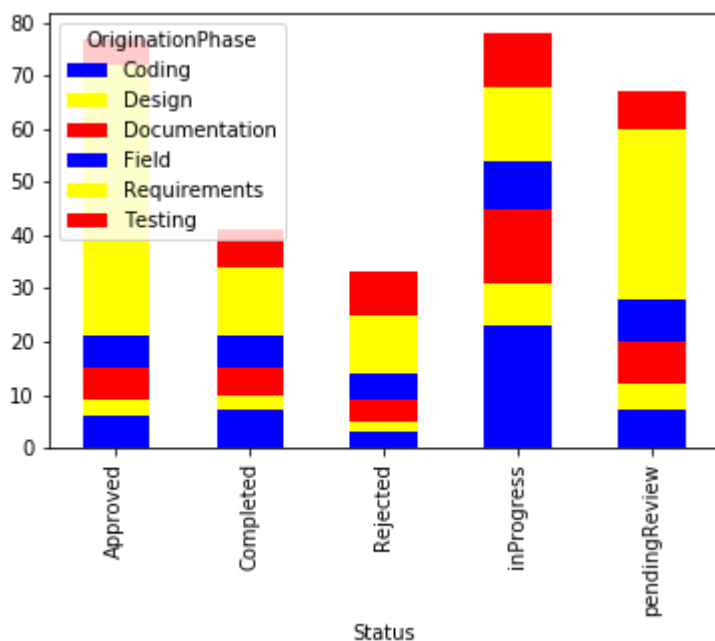
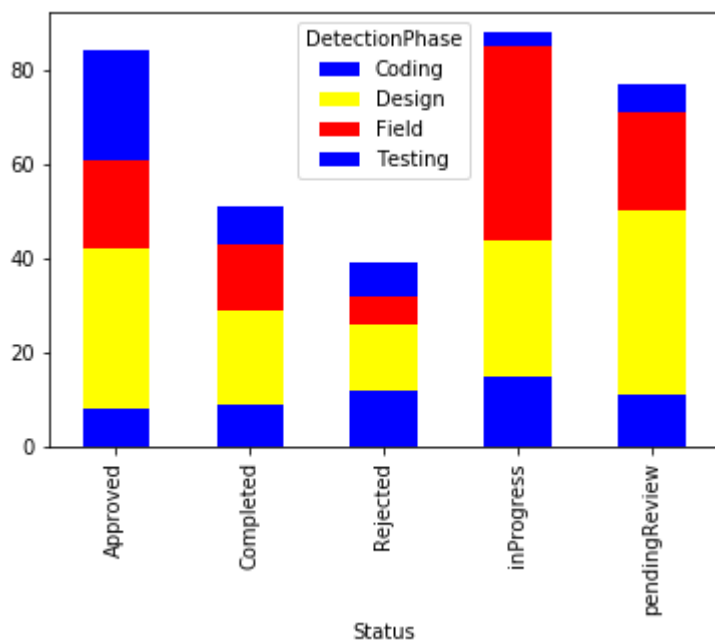
LabelsReviewedByDate = wrangled_issues_df.groupby(['closed_at', 'OriginationPhase'])
dateLabelsFig = LabelsReviewedByDate.unstack().plot(kind='bar', stacked=True, color=
```



Requirement #2: # Plot in Bar Chart the total number of issues created for every Phase based on their Status


```
In [14]: # Requirement #2:
# Plot in Bar Chart the total number of issues created for every Phase based on
#Is seen in 1st graph
LabelsReviewedByDate = wrangled_issues_df.groupby(['Status', 'DetectionPhase']).count()
dateLabelsFig = LabelsReviewedByDate.unstack().plot(kind='bar', stacked=True, color=

#Is seen in 2nd graph
LabelsReviewedByDate = wrangled_issues_df.groupby(['Status', 'OriginationPhase']).count()
dateLabelsFig = LabelsReviewedByDate.unstack().plot(kind='bar', stacked=True, color=
```

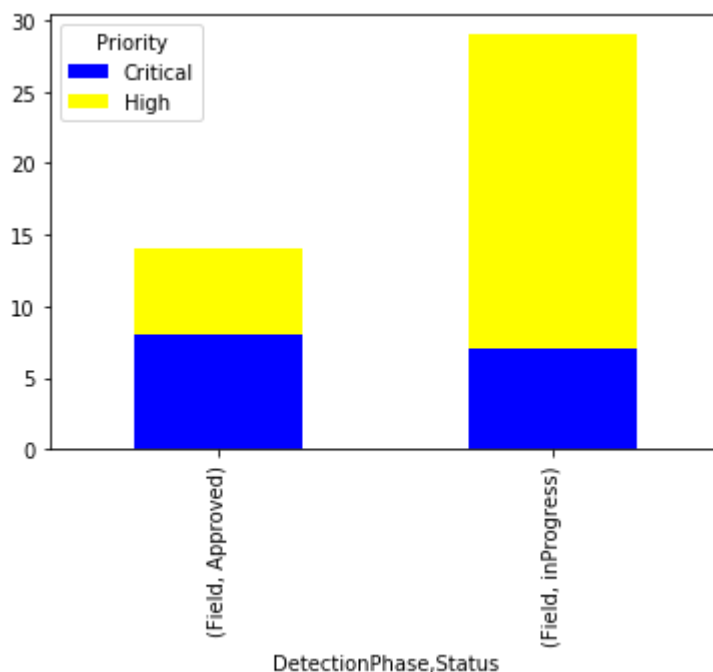
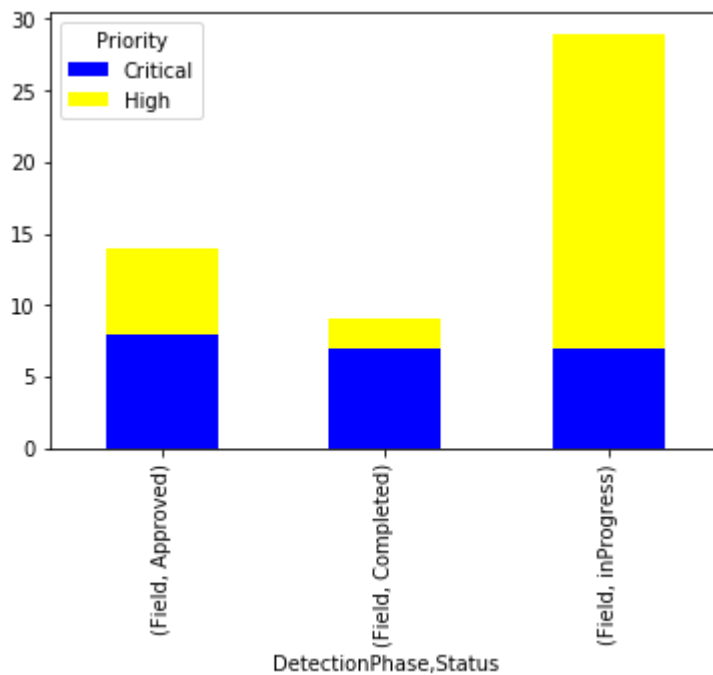


Requirement#3:Plot in Bar Chart the total number of issues for

1) DetectionPhase is Field AND Priority is Critical 2) DetectionPhase is Field AND Status is Completed 3) DetectionPhase is Field AND Priority is Critical AND Status is Approved 4) DetectionPhase is Field AND Priority is Critical or High AND Status is Approved or inProgress

```
In [15]: filteredData = wrangled_issues_df[wrangled_issues_df.Priority.isin(['Critical','High'])]
filteredData = filteredData[filteredData.Status.isin(['Approved', 'Completed', 'InProgress'])]
filteredData = filteredData[filteredData.DetectionPhase.isin(['Field'])]
resultGroup = filteredData.groupby(['DetectionPhase', 'Status', 'Priority']).count()
LabelsReviewedByDate = resultGroup
dateLabelsFig = LabelsReviewedByDate.unstack().plot(kind='bar',stacked=True, color=['blue','yellow'])

filteredData = wrangled_issues_df[wrangled_issues_df.Priority.isin(['Critical','High'])]
filteredData = filteredData[filteredData.Status.isin(['Approved','InProgress'])]
filteredData = filteredData[filteredData.DetectionPhase.isin(['Field'])]
resultGroup = filteredData.groupby(['DetectionPhase', 'Status', 'Priority']).count()
LabelsReviewedByDate = resultGroup
dateLabelsFig = LabelsReviewedByDate.unstack().plot(kind='bar',stacked=True, color=['blue','yellow'])
```



Use Facebook/Prophet package (https://facebook.github.io/prophet/docs/quick_start.html) to forecast the following for our repo

```
In [16]: wrangled_issues_df.to_csv('SCM587SP19.csv', sep=',', encoding='utf-8', index=False)
```

```
In [17]: from fbprophet import Prophet
```

1. The day of the week maximum number of issues created

```
In [18]: df = pd.read_csv('SCM587SP19.csv')
df.head()
df = df.rename(columns={'created_at': 'ds', 'issue_number': 'y'})
print(df)
m = Prophet()
m.fit(df)
future = m.make_future_dataframe(periods=365)
future.tail()
forecast = m.predict(future)
forecast[['ds', 'yhat', 'yhat_lower', 'yhat_upper']].tail()
fig1 = m.plot(forecast)
fig2 = m.plot_components(forecast)
```

344	NaN	NaN	NaN
345	NaN	NaN	NaN
346	NaN	NaN	NaN
347	NaN	NaN	NaN
348	NaN	NaN	NaN
349	NaN	NaN	NaN
350	NaN	NaN	NaN
351	NaN	NaN	NaN
352	NaN	NaN	NaN
353	NaN	NaN	NaN
354	NaN	NaN	NaN
355	NaN	NaN	NaN

[356 rows x 17 columns]

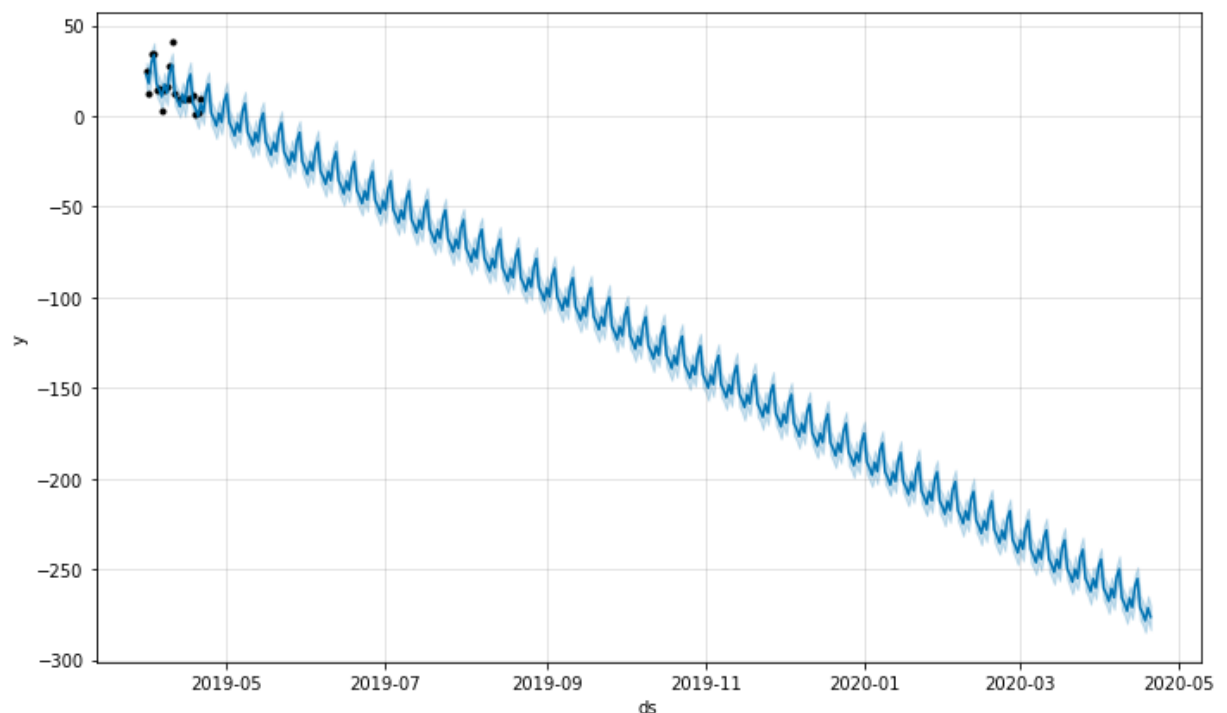


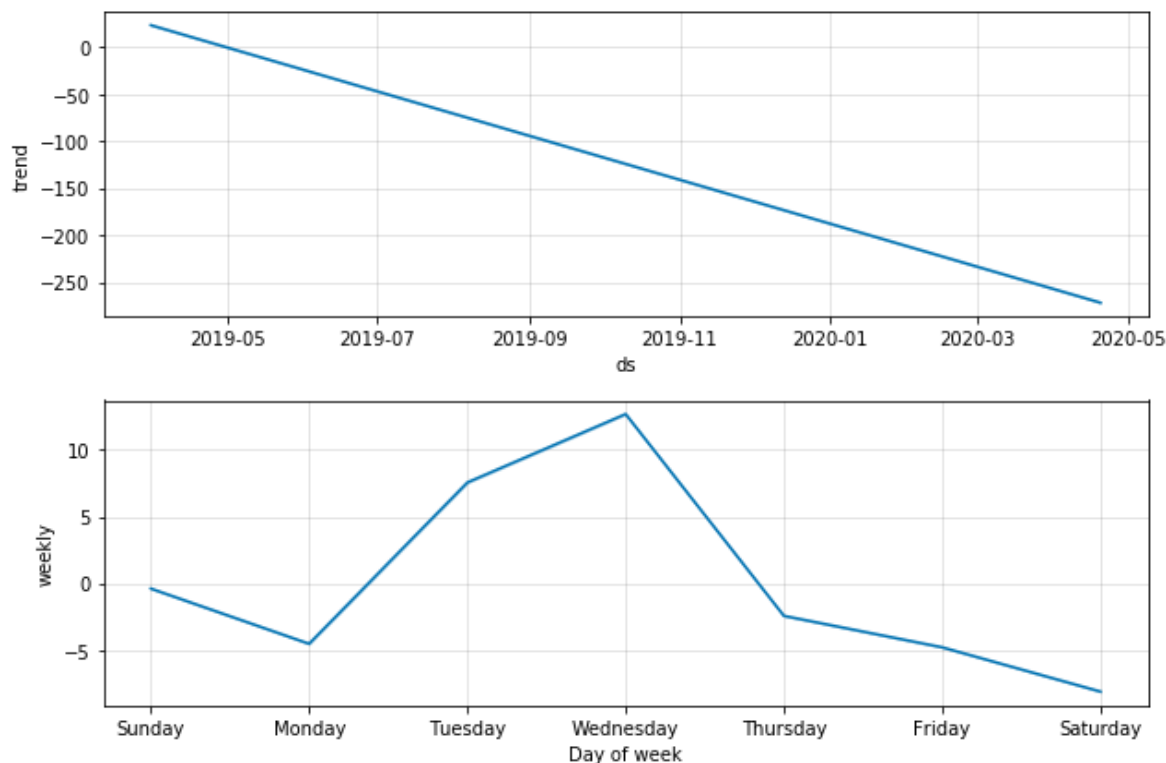
2. The day of the week maximum number of

issues closed

```
In [19]: df = pd.read_csv('SCM587SP19.csv')
df.head()
df_closed = issues_df['closed_at'].value_counts().rename_axis('ds').reset_index()
m = Prophet()
m.fit(df_closed)
future = m.make_future_dataframe(periods=365)
future.tail()
forecast = m.predict(future)
forecast[['ds', 'yhat', 'yhat_lower', 'yhat_upper']].tail()
fig1 = m.plot(forecast)
fig2 = m.plot_components(forecast)
```

C:\Users\17739\Anaconda3\lib\site-packages\fbprophet\forecaster.py:880: FutureWarning: Series.nonzero() is deprecated and will be removed in a future version. Use Series.to_numpy().nonzero() instead
 min_dt = dt.iloc[dt.nonzero()[0]].min()
INFO:fbprophet:Disabling yearly seasonality. Run prophet with yearly_seasonality=True to override this.
INFO:fbprophet:Disabling daily seasonality. Run prophet with daily_seasonality=True to override this.
INFO:fbprophet:n_changepoints greater than number of observations.Using 15.0.



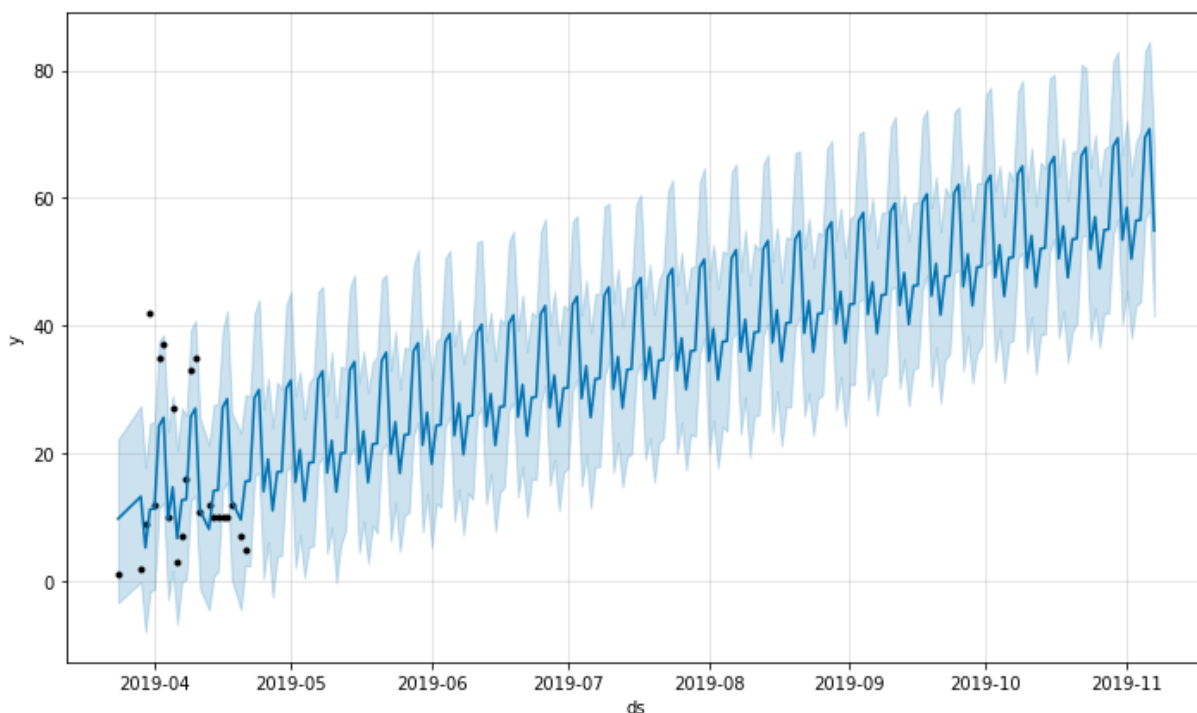


3. Plot the created issues forecast by calling the Prophet.plot method and passing in your forecast dataframe.

```
In [20]: def predict_plot(file):
m = Prophet()
df = pd.read_csv(file)
df_created = df['created_at'].value_counts().rename_axis('ds').reset_index(name='count')
m.fit(df_created)
future = m.make_future_dataframe(periods=200)
forecast = m.predict(future)
createdIssueForecast = m.plot(forecast)
```

```
In [21]: predict_plot('SCM587SP19.csv')
```

```
C:\Users\17739\Anaconda3\lib\site-packages\fbprophet\forecaster.py:880: FutureWarning: Series.nonzero() is deprecated and will be removed in a future version. Use Series.to_numpy().nonzero() instead
  min_dt = dt.iloc[dt.nonzero()[0]].min()
INFO:fbprophet:Disabling yearly seasonality. Run prophet with yearly_seasonality=True to override this.
INFO:fbprophet:Disabling daily seasonality. Run prophet with daily_seasonality=True to override this.
INFO:fbprophet:n_changepoints greater than number of observations.Using 17.0.
```



4. Plot the closed issues forecast; use the Prophet.plot_components method. By default you'll see the trend, yearly seasonality, and weekly seasonality of the time series. If you include holidays, you'll see those here, too.


```
In [22]: m = Prophet()
m.fit(df_closed)
future = m.make_future_dataframe(periods=365)
future.tail()
forecast = m.predict(future)
forecast[['ds', 'yhat', 'yhat_lower', 'yhat_upper']].tail()
closedIssueForecast = m.plot(forecast)
```

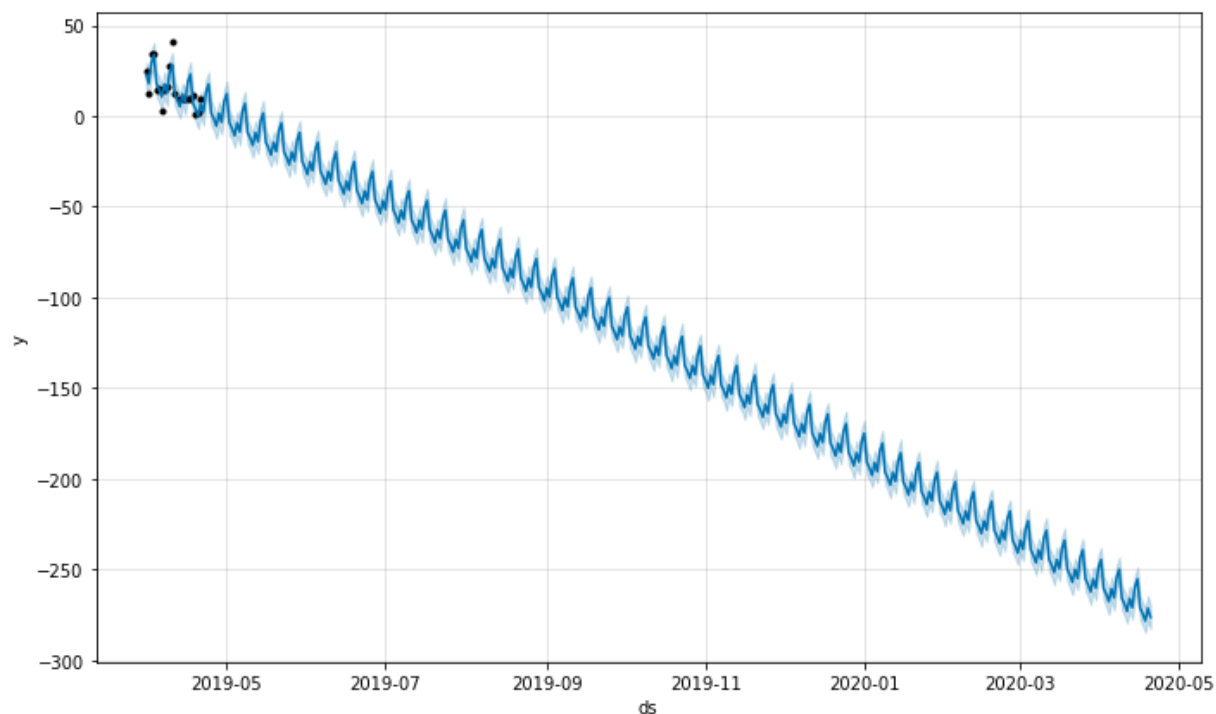
C:\Users\17739\Anaconda3\lib\site-packages\fbprophet\forecaster.py:880: FutureWarning: Series.nonzero() is deprecated and will be removed in a future version. Use Series.to_numpy().nonzero() instead

min_dt = dt.iloc[dt.nonzero()[0]].min()

INFO:fbprophet:Disabling yearly seasonality. Run prophet with yearly_seasonality=True to override this.

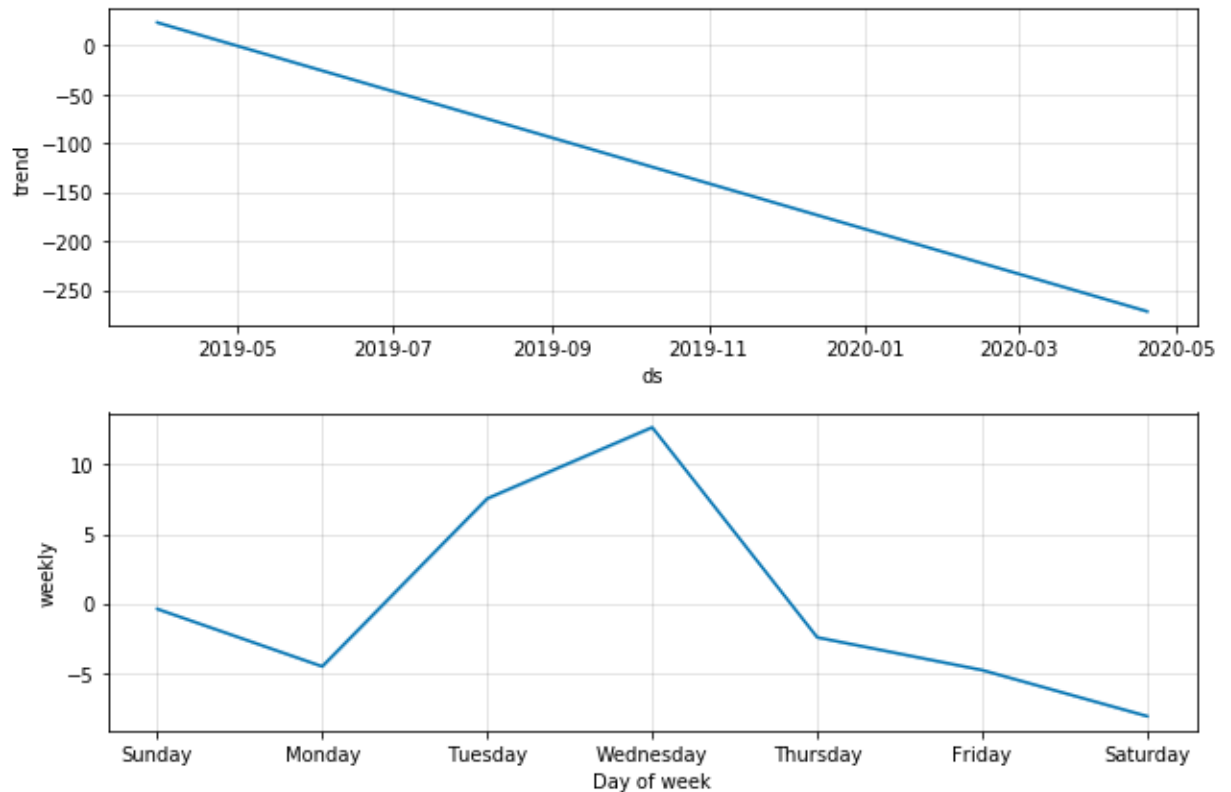
INFO:fbprophet:Disabling daily seasonality. Run prophet with daily_seasonality=True to override this.

INFO:fbprophet:n_changepoints greater than number of observations.Using 15.0.



5. Plot the pulls forecast; use the Prophet.plot_components method. By default you'll see the trend, yearly seasonality, and weekly seasonality of the time series. If you include holidays, you'll see those here, too.

```
In [23]: pullForecast = m.plot_components(forecast)
```



6. Plot the commits forecast; use the Prophet.plot_components method. By default you'll see the trend, yearly seasonality, and weekly seasonality of the time series. If you include holidays, you'll see those here, too.

```
In [24]: df_created = issues_df['created_at'].value_counts().rename_axis('ds').reset_index()
m = Prophet()
m.fit(df_created)
future = m.make_future_dataframe(periods=365)
future.tail()
forecast = m.predict(future)
forecast[['ds', 'yhat', 'yhat_lower', 'yhat_upper']].tail()
commitForecast = m.plot_components(forecast)
```

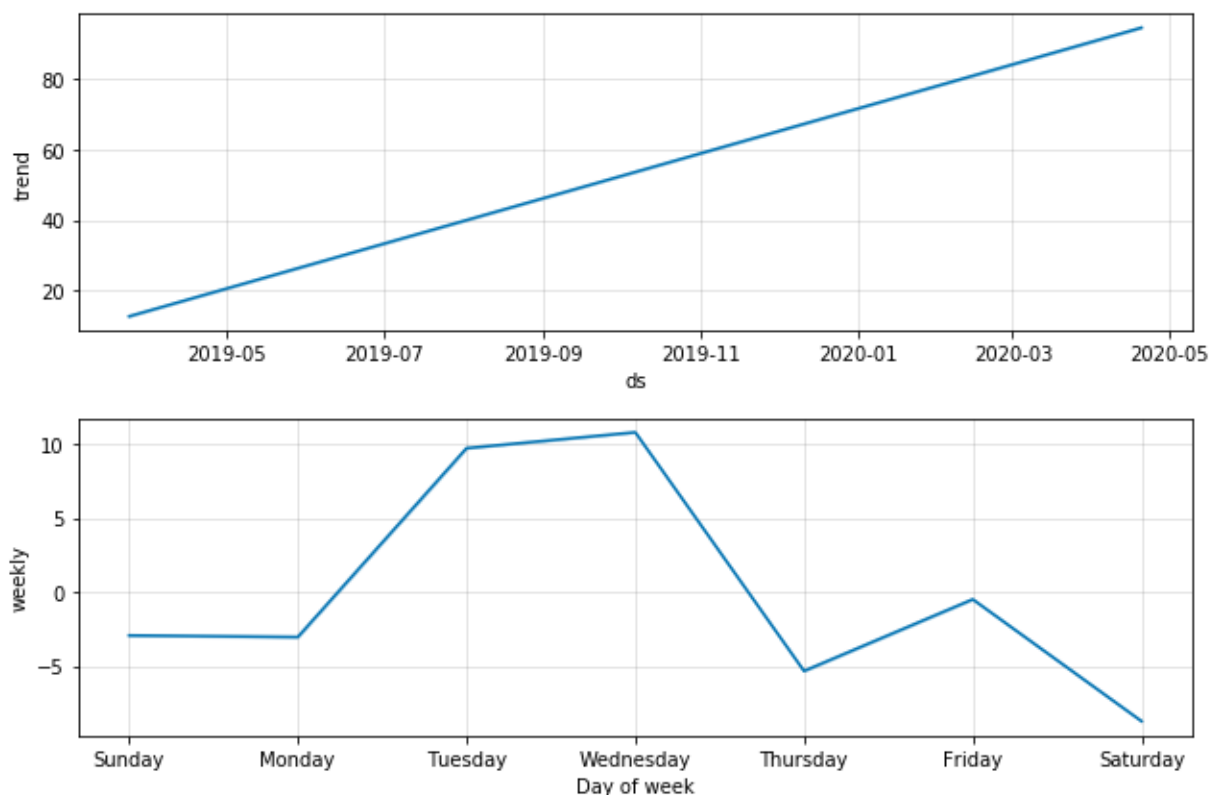
C:\Users\17739\Anaconda3\lib\site-packages\fbprophet\forecaster.py:880: FutureWarning: Series.nonzero() is deprecated and will be removed in a future version. Use Series.to_numpy().nonzero() instead

min_dt = dt.iloc[dt.nonzero()[0]].min()

INFO:fbprophet:Disabling yearly seasonality. Run prophet with yearly_seasonality=True to override this.

INFO:fbprophet:Disabling daily seasonality. Run prophet with daily_seasonality=True to override this.

INFO:fbprophet:n_changepoints greater than number of observations.Using 17.0.



Re-implement the above 6 requirements (listed for Facebook prophet package) using TensorFlow Time Series (TFTS)

<https://github.com/tensorflow/tensorflow/tree/master/tensorflow/contrib/timeseries>

```
In [25]: from __future__ import absolute_import
from __future__ import division
from __future__ import print_function

from os import path
import tempfile
import random
import numpy
import tensorflow as tf
from dateutil import parser
import statsmodels.api as sm
import statsmodels.formula.api as smf
```

```

In [26]: try:
import matplotlib # pylint: disable=g-import-not-at-top
matplotlib.use("TkAgg") # Need Tk for interactive plots.
from matplotlib import pyplot # pylint: disable=g-import-not-at-top
HAS_MATPLOTLIB = True
except ImportError:
# Plotting requires matplotlib, but the unit test running this code may
# execute in an environment without it (i.e. matplotlib is not a build
# dependency). We'd still like to test the TensorFlow-dependent parts of this
# example, namely train_and_predict.
HAS_MATPLOTLIB = False

def multivariate_train_and_sample(export_directory=None, training_steps=500):
    """Trains, evaluates, and exports a multivariate model."""
    estimator = tf.contrib.timeseries.StructuralEnsembleRegressor(
        periodicities=[], num_features=1)

    mys = wrangled_issues_df
    openarr = []
    for index, row in wrangled_issues_df.iterrows():
        if (row["State"] == 'open'):
            openarr.append(parser.parse(row['created_at']).weekday())

    openres = pd.DataFrame(openarr)
    openres.to_csv('RKTopen.csv', header=None)
    reader = tf.contrib.timeseries.CSVReader(
        'RKTopen.csv',
        column_names=('times', 'values'))
    train_input_fn = tf.contrib.timeseries.RandomWindowInputFn(
        # Larger window sizes generally produce a better covariance matrix.
        reader, batch_size=4, window_size=64)
    estimator.train(input_fn=train_input_fn, steps=training_steps)
    evaluation_input_fn = tf.contrib.timeseries.WholeDatasetInputFn(reader)
    current_state = estimator.evaluate(input_fn=evaluation_input_fn, steps=1)
    values = [current_state["observed"]]
    times = [current_state[tf.contrib.timeseries.FilteringResults.TIMES]]
    # Export the model so we can do iterative prediction and filtering without
    # reloading model checkpoints.
    if export_directory is None:
        export_directory = tempfile.mkdtemp()
    input_receiver_fn = estimator.build_raw_serving_input_receiver_fn()
    export_location = estimator.export_saved_model(export_directory,
                                                    input_receiver_fn)

    with tf.Graph().as_default():
        numpy.random.seed(1) # Make the example a bit more deterministic
        with tf.Session() as session:
            signatures = tf.saved_model.loader.load(
                session, [tf.saved_model.tag_constants.SERVING], export_location)
            for _ in range(7):
                current_prediction = (
                    tf.contrib.timeseries.saved_model_utils.predict_continuation(
                        continue_from=current_state, signatures=signatures,
                        session=session, steps=1))
                next_sample = numpy.random.multivariate_normal(
                    # Squeeze out the batch and series length dimensions (both 1).
                    mean=numpy.squeeze(current_prediction["mean"], axis=(0, 1)),

```

```

        cov=numpy.squeeze(current_prediction["covariance"], axis=(0, 1)))
    # Update model state so that future predictions are conditional on the
    # value we just sampled.
    filtering_features = {
        tf.contrib.timeseries.TrainEvalFeatures.TIMES: current_prediction[
            tf.contrib.timeseries.FilteringResults.TIMES],
        tf.contrib.timeseries.TrainEvalFeatures.VALUES: next_sample[
            None, None, :]}
    current_state = (
        tf.contrib.timeseries.saved_model_utils.filter_continuation(
            continue_from=current_state,
            session=session,
            signatures=signatures,
            features=filtering_features))
    values.append(next_sample[None, None, :])
    times.append(current_state["times"])
    all_observations = numpy.squeeze(numpy.concatenate(values, axis=1), axis=0)
    all_times = numpy.squeeze(numpy.concatenate(times, axis=1), axis=0)
    return all_times, all_observations

if not HAS_MATPLOTLIB:
    raise ImportError("Please install matplotlib to generate a plot from this example")
all_times, all_observations = multivariate_train_and_sample()

```

github.com/tensorflow/tensorboard-plugin-example/blob/master/README.md) for more information).

INFO:tensorflow:Summary for np.ndarray is not visible in Tensorboard by default. Consider using a Tensorboard plugin for visualization (see <https://github.com/tensorflow/tensorboard-plugin-example/blob/master/README.md> (<https://github.com/tensorflow/tensorboard-plugin-example/blob/master/README.md>) for more information).

INFO:tensorflow:Summary for np.ndarray is not visible in Tensorboard by default. Consider using a Tensorboard plugin for visualization (see <https://github.com/tensorflow/tensorboard-plugin-example/blob/master/README.md> (<https://github.com/tensorflow/tensorboard-plugin-example/blob/master/README.md>) for more information).

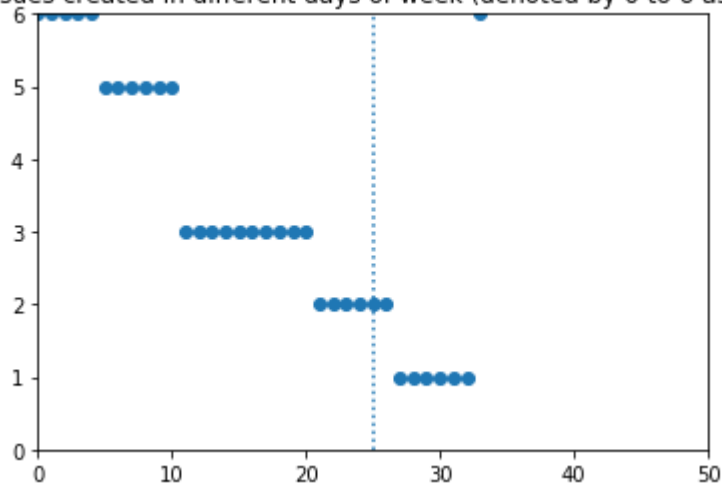
INFO:tensorflow:Summary for np.ndarray is not visible in Tensorboard by default. Consider using a Tensorboard plugin for visualization (see <https://github.com/tensorflow/tensorboard-plugin-example/blob/master/README.md> (<https://github.com/tensorflow/tensorboard-plugin-example/blob/master/README.md>) for more information).

WARNING:tensorflow:Skipping summary for start_tuple, must be a float, np.float32, np.int64, np.int32 or int or np.ndarray or a serialized string of Summary

1. The day of the week maximum number of issues created

```
In [27]: matplotlib.pyplot.ylim(0, 6)
matplotlib.pyplot.xlim(0, 50)
pyplot.axvline(25, linestyle="dotted")
pyplot.title('No of issues created in different days of week (denoted by 0 to 6 as Mon-Fri)')
pyplot.scatter(all_times, all_observations)
pyplot.show()
```

No of issues created in different days of week (denoted by 0 to 6 as Mon-Fri)



In [28]:

```

try:
    import matplotlib # pylint: disable=g-import-not-at-top
    matplotlib.use("TkAgg") # Need Tk for interactive plots.
    from matplotlib import pyplot # pylint: disable=g-import-not-at-top
    HAS_MATPLOTLIB = True
except ImportError:
    # Plotting requires matplotlib, but the unit test running this code may
    # execute in an environment without it (i.e. matplotlib is not a build
    # dependency). We'd still like to test the TensorFlow-dependent parts of this
    # example, namely train_and_predict.
    HAS_MATPLOTLIB = False

def multivariate_train_and_sample(export_directory=None, training_steps=500):
    """Trains, evaluates, and exports a multivariate model."""
    estimator = tf.contrib.timeseries.StructuralEnsembleRegressor(
        periodicities=[], num_features=1)

    mys = wrangled_issues_df
    closedarr = []
    for index, row in wrangled_issues_df.iterrows():
        if(row["State"] == 'closed'):
            closedarr.append(parser.parse(row['closed_at']).weekday())

    closedres = pd.DataFrame(closedarr)
    closedres.to_csv('RKTclosed.csv', header=None)
    reader = tf.contrib.timeseries.CSVReader(
        'RKTclosed.csv',
        column_names=('times', 'values'))
    train_input_fn = tf.contrib.timeseries.RandomWindowInputFn(
        # Larger window sizes generally produce a better covariance matrix.
        reader, batch_size=4, window_size=64)
    estimator.train(input_fn=train_input_fn, steps=training_steps)
    evaluation_input_fn = tf.contrib.timeseries.WholeDatasetInputFn(reader)
    current_state = estimator.evaluate(input_fn=evaluation_input_fn, steps=1)
    values = [current_state["observed"]]
    times = [current_state[tf.contrib.timeseries.FilteringResults.TIMES]]
    # Export the model so we can do iterative prediction and filtering without
    # reloading model checkpoints.
    if export_directory is None:
        export_directory = tempfile.mkdtemp()
    input_receiver_fn = estimator.build_raw_serving_input_receiver_fn()
    export_location = estimator.export_saved_model(export_directory,
                                                    input_receiver_fn)

    with tf.Graph().as_default():
        numpy.random.seed(1) # Make the example a bit more deterministic
        with tf.Session() as session:
            signatures = tf.saved_model.loader.load(
                session, [tf.saved_model.tag_constants.SERVING], export_location)
            for _ in range(7):
                current_prediction = (
                    tf.contrib.timeseries.saved_model_utils.predict_continuation(
                        continue_from=current_state, signatures=signatures,
                        session=session, steps=1))
                next_sample = numpy.random.multivariate_normal(
                    # Squeeze out the batch and series length dimensions (both 1).

```



```

        mean=numpy.squeeze(current_prediction["mean"], axis=(0, 1)),
        cov=numpy.squeeze(current_prediction["covariance"], axis=(0, 1)))
# Update model state so that future predictions are conditional on the
# value we just sampled.
    filtering_features = {
        tf.contrib.timeseries.TrainEvalFeatures.TIMES: current_prediction[
            tf.contrib.timeseries.FilteringResults.TIMES],
        tf.contrib.timeseries.TrainEvalFeatures.VALUES: next_sample[
            None, None, :]}
    current_state = (
        tf.contrib.timeseries.saved_model_utils.filter_continuation(
            continue_from=current_state,
            session=session,
            signatures=signatures,
            features=filtering_features))
    values.append(next_sample[None, None, :])
    times.append(current_state["times"])
    all_observations = numpy.squeeze(numpy.concatenate(values, axis=1), axis=0)
    all_times = numpy.squeeze(numpy.concatenate(times, axis=1), axis=0)
    return all_times, all_observations

if not HAS_MATPLOTLIB:
    raise ImportError("Please install matplotlib to generate a plot from this example")
all_times, all_observations = multivariate_train_and_sample()

```

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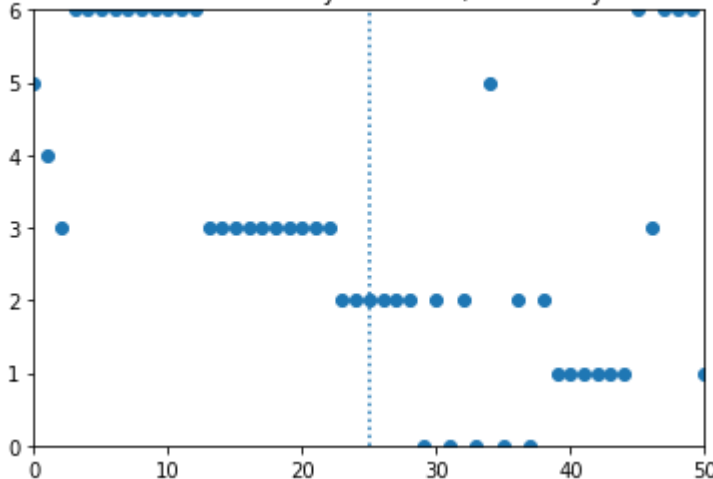
[6.]

[6.]

2. The day of the week maximum number of issues closed

```
In [29]: matplotlib.pyplot.ylim(0, 6)
matplotlib.pyplot.xlim(0, 50)
pyplot.axvline(25, linestyle="dotted")
pyplot.title('No of issues closed in different days of week (denoted by 0 to 6 as Mon-Fri)')
pyplot.scatter(all_times, all_observations)
pyplot.show()
```

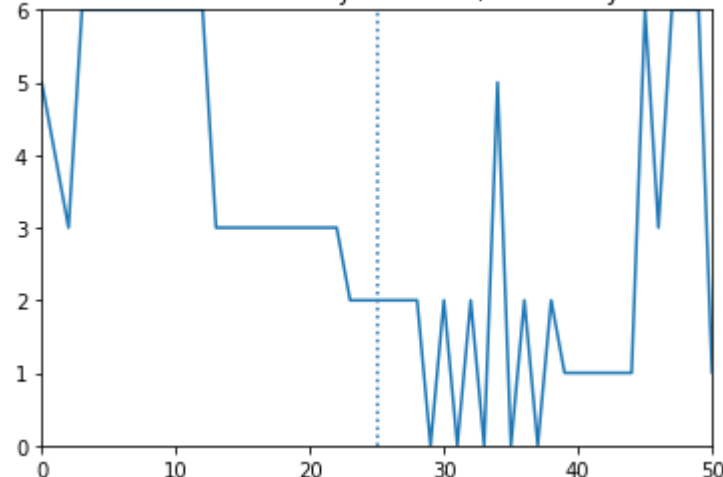
No of issues closed in different days of week (denoted by 0 to 6 as Mon-Fri)



3. Plot the created issues forecast by calling the Prophet.plot method and passing in your forecast dataframe.

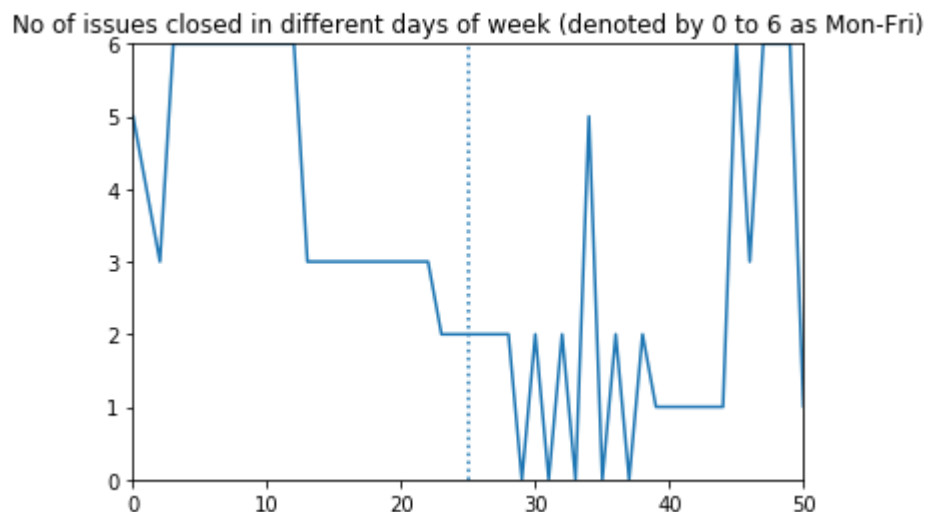
```
In [30]: matplotlib.pyplot.ylim(0, 6)
matplotlib.pyplot.xlim(0, 50)
pyplot.axvline(25, linestyle="dotted")
pyplot.title('No of issues created in different days of week (denoted by 0 to 6 as Mon-Fri)')
pyplot.plot(all_times, all_observations)
pyplot.show()
```

No of issues created in different days of week (denoted by 0 to 6 as Mon-Fri)



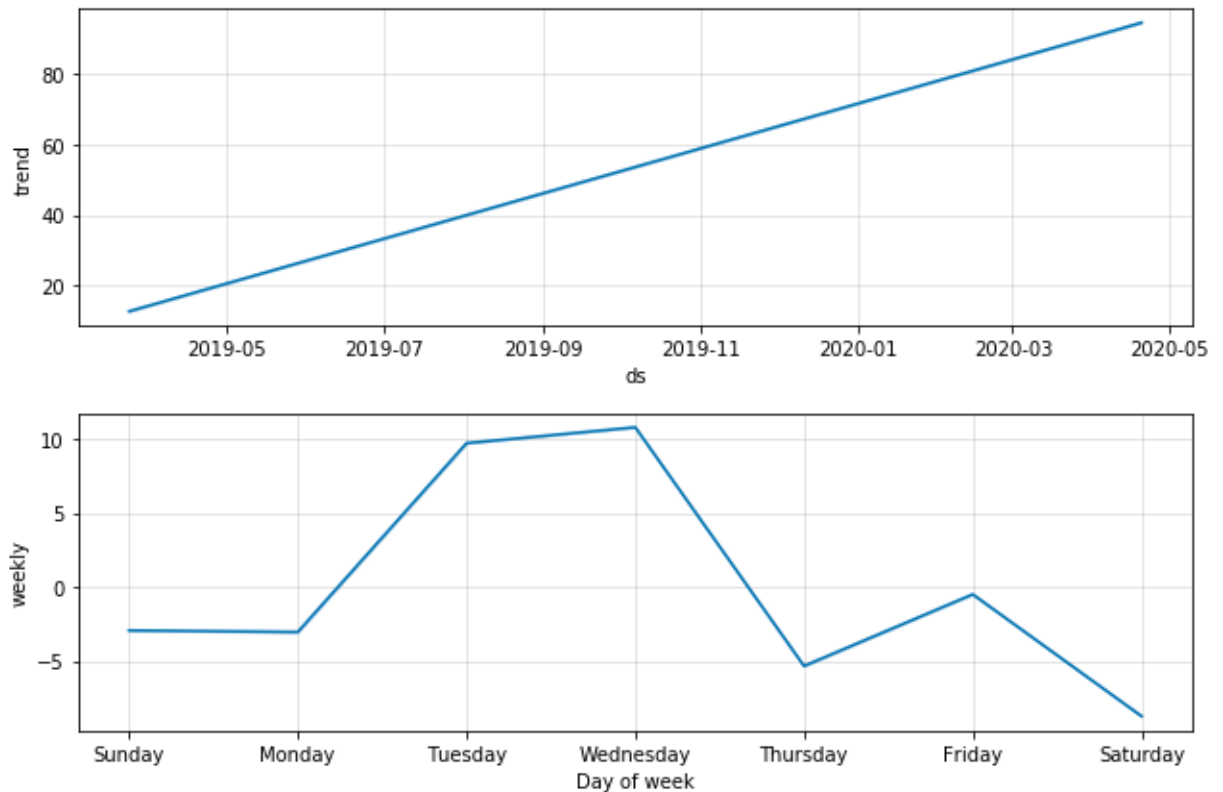
4. Plot the closed issues forecast; use the Prophet.plot_components method. By default you'll see the trend, yearly seasonality, and weekly seasonality of the time series. If you include holidays, you'll see those here, too.

```
In [31]: matplotlib.pyplot.ylim(0, 6)
matplotlib.pyplot.xlim(0, 50)
pyplot.axvline(25, linestyle="dotted")
pyplot.title('No of issues closed in different days of week (denoted by 0 to 6 as Mon-Fri)')
pyplot.plot(all_times, all_observations)
pyplot.show()
```



5. Plot the pulls forecast; use the Prophet.plot_components method. By default you'll see the trend, yearly seasonality, and weekly seasonality of the time series. If you include holidays, you'll see those here, too.

```
In [32]: pullForecast = m.plot_components(forecast)
```



6. Plot the commits forecast; use the Prophet.plot_components method. By default you'll see the trend, yearly seasonality, and weekly seasonality of the time series. If you include holidays, you'll see those here, too.

```
In [33]: df_created = issues_df['created_at'].value_counts().rename_axis('ds').reset_index()
m = Prophet()
m.fit(df_created)
future = m.make_future_dataframe(periods=365)
future.tail()
forecast = m.predict(future)
forecast[['ds', 'yhat', 'yhat_lower', 'yhat_upper']].tail()
commits_forecast = m.plot_components(forecast)
```

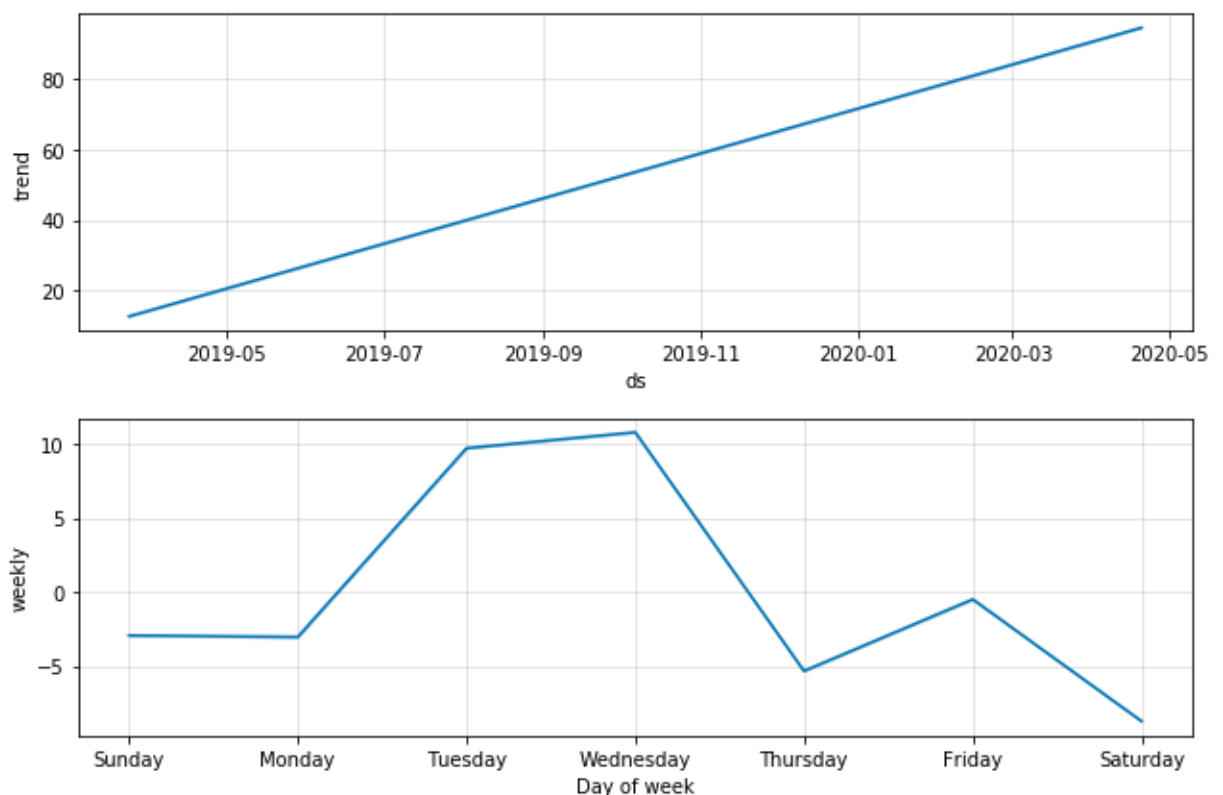
C:\Users\17739\Anaconda3\lib\site-packages\fbprophet\forecaster.py:880: FutureWarning: Series.nonzero() is deprecated and will be removed in a future version. Use Series.to_numpy().nonzero() instead

min_dt = dt.iloc[dt.nonzero()[0]].min()

INFO:fbprophet:Disabling yearly seasonality. Run prophet with yearly_seasonality=True to override this.

INFO:fbprophet:Disabling daily seasonality. Run prophet with daily_seasonality=True to override this.

INFO:fbprophet:n_changepoints greater than number of observations.Using 17.0.



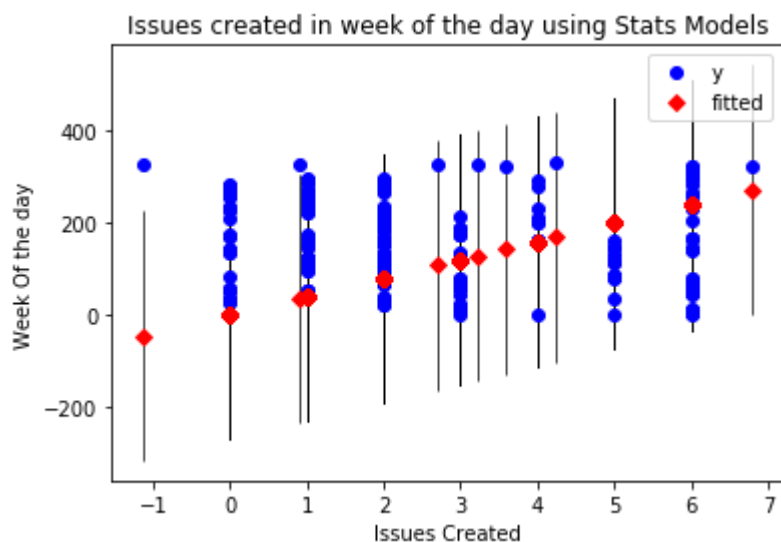
Re-implement the above 6 requirements (listed for Facebook prophet package) using StatsModel
: <https://www.statsmodels.org/stable/index.html>
(<https://www.statsmodels.org/stable/index.html>)

```
In [34]: import numpy as np
import statsmodels.api as sm
from datetime import datetime
```

1. The day of the week maximum number of issues created

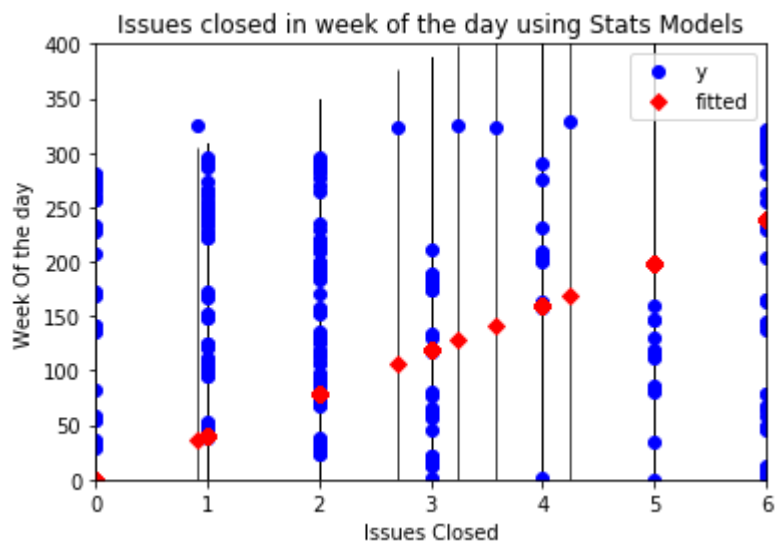
```
In [35]: model = sm.OLS(all_times, all_observations)
results = model.fit()
fig, ax = plt.subplots()
fig = sm.graphics.plot_fit(results, 0, ax=ax)
ax.set_ylabel("Week Of the day")
ax.set_xlabel("Issues Created")
ax.set_title("Issues created in week of the day using Stats Models")
```

Out[35]: Text(0.5, 1.0, 'Issues created in week of the day using Stats Models')



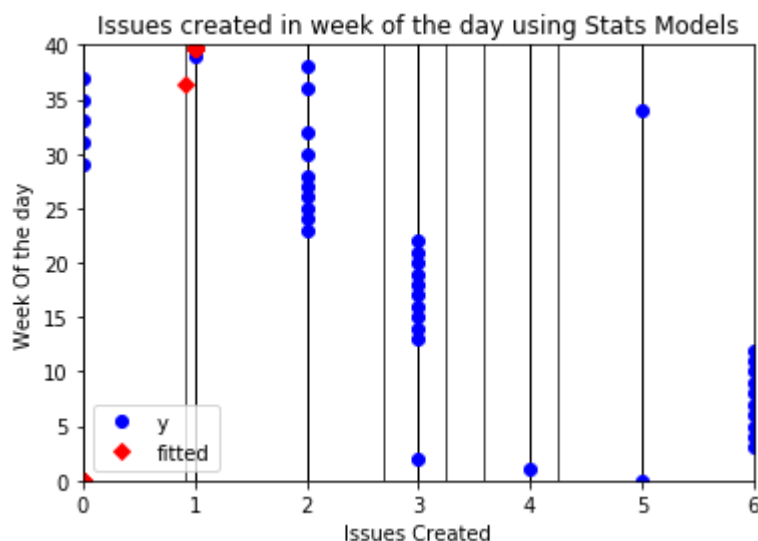
2. The day of the week maximum number of issues closed

```
In [36]: model = sm.OLS(all_times, all_observations)
results = model.fit()
fig, ax = plt.subplots()
fig = sm.graphics.plot_fit(results, 0, ax=ax)
ax.set_ylabel("Week Of the day")
ax.set_xlabel("Issues Closed")
ax.set_title("Issues closed in week of the day using Stats Models")
matplotlib.pyplot.ylim(0, 400)
matplotlib.pyplot.xlim(0, 6)
plt.show()
```



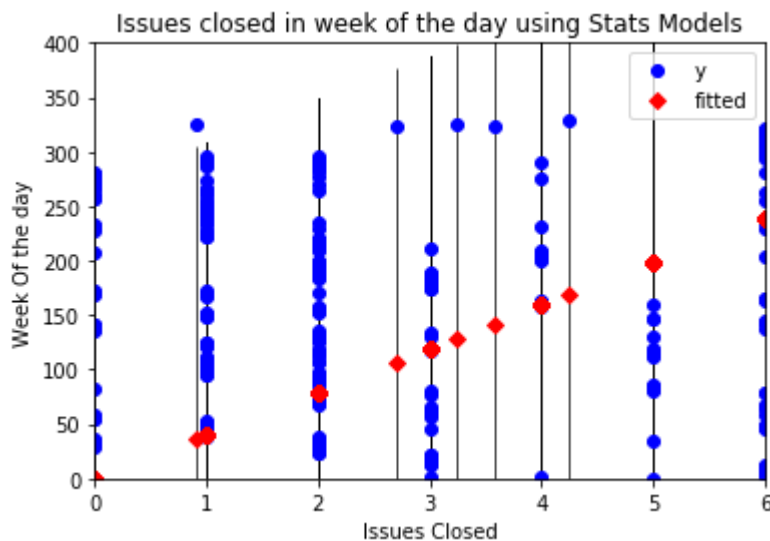
3. Plot the created issues forecast by calling the Prophet.plot method and passing in your forecast dataframe.

```
In [37]: model = sm.OLS(all_times, all_observations)
results = model.fit()
fig, ax = plt.subplots()
fig = sm.graphics.plot_fit(results, 0, ax=ax)
ax.set_ylabel("Week Of the day")
ax.set_xlabel("Issues Created")
ax.set_title("Issues created in week of the day using Stats Models")
matplotlib.pyplot.ylim(0, 40)
matplotlib.pyplot.xlim(0, 6)
plt.show()
```



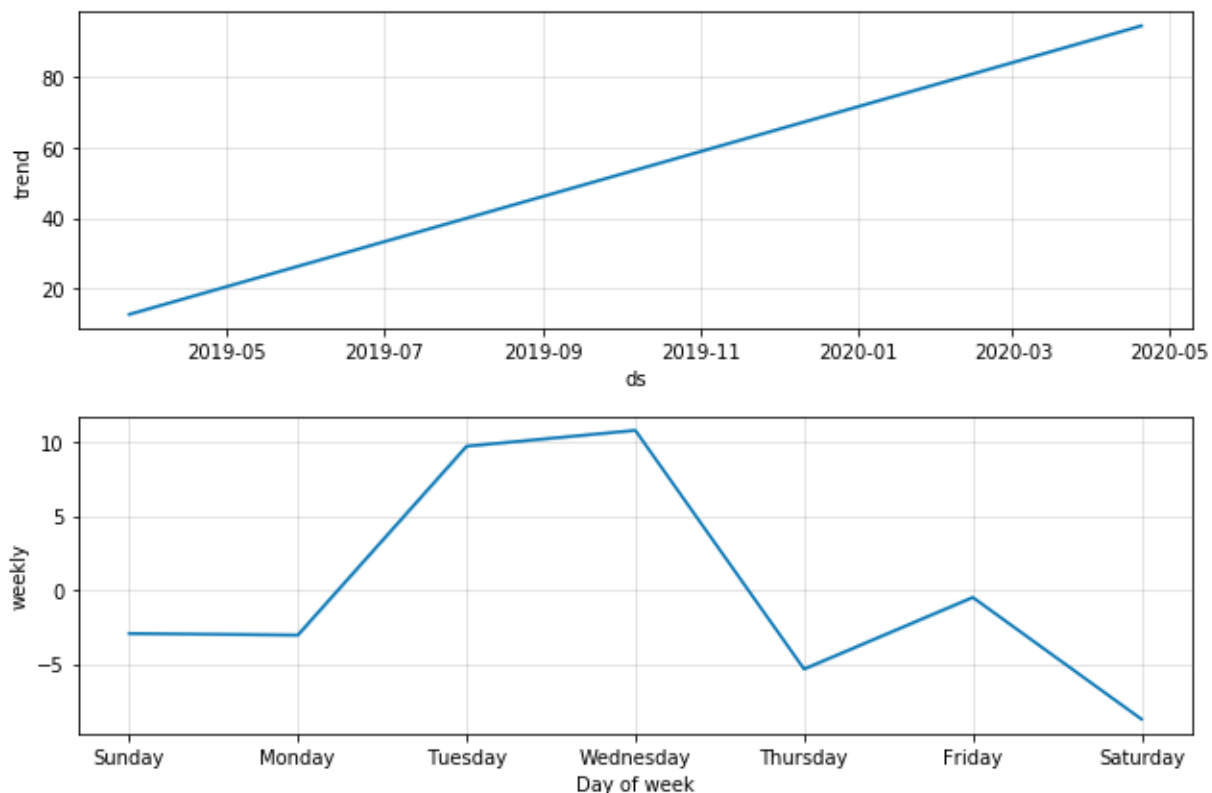
4. Plot the closed issues forecast; use the Prophet.plot_components method. By default you'll see the trend, yearly seasonality, and weekly seasonality of the time series. If you include holidays, you'll see those here, too.


```
In [38]: model = sm.OLS(all_times, all_observations)
results = model.fit()
fig, ax = plt.subplots()
fig = sm.graphics.plot_fit(results, 0, ax=ax)
ax.set_ylabel("Week Of the day")
ax.set_xlabel("Issues Closed")
ax.set_title("Issues closed in week of the day using Stats Models")
matplotlib.pyplot.ylim(0, 400)
matplotlib.pyplot.xlim(0, 6)
plt.show()
```



5. Plot the pulls forecast; use the Prophet.plot_components method. By default you'll see the trend, yearly seasonality, and weekly seasonality of the time series. If you include holidays, you'll see those here, too.

```
In [39]: pulls_forecast = m.plot_components(forecast)
```



6. Plot the commits forecast; use the Prophet.plot_components method. By default you'll see the trend, yearly seasonality, and weekly seasonality of the time series. If you include holidays, you'll see those here, too.

```
In [40]: df_created = issues_df['created_at'].value_counts().rename_axis('ds').reset_index()
m = Prophet()
m.fit(df_created)
future = m.make_future_dataframe(periods=365)
future.tail()
forecast = m.predict(future)
forecast[['ds', 'yhat', 'yhat_lower', 'yhat_upper']].tail()
commits_forecast = m.plot_components(forecast)
```

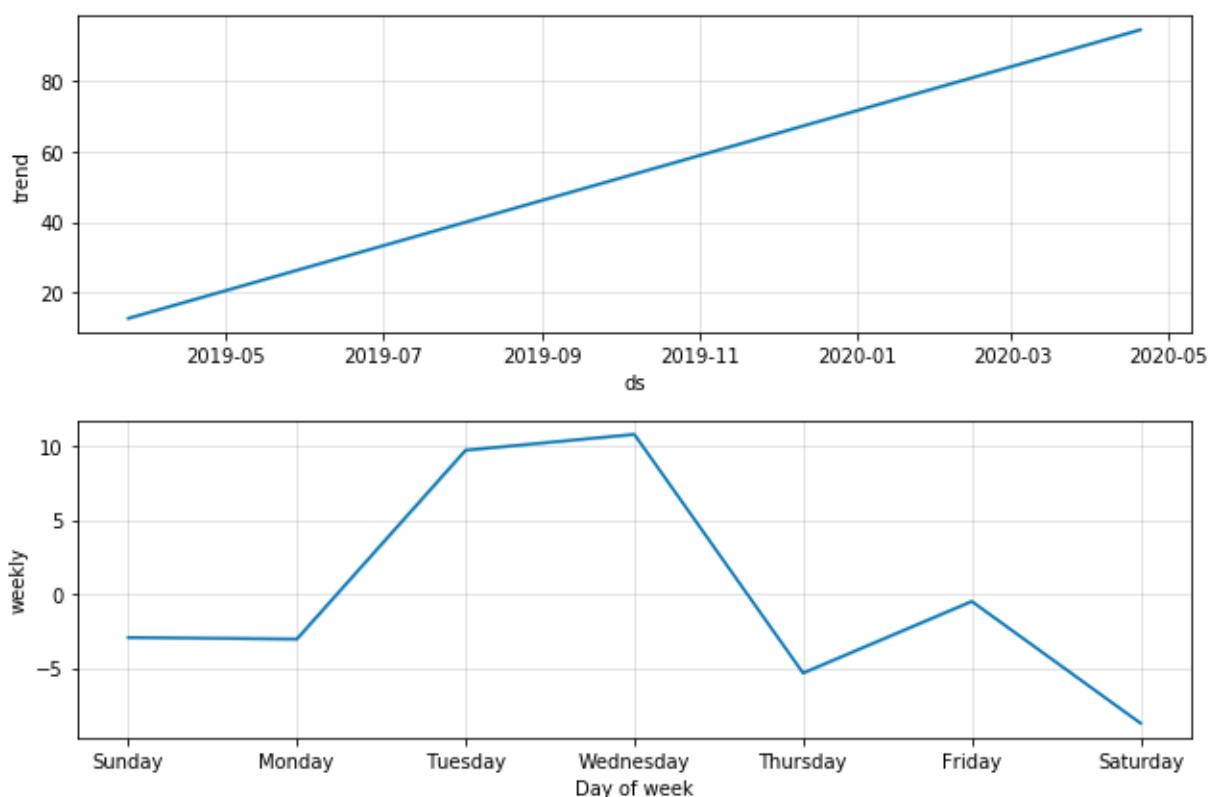
C:\Users\17739\Anaconda3\lib\site-packages\fbprophet\forecaster.py:880: FutureWarning: Series.nonzero() is deprecated and will be removed in a future version. Use Series.to_numpy().nonzero() instead

min_dt = dt.iloc[dt.nonzero()[0]].min()

INFO:fbprophet:Disabling yearly seasonality. Run prophet with yearly_seasonality=True to override this.

INFO:fbprophet:Disabling daily seasonality. Run prophet with daily_seasonality=True to override this.

INFO:fbprophet:n_changepoints greater than number of observations.Using 17.0.



In []: