



Model Development Phase Template

Date	23 June 2025
Team ID	178047
Project Title	Unemployed Insurance Beneficiary Forecasting
Maximum Marks	10 Marks

Initial Model Training Code, Model Validation and Evaluation Report

The initial model training code will be showcased in the future through a screenshot. The model validation and evaluation report will include a summary and training and validation performance metrics for multiple models, presented through respective screenshots.

Initial Model Training Code (5 marks):

```
Training and testing

df.dropna(inplace=True)

train_size=int(len(df)*0.8)
train,test=df[:train_size],df[train_size:]
```





Model building

```
[40] #Augmented Dickey-Fuller test(used to convert non-stationary data to
     #stationary data)
     adf=adfuller(df['Beneficiaries'],autolag='AIC')
     print("P-Value",adf[1])
     adf=adfuller(train['Beneficiaries diff'],autolag='AIC')
     print("P-Value",adf[1])
     #ACF and PACF(to check how our data is correlated with ACF &PACF)
     plot_acf(train['Beneficiaries'], lags=30, title='Original ACF')
     plot pacf(train['Beneficiaries'], lags=30, title='Original PACF')
     plt.show()
     #differenced ACF and PCAF
     plot_acf(train['Beneficiaries_diff'], lags=30, title='Differenced ACF')
     plot_pacf(train['Beneficiaries_diff'], lags=30, title='Differenced PACF')
     plt.show()
     #Augmented Dickey-Fuller test(used to convert non-stationary data to
     #stationary data)
     adf=adfuller(df['Beneficiaries'],autolag='AIC')
     print("P-Value",adf[1])
     adf=adfuller(train['Beneficiaries_diff'],autolag='AIC')
     print("P-Value",adf[1])
     #ACF and PACF(to check how our data is correlated with ACF &PACF)
     plot_acf(train['Beneficiaries'], lags=30, title='Original ACF')
     plot_pacf(train['Beneficiaries'], lags=30, title='Original PACF')
     plt.show()
     #differenced ACF and PCAF
     plot_acf(train['Beneficiaries_diff'].dropna(), lags=30, title='Differenced ACF')
     plot pacf(train['Beneficiaries diff'].dropna(), lags=30, title='Differenced PACF')
     plt.show()
     #smoothing out our data(visual representation)
     plt.plot(train['Beneficiaries'])
     plt.plot(train['Beneficiaries_diff'])
     plt.show()
```





```
from statsmodels.tsa.arima.model import ARIMA
model = ARIMA(train['Beneficiaries_diff'].dropna(), order=(5,0,0)) # Pass the series with the correct index
model_arima=model.fit()
model_arima.summary()
#SARIMA
model=SARIMAX(train['Beneficiaries_diff'].dropna(),order=(5,0,0),seasonal_order=(0,1,2,3))
model sarima=model.fit()
model sarima.summary()
#Auto Regression
model ar=AutoReg(train['Beneficiaries diff'].dropna(), lags=10).fit()
model ar.summary()
#VAR
model =VAR(train[['Beneficiaries_diff','Benefit Amounts (Dollars)']].dropna()) # Add dropna()
model AR = model.fit(maxlags=10)
model_AR.summary()
import prophet
# Prepare the data for Prophet
# Prophet requires the dataframe to have columns named 'ds' and 'y'
prophet_df = df[['Year', 'Month', 'Beneficiaries']].copy()
prophet_df['ds'] = pd.to_datetime(prophet_df['Year'].astype(str) + '-' + prophet_df['Month'].astype(str))
prophet_df = prophet_df[['ds', 'Beneficiaries']].rename(columns={'Beneficiaries': 'y'})
# Instantiate and fit the Prophet model
model_prophet = prophet.Prophet()
model_prophet.fit(prophet_df)
```





Model Validation and Evaluation Report (5 marks):

Model	Summary	Training and Validation Performance Metrics (MSE,MAE,R2 score)
ARIMA	SARIPMX Results Dep. Variable: Beneficiaries diff No. Observations: 11007 No. Observations: 11007 No. Observations: 11007 No. Observations: 12007	102763733.35841592 5691.3713943589655 -8.184450493908813e-05
SARIMA	Dep. Variable: December Dec	103407466.6558495 5833.0826353155 R2-score: NA
AutoReg	AutoReg Model Results Dep. Variable: Beneficiaries.diff No. Observations: 11007	102771796.73461813 5692.993043170196 R2-score: NA





VAR	Summary of Regression Results	NA
Prophet	ds yhat yhat_lower yhat_upper 0 2001-01-01 4850.952377 -2818.407480 13519.833896 1 2001-02-01 4636.011708 -3680.946446 12425.524121 2 2001-03-01 4531.799997 -3757.700050 12761.117768 3 2001-04-01 4423.377526 -4139.662938 13044.800193 4 2001-05-01 3827.318423 -4071.144641 12201.204735	49080877.302183054 4541.665548954107 0.003339211190899194