

# This Jupyter Notebook contains the visualisations

Simulation models were created in simul8 and ROI distributions were done using AtRisk in Excel

```
In [1]: import pandas as pd
import matplotlib.pyplot as plt
import numpy as np
```

```
In [3]: df0 = pd.read_excel('C:/Users/123/Desktop/OA/DataSets/Current_Scenerio.xlsx')
queue_cols0 = ['CSI Queue', 'Data Input Queue', 'Transport Queue', 'Sample Prep.1',
avg_queues0 = df0[queue_cols0].mean()

df1 = pd.read_excel('C:/Users/123/Desktop/OA/DataSets/s1.xlsx')
queue_cols1 = ['CSI_Sample_Collection_Queue', 'Data_Input_Queue', 'Transport_Queue'
avg_queues1 = df1[queue_cols1].mean()

df2 = pd.read_excel('C:/Users/123/Desktop/OA/DataSets/S2.xlsx')
queue_cols2 = ['CSI_Sample_Collection_Queue_Time', 'Courier_Collection_Queue_Time',
avg_queues2 = df2[queue_cols2].mean()

df3 = pd.read_excel('C:/Users/123/Desktop/OA/DataSets/S3.xlsx')
queue_cols3 = ['Time_in_CSI_Queue', 'Time_in_Courier_Queue', 'Time_in_Sample_Prep_Q
avg_queues3 = df3[queue_cols3].mean()

df4 = pd.read_excel('C:/Users/123/Desktop/OA/DataSets/S4.xlsx')
queue_cols4 = ['CSI_Queue_Time', 'Courier_Queue_Time', 'Sample_Prep_Queue_Time', 'R
avg_queues4 = df4[queue_cols4].mean()

df5 = pd.read_excel('C:/Users/123/Desktop/OA/DataSets/S5.xlsx')
queue_cols5 = ['CSI_Queue_Time', 'Courier_Queue_Time', 'Sample_Prep_Queue_Time', 'R
avg_queues5 = df5[queue_cols5].mean()
```

## Current Scenerio

```
In [9]: all_data = pd.concat([df0[col] for col in queue_cols0])
xlim = (0, all_data.max().max() * 1.1) # Add 10% margin
ylim = (0, 1.1 * max(df0[col].value_counts().max() for col in queue_cols0))

fig, axs = plt.subplots(nrows=3, ncols=3, figsize=(16, 12))
axs = axs.flatten()
```

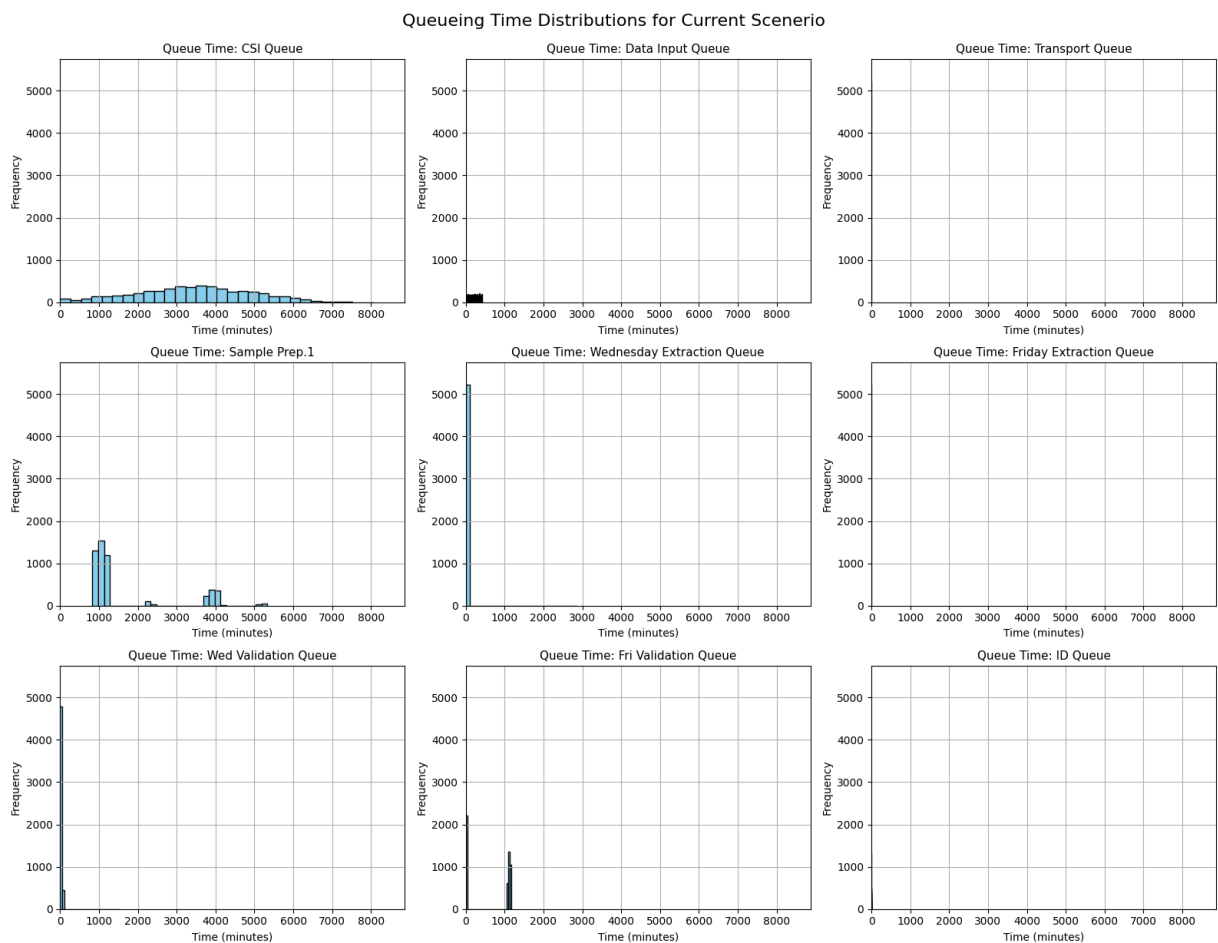
```

for i, col in enumerate(queue_cols0):
    axs[i].hist(df0[col].dropna(), bins=30, edgecolor='black', color='skyblue')
    axs[i].set_title(f'Queue Time: {col}', fontsize=11)
    axs[i].set_xlim(xlim)
    axs[i].set_ylim(ylim)
    axs[i].set_xlabel('Time (minutes)')
    axs[i].set_ylabel('Frequency')
    axs[i].grid(True)

# Hide any unused subplots (e.g. if only 8 plots)
for j in range(len(queue_cols0), len(axs)):
    fig.delaxes(axs[j])

plt.tight_layout()
plt.suptitle('Queueing Time Distributions for Current Scenerio', fontsize=16, y=1.0)
plt.show()

```



```

In [10]: fig, axs = plt.subplots(3, 3, figsize=(18, 12))
         axs = axs.flatten()

         for i, col in enumerate(queue_cols0):
             data = df0[col].dropna()

             # Clip to 95th percentile to avoid huge x-axis spread
             max_clip = data.quantile(0.95)
             data_clipped = data[data <= max_clip]

```

```

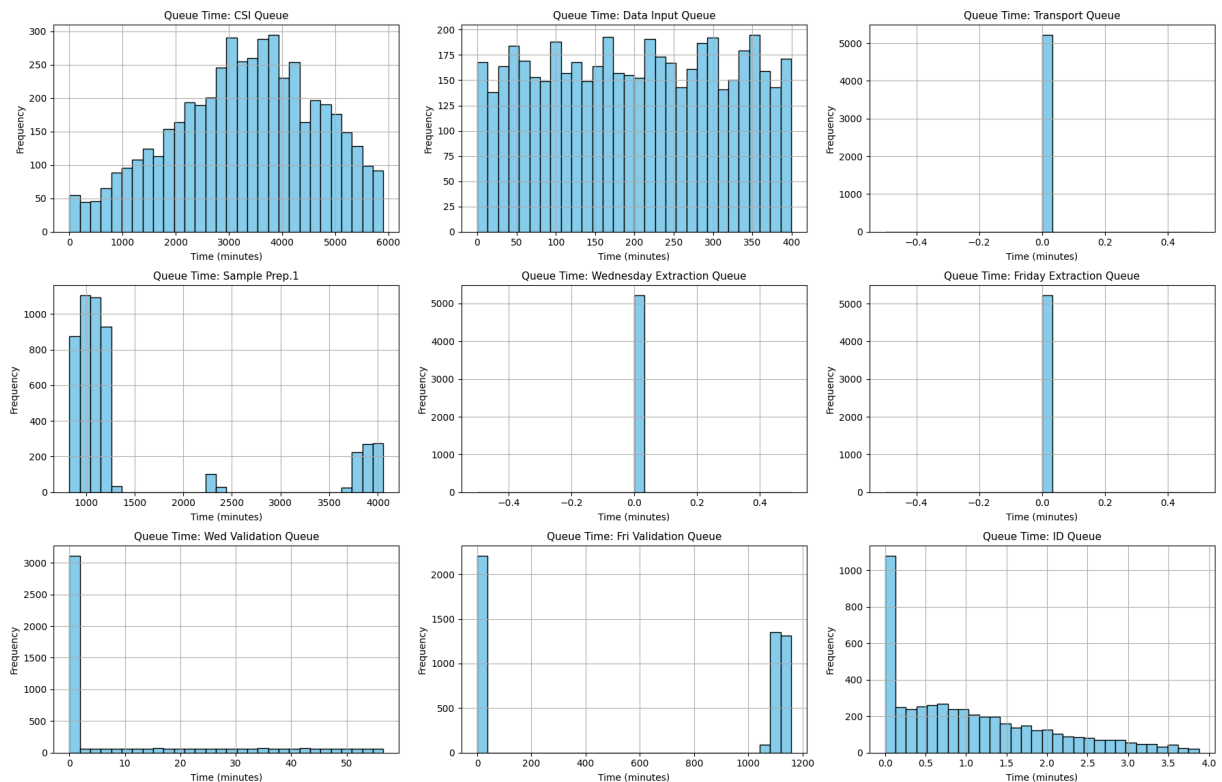
    axs[i].hist(data_clipped, bins=30, edgecolor='black', color='skyblue')
    axs[i].set_title(f'Queue Time: {col}', fontsize=11)
    axs[i].set_xlabel('Time (minutes)')
    axs[i].set_ylabel('Frequency')
    axs[i].grid(True)

# Hide unused subplots
for j in range(len(queue_cols0), len(axs)):
    fig.delaxes(axs[j])

plt.suptitle('Queueing Time Distributions for Current Scenerio(Scaled)', fontsize=
plt.tight_layout()
plt.show()

```

Queueing Time Distributions for Current Scenerio(Scaled))



## Scenerio 1

```

In [12]: all_values = pd.concat([df1[col].dropna() for col in queue_cols1])
xlim = (0, all_values.max() * 1.1)
ylim = (0, max(df1[col].value_counts().max() for col in queue_cols1) * 1.1)

fig, axs = plt.subplots(3, 3, figsize=(18, 12)) # 3x3 for 9 stages
axs = axs.flatten()

for i, col in enumerate(queue_cols1):
    axs[i].hist(df1[col].dropna(), bins=30, edgecolor='black', color='skyblue')
    axs[i].set_title(f'{col}', fontsize=11)
    axs[i].set_xlim(xlim)
    axs[i].set_ylim(ylim)

```

```

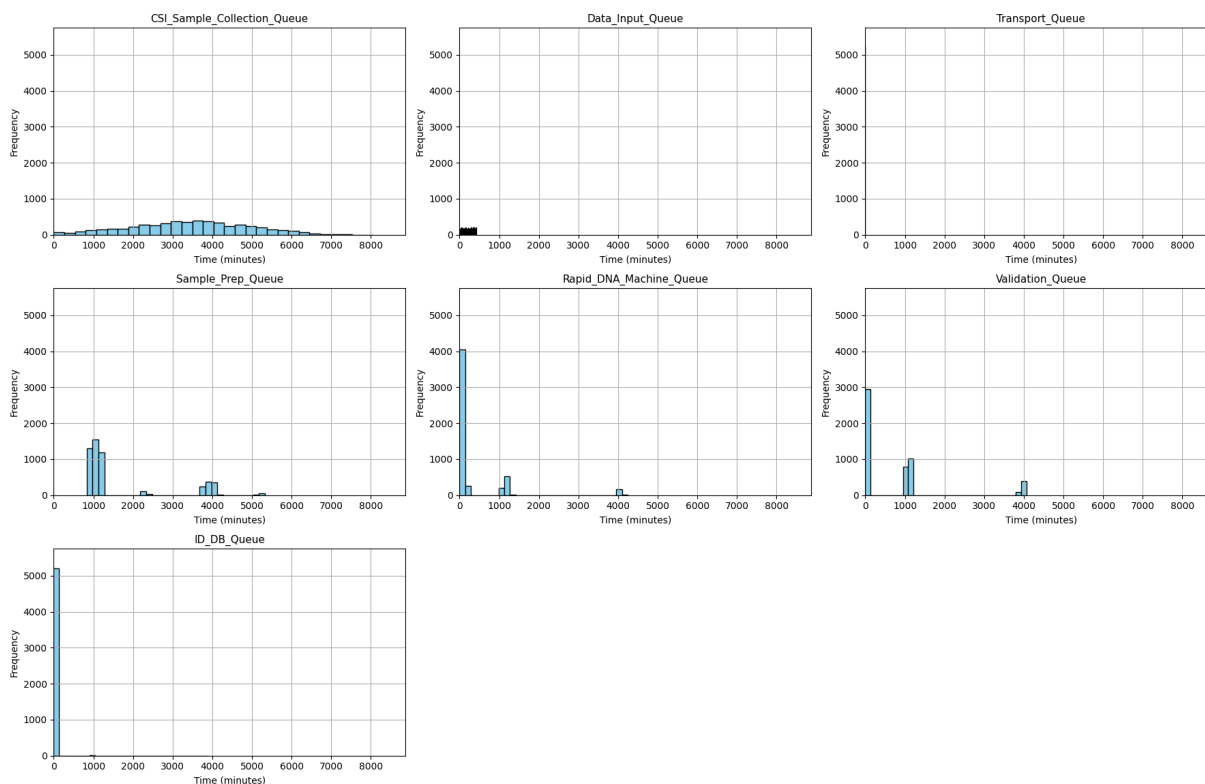
    axs[i].set_xlabel('Time (minutes)')
    axs[i].set_ylabel('Frequency')
    axs[i].grid(True)

# Hide any extra subplot boxes (if fewer than 9 columns)
for j in range(len(queue_cols1), len(axs)):
    fig.delaxes(axs[j])

plt.suptitle('Queueing Time Distributions for Scenerio 1', fontsize=18, y=1.02)
plt.tight_layout()
plt.show()

```

Queueing Time Distributions for Scenerio 1



```

In [13]: fig, axs = plt.subplots(3, 3, figsize=(18, 12))
        axs = axs.flatten()

        for i, col in enumerate(queue_cols1):
            data = df1[col].dropna()

            # Clip to 95th percentile to avoid huge x-axis spread
            max_clip = data.quantile(0.95)
            data_clipped = data[data <= max_clip]

            axs[i].hist(data_clipped, bins=30, edgecolor='black', color='skyblue')
            axs[i].set_title(f'Queue Time: {col}', fontsize=11)
            axs[i].set_xlabel('Time (minutes)')
            axs[i].set_ylabel('Frequency')
            axs[i].grid(True)

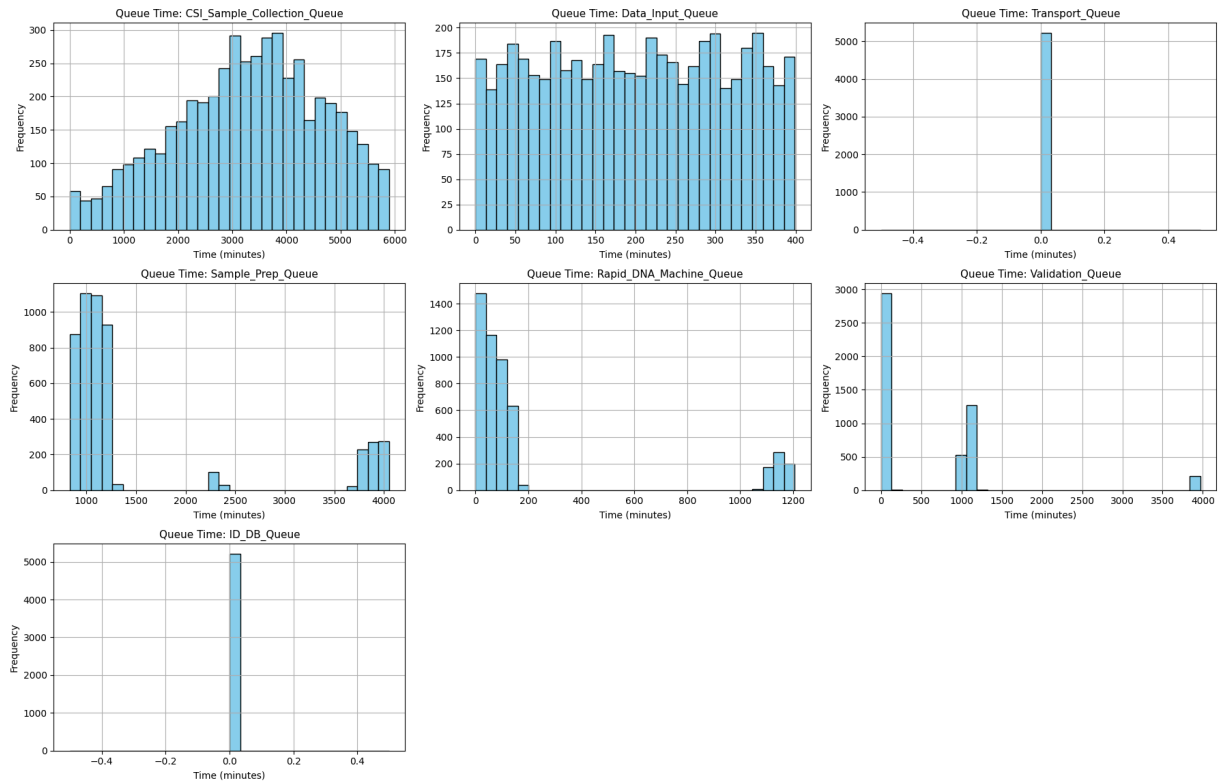
        # Hide unused subplots
        for j in range(len(queue_cols1), len(axs)):

```

```
fig.delaxes(axs[j])
```

```
plt.suptitle('Queueing Time Distributions for Scenerio 1 (Scaled)', fontsize=18, y
plt.tight_layout()
plt.show()
```

Queueing Time Distributions for Scenerio 1 (Scaled))



## Scenerio 2

```
In [15]: all_values = pd.concat([df2[col].dropna() for col in queue_cols2])
xlim = (0, all_values.max() * 1.1)
ylim = (0, max(df2[col].value_counts().max() for col in queue_cols2) * 1.1)

fig, axs = plt.subplots(3, 3, figsize=(18, 12)) # 3x3 for 9 stages
axs = axs.flatten()

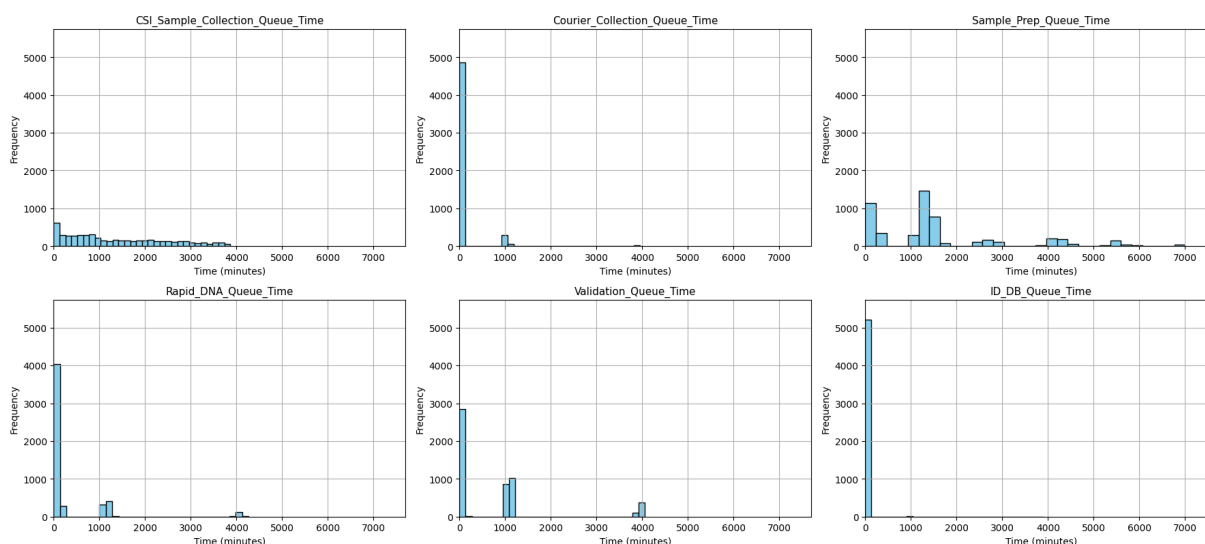
for i, col in enumerate(queue_cols2):
    axs[i].hist(df2[col].dropna(), bins=30, edgecolor='black', color='skyblue')
    axs[i].set_title(f'{col}', fontsize=11)
    axs[i].set_xlim(xlim)
    axs[i].set_ylim(ylim)
    axs[i].set_xlabel('Time (minutes)')
    axs[i].set_ylabel('Frequency')
    axs[i].grid(True)

# Hide any extra subplot boxes (if fewer than 9 columns)
for j in range(len(queue_cols2), len(axs)):
    fig.delaxes(axs[j])

plt.suptitle('Queueing Time Distributions for Scenerio 2', fontsize=18, y=1.02)
```

```
plt.tight_layout()
plt.show()
```

Queueing Time Distributions for Scenerio 2



```
In [16]: fig, axs = plt.subplots(3, 3, figsize=(18, 12))
         axs = axs.flatten()

         for i, col in enumerate(queue_cols2):
             data = df2[col].dropna()

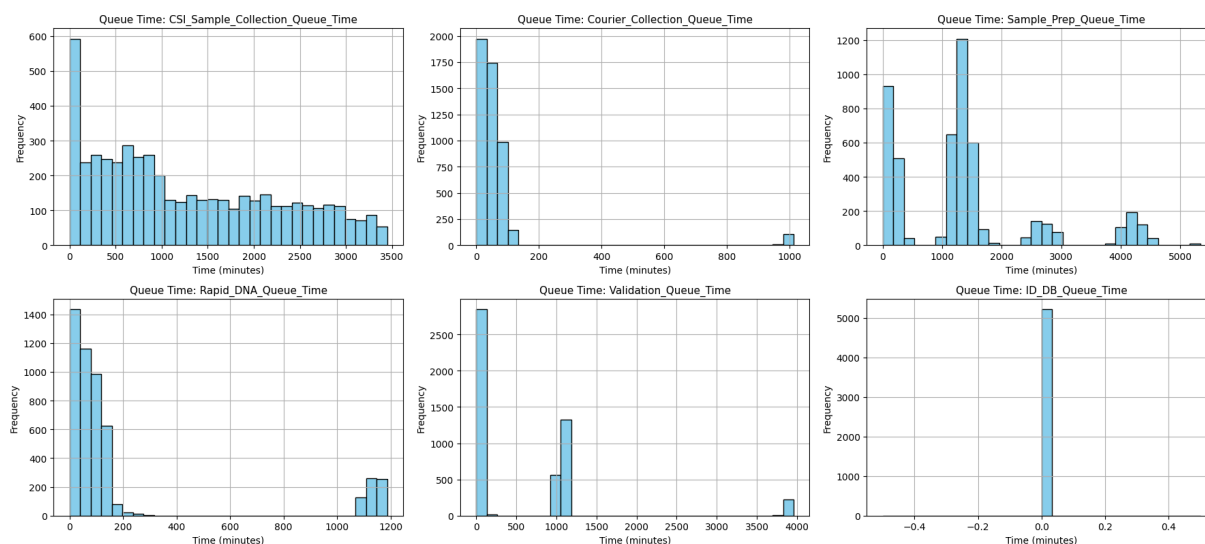
             # Clip to 95th percentile to avoid huge x-axis spread
             max_clip = data.quantile(0.95)
             data_clipped = data[data <= max_clip]

             axs[i].hist(data_clipped, bins=30, edgecolor='black', color='skyblue')
             axs[i].set_title(f'Queue Time: {col}', fontsize=11)
             axs[i].set_xlabel('Time (minutes)')
             axs[i].set_ylabel('Frequency')
             axs[i].grid(True)

         # Hide unused subplots
         for j in range(len(queue_cols2), len(axs)):
             fig.delaxes(axs[j])

         plt.suptitle('Queueing Time Distributions for Scenerio 2 (Scaled)', fontsize=18, y
         plt.tight_layout()
         plt.show()
```

Queueing Time Distributions for Scenerio 2 (Scaled)



## Scenerio 3

```
In [17]: all_values = pd.concat([df3[col].dropna() for col in queue_cols3])
xlim = (0, all_values.max() * 1.1)
ylim = (0, max(df3[col].value_counts().max() for col in queue_cols3) * 1.1)

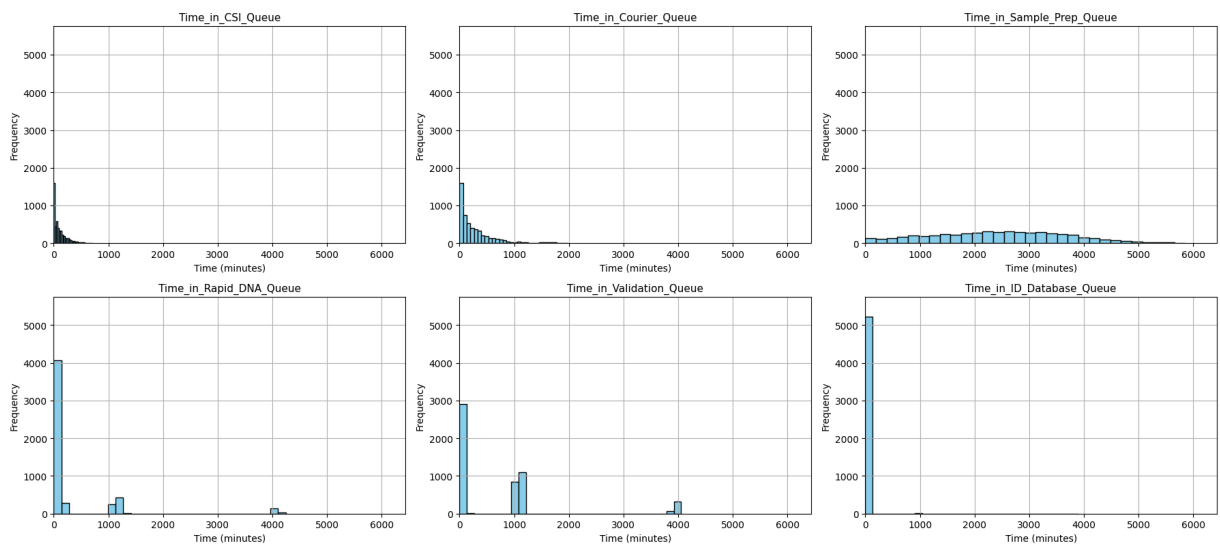
fig, axs = plt.subplots(3, 3, figsize=(18, 12)) # 3x3 for 9 stages
axs = axs.flatten()

for i, col in enumerate(queue_cols3):
    axs[i].hist(df3[col].dropna(), bins=30, edgecolor='black', color='skyblue')
    axs[i].set_title(f'{col}', fontsize=11)
    axs[i].set_xlim(xlim)
    axs[i].set_ylim(ylim)
    axs[i].set_xlabel('Time (minutes)')
    axs[i].set_ylabel('Frequency')
    axs[i].grid(True)

# Hide any extra subplot boxes (if fewer than 9 columns)
for j in range(len(queue_cols3), len(axs)):
    fig.delaxes(axs[j])

plt.suptitle('Queueing Time Distributions for Scenerio 3', fontsize=18, y=1.02)
plt.tight_layout()
plt.show()
```

## Queueing Time Distributions for Scenerio 3



```
In [18]: fig, axs = plt.subplots(3, 3, figsize=(18, 12))
axs = axs.flatten()

for i, col in enumerate(queue_cols3):
    data = df3[col].dropna()

    # Clip to 95th percentile to avoid huge x-axis spread
    max_clip = data.quantile(0.95)
    data_clipped = data[data <= max_clip]

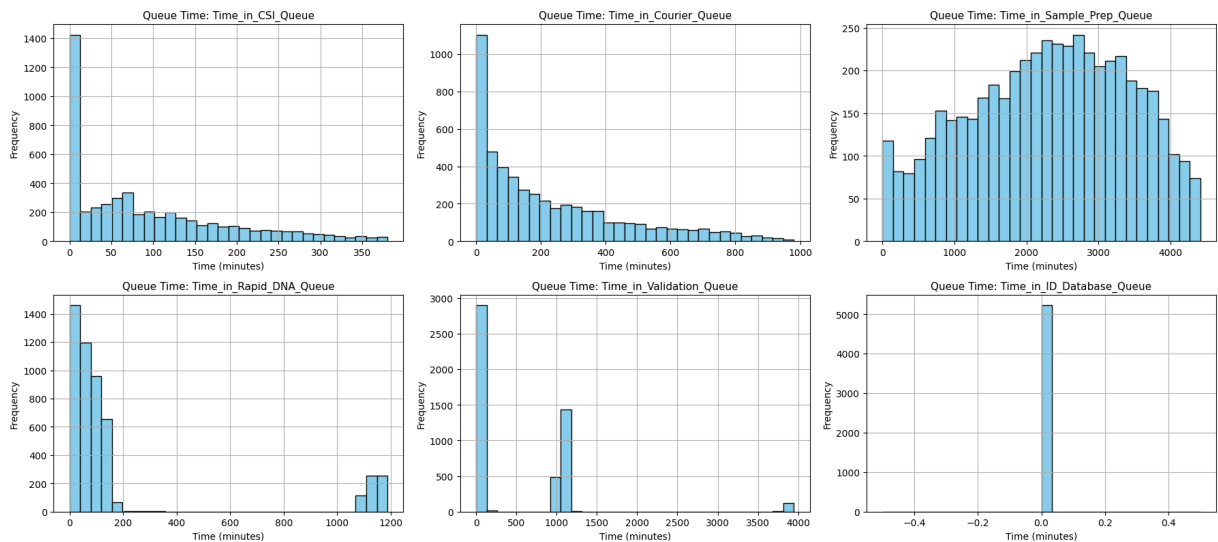
    axs[i].hist(data_clipped, bins=30, edgecolor='black', color='skyblue')
    axs[i].set_title(f'Queue Time: {col}', fontsize=11)
    axs[i].set_xlabel('Time (minutes)')
    axs[i].set_ylabel('Frequency')
    axs[i].grid(True)

# Hide unused subplots
for j in range(len(queue_cols3), len(axs)):
    fig.delaxes(axs[j])

plt.suptitle('Queueing Time Distributions for Scenerio 3 (Scaled)', fontsize=18, y
plt.tight_layout()
plt.show()
```



Queueing Time Distributions for Scenerio 3 (Scaled)



## Scenerio 4

```
In [19]: all_values = pd.concat([df4[col].dropna() for col in queue_cols4])
xlim = (0, all_values.max() * 1.1)
ylim = (0, max(df4[col].value_counts().max() for col in queue_cols4) * 1.1)

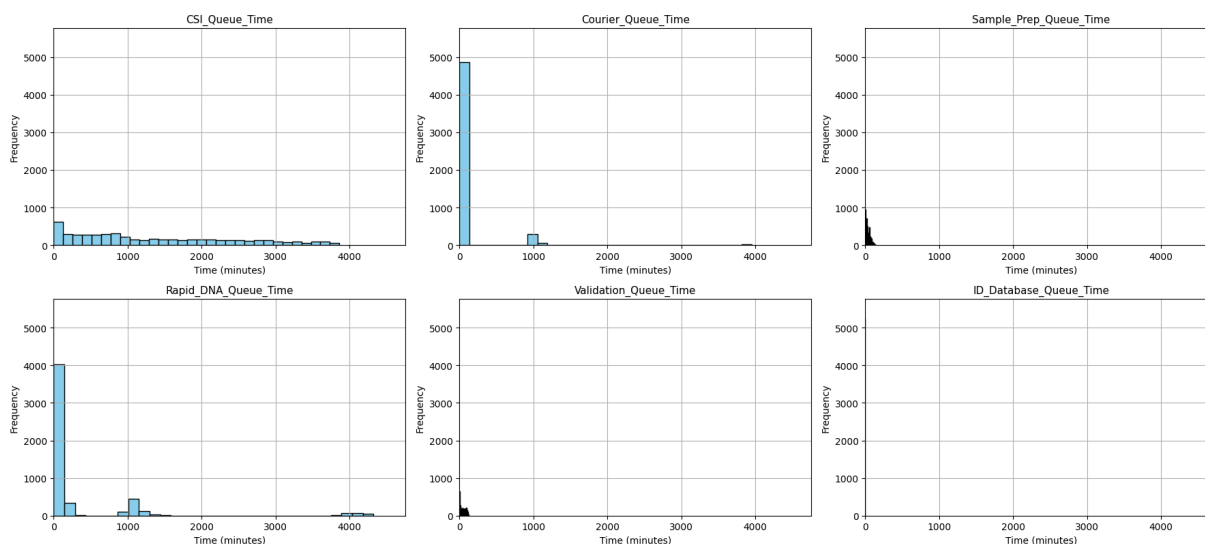
fig, axs = plt.subplots(3, 3, figsize=(18, 12)) # 3x3 for 9 stages
axs = axs.flatten()

for i, col in enumerate(queue_cols4):
    axs[i].hist(df4[col].dropna(), bins=30, edgecolor='black', color='skyblue')
    axs[i].set_title(f'{col}', fontsize=11)
    axs[i].set_xlim(xlim)
    axs[i].set_ylim(ylim)
    axs[i].set_xlabel('Time (minutes)')
    axs[i].set_ylabel('Frequency')
    axs[i].grid(True)

# Hide any extra subplot boxes (if fewer than 9 columns)
for j in range(len(queue_cols4), len(axs)):
    fig.delaxes(axs[j])

plt.suptitle('Queueing Time Distributions for Scenerio 4', fontsize=18, y=1.02)
plt.tight_layout()
plt.show()
```

## Queueing Time Distributions for Scenerio 4



```
In [20]: fig, axs = plt.subplots(3, 3, figsize=(18, 12))
axs = axs.flatten()

for i, col in enumerate(queue_cols4):
    data = df4[col].dropna()

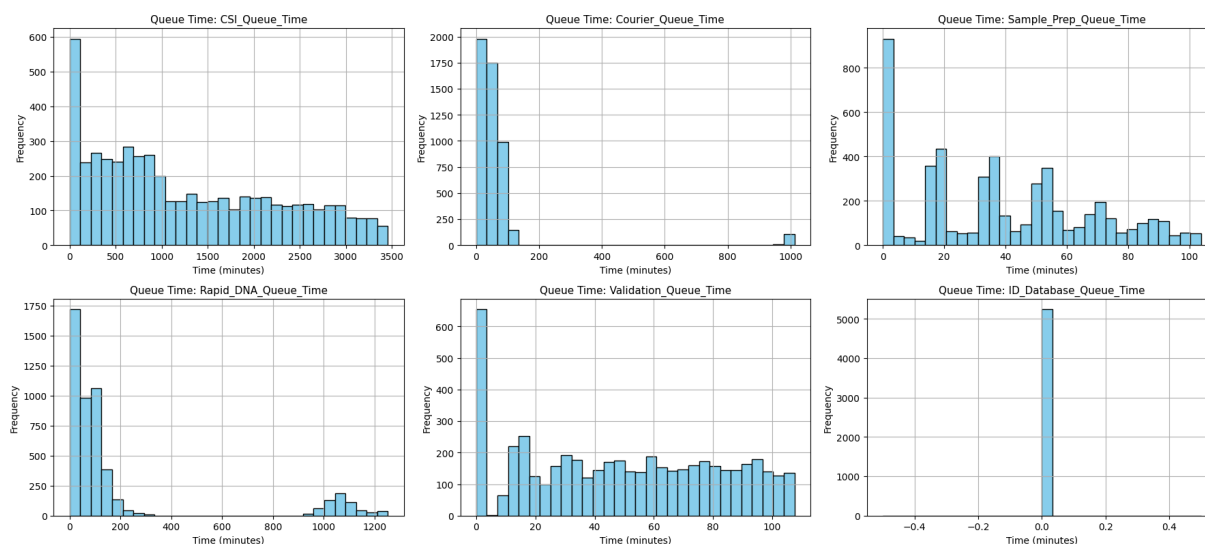
    # Clip to 95th percentile to avoid huge x-axis spread
    max_clip = data.quantile(0.95)
    data_clipped = data[data <= max_clip]

    axs[i].hist(data_clipped, bins=30, edgecolor='black', color='skyblue')
    axs[i].set_title(f'Queue Time: {col}', fontsize=11)
    axs[i].set_xlabel('Time (minutes)')
    axs[i].set_ylabel('Frequency')
    axs[i].grid(True)

# Hide unused subplots
for j in range(len(queue_cols4), len(axs)):
    fig.delaxes(axs[j])

plt.suptitle('Queueing Time Distributions for Scenerio 4 (Scaled)', fontsize=18, y
plt.tight_layout()
plt.show()
```

Queueing Time Distributions for Scenerio 4 (Scaled)



## Scenerio 5

```
In [21]: all_values = pd.concat([df5[col].dropna() for col in queue_cols5])
xlim = (0, all_values.max() * 1.1)
ylim = (0, max(df5[col].value_counts().max() for col in queue_cols5) * 1.1)

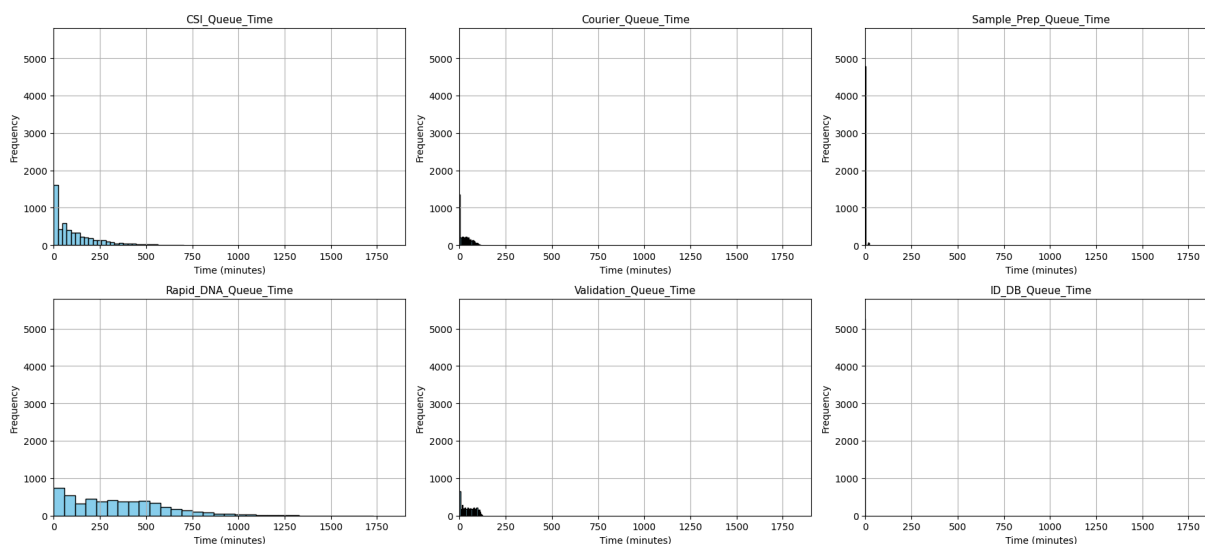
fig, axs = plt.subplots(3, 3, figsize=(18, 12)) # 3x3 for 9 stages
axs = axs.flatten()

for i, col in enumerate(queue_cols5):
    axs[i].hist(df5[col].dropna(), bins=30, edgecolor='black', color='skyblue')
    axs[i].set_title(f'{col}', fontsize=11)
    axs[i].set_xlim(xlim)
    axs[i].set_ylim(ylim)
    axs[i].set_xlabel('Time (minutes)')
    axs[i].set_ylabel('Frequency')
    axs[i].grid(True)

# Hide any extra subplot boxes (if fewer than 9 columns)
for j in range(len(queue_cols5), len(axs)):
    fig.delaxes(axs[j])

plt.suptitle('Queueing Time Distributions for Scenerio 5', fontsize=18, y=1.02)
plt.tight_layout()
plt.show()
```

## Queueing Time Distributions for Scenerio 5



```
In [22]: fig, axs = plt.subplots(3, 3, figsize=(18, 12))
axs = axs.flatten()

for i, col in enumerate(queue_cols5):
    data = df5[col].dropna()

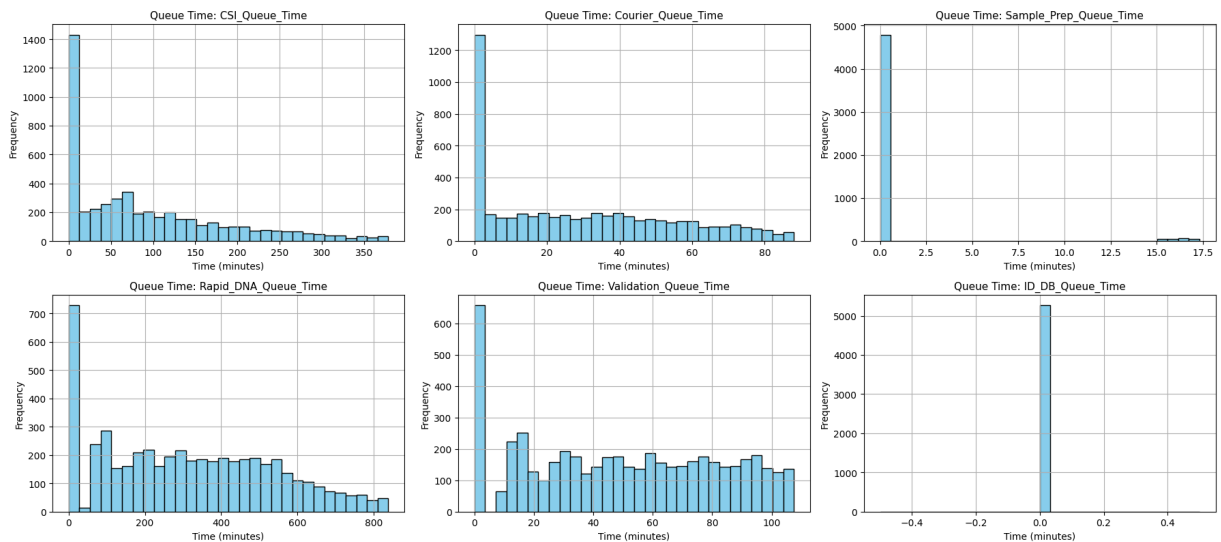
    # Clip to 95th percentile to avoid huge x-axis spread
    max_clip = data.quantile(0.95)
    data_clipped = data[data <= max_clip]

    axs[i].hist(data_clipped, bins=30, edgecolor='black', color='skyblue')
    axs[i].set_title(f'Queue Time: {col}', fontsize=11)
    axs[i].set_xlabel('Time (minutes)')
    axs[i].set_ylabel('Frequency')
    axs[i].grid(True)

# Hide unused subplots
for j in range(len(queue_cols5), len(axs)):
    fig.delaxes(axs[j])

plt.suptitle('Queueing Time Distributions for Scenerio 5 (Scaled)', fontsize=18, y=
plt.tight_layout()
plt.show()
```

Queueing Time Distributions for Scenerio 5 (Scaled)



////////////////////////////////////

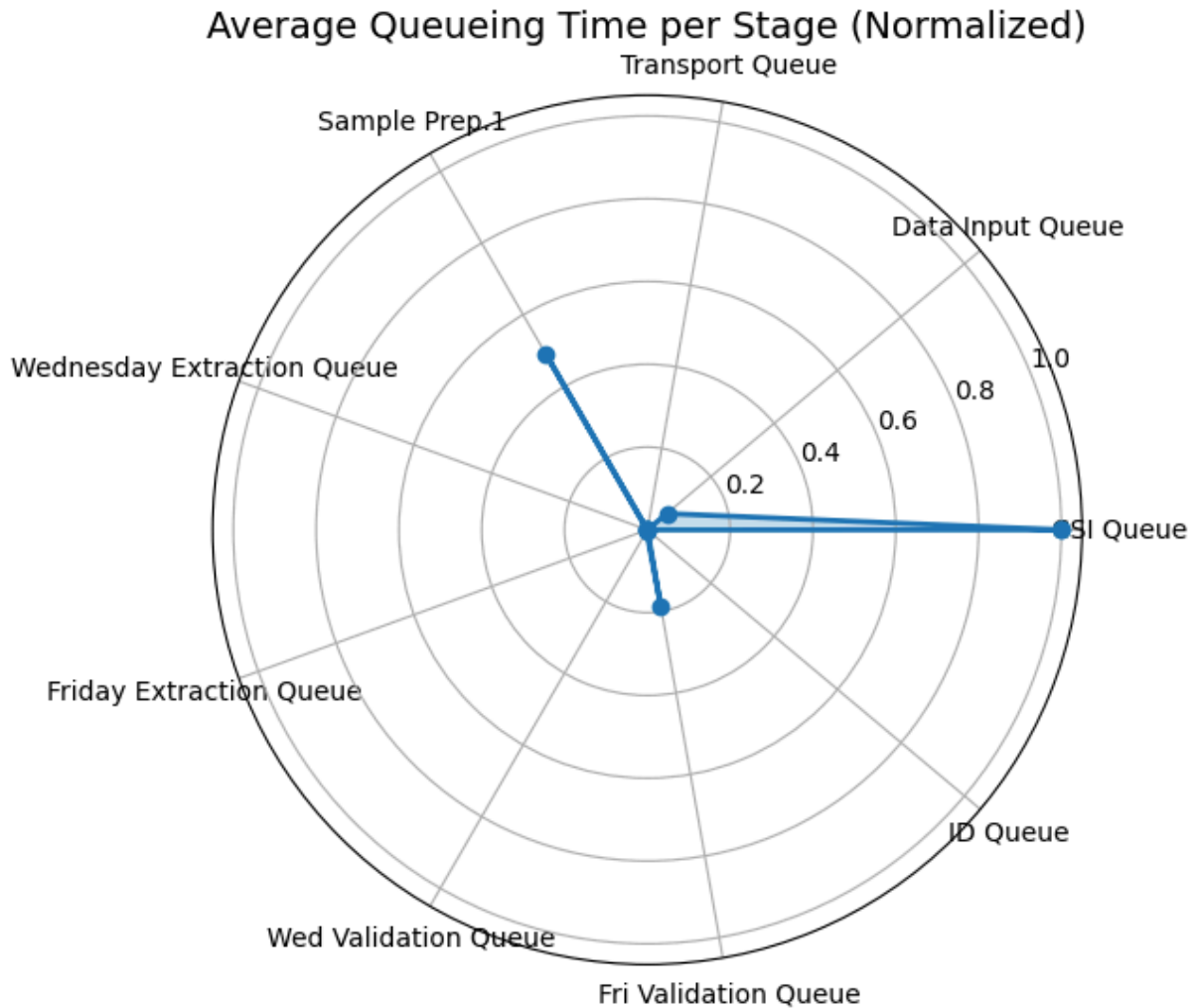
## Queue times spidergram for current scenerio

```
In [25]: queue_cols = ['CSI Queue', 'Data Input Queue', 'Transport Queue', 'Sample Prep.1',
avg_queues = df0[queue_cols0].mean()
normalized = avg_queues / avg_queues.max()

labels = normalized.index.tolist()
values = normalized.values.tolist()

# Close the Loop
values += values[:1]
angles = np.linspace(0, 2 * np.pi, len(labels), endpoint=False).tolist()
angles += angles[:1]

# Plot
fig, ax = plt.subplots(figsize=(6, 6), subplot_kw=dict(polar=True))
ax.plot(angles, values, 'o-', linewidth=2)
ax.fill(angles, values, alpha=0.25)
ax.set_thetagrids(np.degrees(angles[:-1]), labels)
ax.set_title('Average Queueing Time per Stage (Normalized)', size=14)
plt.show()
```



## Comparative Spidergram for Queue Times

```
In [27]: labels = ['CSI', 'Transport', 'Prep', 'Sequencing', 'Validation', 'DB']

# Ensure all scenario values have EXACTLY 6 values
spider_data = {
    'Current Sc': [
        avg_queues0.loc['CSI Queue'].item(),
        avg_queues0.loc['Transport Queue'].item(),
        avg_queues0.loc['Sample Prep.1'].item(),
        avg_queues0.loc[['Wednesday Extraction Queue', 'Friday Extraction Queue']].mean().item(),
        avg_queues0.loc[['Wed Validation Queue', 'Fri Validation Queue']].mean().item(),
        avg_queues0.loc['ID Queue'].item()
    ],
    'Scenario 1': [
        avg_queues1.loc['CSI_Sample_Collection_Queue'].item(),
        avg_queues1.loc[['Data_Input_Queue', 'Transport_Queue']].mean().item(),
        avg_queues1.loc['Sample_Prep_Queue'].item(),
        avg_queues1.loc['Rapid_DNA_Machine_Queue'].item(),
        avg_queues1.loc[['Validation_Queue']].mean().item(),
        avg_queues1.loc['ID_DB_Queue'].item()
    ],
    'Scenario 2': avg_queues2.values.tolist()[6],
}
```

```

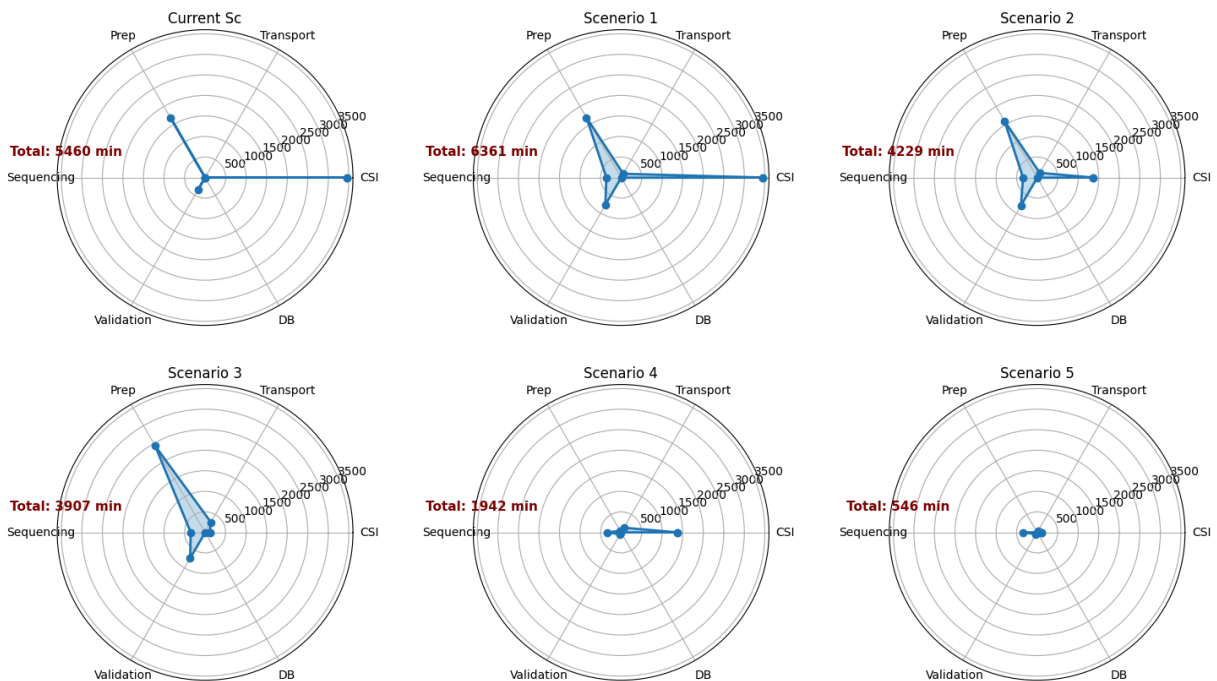
'Scenario 3': avg_queues3.values.tolist()[:6],
'Scenario 4': avg_queues4.values.tolist()[:6],
'Scenario 5': avg_queues5.values.tolist()[:6]
}

# Build consistent angles
angles = np.linspace(0, 2 * np.pi, len(labels), endpoint=False).tolist()
angles += angles[:1]

# Plotting
fig, axs = plt.subplots(2, 3, subplot_kw=dict(polar=True), figsize=(18, 10))
angles = np.linspace(0, 2 * np.pi, len(labels), endpoint=False).tolist()
angles += angles[:1]

for ax, (scenario, values) in zip(axs.flat, spider_data.items()):
    values_plot = values + values[:1]
    ax.plot(angles, values_plot, 'o-', linewidth=2)
    ax.fill(angles, values_plot, alpha=0.25)
    ax.set_thetagrids(np.degrees(angles[:-1]), labels)
    ax.set_title(scenario)
    ax.set_ylim(0, 3600)
    total_time = sum(values)
    ax.text(3, ax.get_ylim()[1] * 0.95, f"Total: {int(total_time)} min",
           ha='center', va='bottom', fontsize=11, fontweight='bold', color='maroon')

```



```

In [16]: data = {
    'Scenario': ['Current Scenario', 'Scenario 1', 'Scenario 2', 'Scenario 3', 'Scenario 4', 'Scenario 5'],
    'Avg_Time_in_System': [10971, 7211, 5027, 4372, 2584, 871],
    'Cost_per_Arrest_m': [2.5, 3.12, 2.38, 2.33, 1.43, 1.1]
}

df = pd.DataFrame(data)

# Create figure and axis
fig, ax1 = plt.subplots(figsize=(10, 6))

```

```

# Bar plot for average time
bar = ax1.bar(df['Scenario'], df['Avg_Time_in_System'], color='royalblue', label='A')
ax1.set_xlabel('Scenario')
ax1.set_ylabel('Avg Time in System', color='royalblue')
ax1.tick_params(axis='y', labelcolor='royalblue')

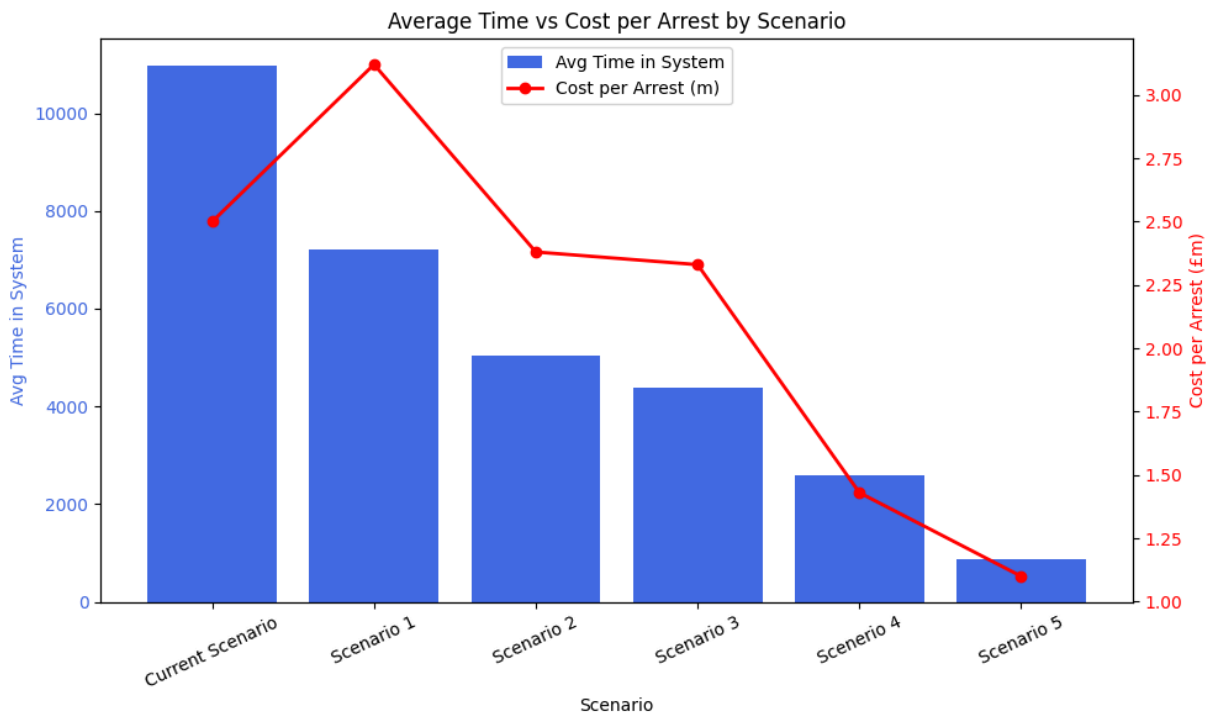
# Rotate x Labels if needed
plt.xticks(rotation=25)

# Secondary axis for cost per arrest
ax2 = ax1.twinx()
line = ax2.plot(df['Scenario'], df['Cost_per_Arrest_m'], color='red', marker='o', 1
ax2.set_ylabel('Cost per Arrest (£m)', color='red')
ax2.tick_params(axis='y', labelcolor='red')

# Combine Legends
lines, labels = ax1.get_legend_handles_labels()
lines2, labels2 = ax2.get_legend_handles_labels()
ax1.legend(lines + lines2, labels + labels2, loc='upper center')

plt.title('Average Time vs Cost per Arrest by Scenario')
plt.tight_layout()
plt.show()

```



In [ ]: