```
In [1]: import pandas as pd
        import numpy as np
        from lifelines import CoxPHFitter
        import matplotlib.pyplot as plt
        # Load the data from the .xlsx file
        data = pd.read_excel('data1.xlsx')
        # Perform univariate analysis
        univariate_results = []
        univariate_aic_bic = []
        for covariate in data.columns:
            if covariate not in ['Months', 'DEATH']:
                cph_univariate = CoxPHFitter()
                cph_univariate.fit(data[[covariate, 'Months', 'DEATH']], duration_col='M
                # Calculate log-likelihood
                llf = cph_univariate.log_likelihood_
                # Calculate number of parameters
                k = cph_univariate.params_.shape[0]
                # Calculate AIC and BIC
                n = len(data)
                aic = -2 * 11f + 2 * k
                bic = -2 * 11f + k * np.log(n)
                univariate_results.append((covariate, cph_univariate.summary['p'][covari
                univariate_aic_bic.append((covariate, aic, bic))
        # Sort the univariate models based on AIC
        univariate aic bic.sort(key=lambda x: x[1])
        # Select significant variables based on p-values
        significant_variables = [(var, p_value) for var, p_value in univariate_results i
        # Fit the multivariate model
        multivariate_data = data[[var for var, _ in significant_variables] + ['Months',
        cph_multivariate = CoxPHFitter()
        cph_multivariate.fit(multivariate_data, duration_col='Months', event_col='DEATH'
        # Calculate log-likelihood for multivariate model
        llf_multivariate = cph_multivariate.log_likelihood_
        # Calculate number of parameters for multivariate model
        k_multivariate = cph_multivariate.params_.shape[0]
        # Calculate AIC and BIC for multivariate model
        n = len(data)
        aic_multivariate = -2 * llf_multivariate + 2 * k_multivariate
        bic_multivariate = -2 * llf_multivariate + k_multivariate * np.log(n)
        # Print AIC and BIC for univariate models
        print("AIC and BIC for univariate models:")
        for covariate, aic, bic in univariate_aic_bic:
            print(f"{covariate}: AIC={aic}, BIC={bic}")
        # Print significant variables from univariate analysis
```

```
print("\nSignificant variables from univariate analysis:")
 for var, p_value in significant_variables:
     print(f"{var}: p-value={p_value}")
 # Print AIC and BIC for multivariate model
 print("\nAIC and BIC for multivariate model:")
 print("AIC of the multivariate model:", aic_multivariate)
 print("BIC of the multivariate model:", bic_multivariate)
AIC and BIC for univariate models:
CompositeStage: AIC=1801.0425436924681, BIC=1804.8802741396341
ID: AIC=1854.1291924816765, BIC=1857.9669229288425
LNInvolment: AIC=1877.7266674202021, BIC=1881.5643978673681
AGE: AIC=1887.3267837410287, BIC=1891.1645141881947
Comorbidity: AIC=1888.099387799303, BIC=1891.937118246469
FamiliyHistoryOfCancer: AIC=1889.2113357577587, BIC=1893.0490662049247
SEX: AIC=1889.2852518349302, BIC=1893.1229822820962
Significant variables from univariate analysis:
ID: p-value=5.656542610612734e-09
CompositeStage: p-value=2.724672855669092e-18
```

AIC and BIC for multivariate model:

LNInvolment: p-value=0.0011578069966356516

AIC of the multivariate model: 1768.6487964813414 BIC of the multivariate model: 1780.1619878228391

In []: