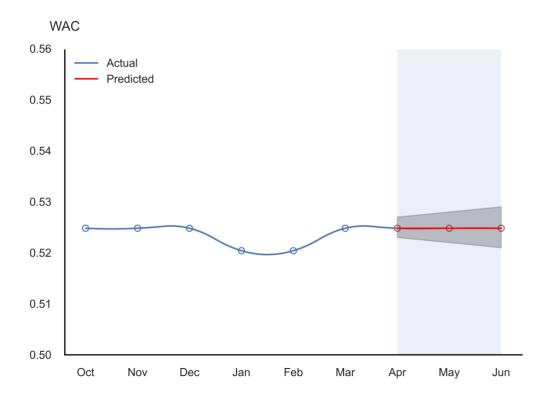
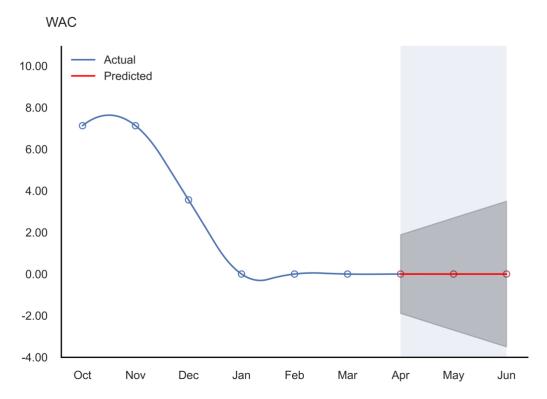
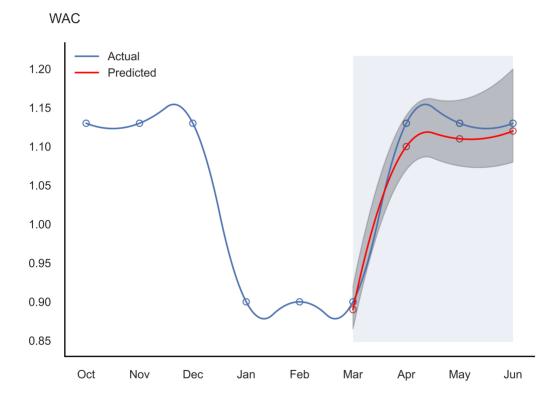
The codes on the next pages serve to build the following visualizations:

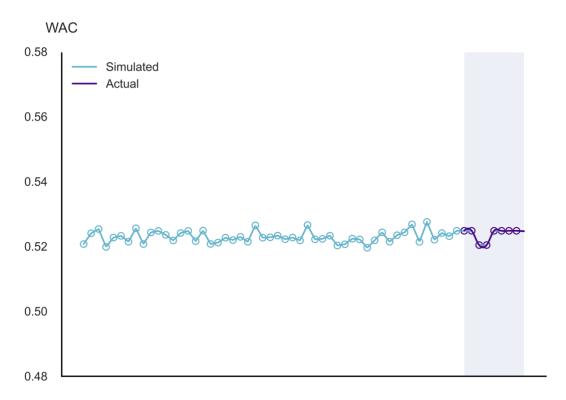
1) Depicting model performance of fitted ARIMA models:







2) Depicting data simulations



Data Visualizations

July 12, 2021

1 Visualising Model Performance

```
[]: def optimal_arima_build_viz(row, top_5_best_results_with_ci_df):
         X = np.array(row[2:58], dtype='float') #sim wac 1 to wac 56: used to train_
      \rightarrow model
         y = row[58:].reset_index(drop=True) #wac 57 to wac 59: to be predicted by
      \rightarrowmodel
         #optimal arima order is the one that gives the smallest ci width. try the
      →most optimal order onwards, be the most optimal order may throw an error for
      → cases where WAC doesnt vary at all
         sorted_df = top_5_best_results_with_ci_df.sort_values(by='next_months ci_
      →width', ascending=True)
         for i in range(len(sorted_df)):
             optimal_arima_order = sorted_df.iloc[i]['arima order']
             try:
                 model = ARIMA(X, order=optimal_arima_order)
                 model_fit = model.fit()
                 y_pred_df = model_fit.get_forecast(steps=3).summary_frame(alpha=0.
      →05)
                 y_pred_df = y_pred_df[['mean', 'mean_ci_lower', 'mean_ci_upper', |

    'mean_se']]

                 y_pred_df.fillna('NA', inplace=True)
                 if y_pred_df['mean_ci_lower'] != 'NA':
                 else:
                     continue
             except:
                 continue
         y_results = pd.concat([y, y_pred_df, pd.Series(f'{optimal_arima_order}')],__
      ⇒axis=1)
```

```
1.1 1) WAC stays constant at some positive value: 395 cases (52%)
[]: row = wac df for model.iloc[452:453]
    row
[]: top_5_best_results_with_ci_df = try_combinations(wac_df_for_model.iloc[452])
    top_5_best_results_with_ci_df
[]: y_results_452 = optimal_arima_build_viz(wac_df_for_model.iloc[452],_
     →top_5_best_results_with_ci_df)
    y_results_452.index = [6, 7, 8]
    y_results_452
[]: actual_y_452 = wac_df_for_model.iloc[452][52:].reset_index(drop=True)
    actual_y_452
[]: import matplotlib.pyplot as plt
    plt.style.use('seaborn')
    from matplotlib.ticker import FormatStrFormatter
[]: #create a smooth curve plot from the data
    from scipy.interpolate import make interp spline
    spline = make_interp_spline(actual_y_452.index, list(actual_y_452), k=2)
    X = np.linspace(0, 8, 500)
    Y = spline(X)
    plt.plot(X, Y, label='Actual')
    #plot the actual datapoints too as scatter points
    actual_y_452_scatter = actual_y_452[:6]
    plt.scatter(actual_y_452_scatter.index, actual_y_452_scatter,_

→facecolors='none', edgecolors='b')
    #plot predicted WAC data and its scatter points
    y_results_452['predicted'].plot(label='Predicted', color='red')
    plt.scatter(y_results_452['predicted'].index, y_results_452['predicted'],u

→facecolors='none', edgecolors='r')
```

```
#shaded area for SD
ax = plt.gca() #qet current plot's axes
ax.fill_between([6, 7, 8],
               y_results_452['prediction_lower'],
               y_results_452['prediction_upper'],
               color='k', alpha=.25)
#format legend location
plt.legend(loc='upper left')
#y ticks
plt.ylim(1.649, 1.651)
ax.yaxis.set_major_formatter(FormatStrFormatter('%.2f')) #set y ticks to 4dp
#blue shaded area to signal prediction starting
ax.fill_betweenx(ax.get_ylim(), x1=6, x2=8, alpha=0.1)
#x ticks
plt.xticks([0,1,2,3,4,5,6,7,8], ['Oct', 'Nov', 'Dec', 'Jan', 'Feb', 'Mar', __
#format y axis title
plt.ylabel('WAC', rotation=0, fontsize=13)
ax.yaxis.set_label_coords(0,1.05)
#format y axis and x axis lines
ax.spines['left'].set_color('black')
ax.spines['bottom'].set_color('black')
plt.savefig('1.png', dpi=1000)
```

1.2 2) WAC changes between different positive values: 269 cases (35%)

```
[]: actual_y_0 = wac_df_for_model.iloc[0][52:].reset_index(drop=True)
    actual_y_0
[]: #create a smooth curve plot from the data
    from scipy.interpolate import make interp spline
    spline = make_interp_spline(actual_y_0.index, list(actual_y_0), k=2)
    X = np.linspace(0, 8, 500)
    Y = spline(X)
    plt.plot(X, Y, label='Actual')
    #plot the actual datapoints too as scatter points
    actual_y_0_scatter = actual_y_0[:6]
    plt.scatter(actual_y_0_scatter.index, actual_y_0_scatter, facecolors='none',_
     →edgecolors='b')
    #plot predicted WAC data and its scatter points
    y results 0['predicted'].plot(label='Predicted', color='red')
    plt.scatter(y_results_0['predicted'].index, y_results_0['predicted'],__

¬facecolors='none', edgecolors='r')
    #shaded area for SD
    ax = plt.gca() #get current plot's axes
    ax.fill_between([6, 7, 8],
                    y_results_0['prediction_lower'],
                    y_results_0['prediction_upper'],
                    color='k', alpha=.25)
     #format legend location
    plt.legend(loc='upper left')
    #y ticks
    plt.ylim(0.50, 0.56)
    ax.yaxis.set_major_formatter(FormatStrFormatter('%.2f')) #set y ticks to 4dp
    #blue shaded area to signal prediction starting
    ax.fill_betweenx(ax.get_ylim(), x1=6, x2=8, alpha=0.1)
    #x ticks
    plt.xticks([0,1,2,3,4,5,6,7,8], ['Oct', 'Nov', 'Dec', 'Jan', 'Feb', 'Mar', |
     #format y axis title
    plt.ylabel('WAC', rotation=0, fontsize=13)
    ax.yaxis.set_label_coords(0,1.05)
    #format y axis and x axis lines
    ax.spines['left'].set_color('black')
```

```
ax.spines['bottom'].set_color('black')
plt.savefig('2.png', dpi=1000)
```

```
1.3 3) WAC varies a LOT: 59 cases (8%)
[]: row = wac_df_for_model.iloc[612:613]
     row
[]: top_5 best_results_with ci_df = try_combinations(wac_df_for_model.iloc[612])
     top_5_best_results_with_ci_df
[]: y_results_612 = optimal_arima_build_viz(wac_df_for_model.iloc[612],_
     →top_5_best_results_with_ci_df)
     y_results_612.index = [6, 7, 8]
[]: #create a smooth curve plot from the data
    from scipy.interpolate import make_interp_spline
     spline = make_interp_spline(actual_y_612.index, list(actual_y_612), k=2)
     X = np.linspace(0, 8, 500)
     Y = spline(X)
     plt.plot(X, Y, label='Actual')
     #plot the actual datapoints too as scatter points
     actual_y_612_scatter = actual_y_612[:6]
     plt.scatter(actual_y_612_scatter.index, actual_y_612_scatter,_

→facecolors='none', edgecolors='b')
     #plot predicted WAC data and scatter points
     y_results_612['predicted'].plot(label='Predicted', color='red')
     plt.scatter(y_results_612['predicted'].index, y_results_612['predicted'],u
     →facecolors='none', edgecolors='r')
     #shaded area for SD
     ax = plt.gca() #get current plot's axes
     ax.fill_between([6, 7, 8],
                    y_results_612['prediction_lower'],
                     y_results_612['prediction_upper'],
                     color='k', alpha=.25)
     #y ticks
     plt.ylim(-4, 11)
     ax.yaxis.set_major_formatter(FormatStrFormatter('%.2f')) #set y ticks to 4dp
```

2 4) WAC changes between the prediction months

```
[]: row = wac_df_for_model.loc[1075:1076]
     row
[]: top_5_best_results_with_ci_df = try_combinations(wac_df_for_model.loc[1075])
     top_5_best_results_with_ci_df
[]: y_results_1075 = optimal_arima_build_viz(wac_df_for_model.loc[1075],_
     →top_5_best_results_with_ci_df)
     y_results_1075.index = [5, 6, 7, 8]
[]: #create a smooth curve plot from the data
     from scipy.interpolate import make_interp_spline
     spline = make_interp_spline(actual_y_1075.index, list(actual_y_1075), k=2)
     X = np.linspace(0, 8, 500)
     Y = spline(X)
     plt.plot(X, Y, label='Actual')
     #plot the actual datapoints too as scatter points
     plt.scatter(actual_y_1075.index, actual_y_1075, facecolors='none',_
     →edgecolors='b')
```

```
#plot predicted WAC data and scatter points
spline = make_interp_spline(y_results_1075.index,__
→list(y_results_1075['predicted']), k=2)
X = np.linspace(5, 8, 500)
Y = spline(X)
plt.plot(X, Y, label='Predicted', color='red')
plt.scatter(y_results_1075['predicted'].index, y_results_1075['predicted'],_u

¬facecolors='none', edgecolors='r')
#shaded area for SD
spline = make_interp_spline(y_results_1075.index,_
X = np.linspace(5, 8, 500)
Y_pred_lower = spline(X)
spline = make_interp_spline(y_results_1075.index,__
→y_results_1075['prediction_upper'], k=2)
X = np.linspace(5, 8, 500)
Y_pred_upper = spline(X)
ax = plt.gca() #get current plot's axes
ax.fill_between(X,
               Y_pred_lower,
               Y_pred_upper,
               color='k', alpha=.25)
#blue shaded area to signal prediction starting
ax.fill_betweenx(ax.get_ylim(), x1=5, x2=8, alpha=0.1)
#format legend location
plt.legend(loc='upper left')
#x ticks
plt.xticks([0,1,2,3,4,5,6,7,8], ['Oct', 'Nov', 'Dec', 'Jan', 'Feb', 'Mar', u
#y ticks
ax.yaxis.set_major_formatter(FormatStrFormatter('%.2f')) #set y ticks to 4dp
#format y axis title
plt.ylabel('WAC', rotation=0, fontsize=13)
ax.yaxis.set_label_coords(0,1.05)
#format y axis and x axis lines
ax.spines['left'].set_color('black')
```

```
ax.spines['bottom'].set_color('black')

#save fig
plt.savefig('4.png', dpi=1000)
```

3 Visualising Data Simulation

```
[]: data = wac_df_for_model.iloc[0][2:]
[]: len(data)
[]: #plot simulated data
     spline = make_interp_spline(range(59), list(data), k=1)
     X = np.linspace(0, 59, 500)
     Y = spline(X)
     plt.plot(X, Y, label='Simulated', color='c')
     plt.scatter(range(59), data, facecolors='none', edgecolors='c')
     #plot actual data
     spline = make_interp_spline(range(51, 59), list(data[51:59]), k=2)
     X = np.linspace(51, 59, 200)
     Y = spline(X)
     plt.plot(X, Y, label='Actual', color='indigo')
     plt.scatter(range(51, 59), data[51:59], facecolors='none', edgecolors='indigo')
     #change y axis
     plt.ylim(0.48, 0.58)
     #blue shaded area to signal actual value starting
     ax = plt.gca() #get current plot's axes
     ax.fill_betweenx(ax.get_ylim(), x1=51, x2=59, alpha=0.1)
     #format legend location
     plt.legend(loc='upper left')
     #x ticks: remove
     plt.xticks([], [])
     #format y axis title
     plt.ylabel('WAC', rotation=0, fontsize=13)
     ax.yaxis.set_label_coords(0,1.05)
     #format y axis and x axis lines
```

```
ax.spines['left'].set_color('black')
ax.spines['bottom'].set_color('black')

#save fig
plt.savefig('simulation.png', dpi=1000)
```