Forecasts Prediksi Global Sales Video Games

UAS STATISTIKA Ilmu Komputer 3B

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Source Dan Dependency

- Train data dan Predict data menggunakan library dari Facebook Prophet (https://facebook.github.io/prophet/)
- Data yang digunakan Video Game Sales
 (https://www.kaggle.com/datasets/thedevastator/video-game-sales-and-ratings?
 resource=download)

Install Dan Import Dependency

```
In [ ]: %pip install prophet pystan

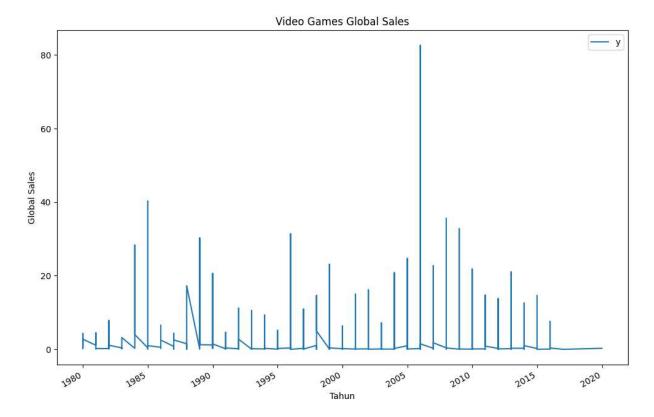
In [ ]: import pandas as pd # import pandas
    from prophet import Prophet # import prophet
    import matplotlib.pyplot as plt # import matplotlib
```

Membaca dan Memproses Tahun Dari Data

```
In [ ]: df = pd.read_csv('Video_Games.csv') # membaca data
    df.head() # melihat 5 data teratas
```

Out[]:		index	Name	Platform	Year_of_Release	Genre	Publisher	NA_Sales	EU_Sales
	0	0	Wii Sports	Wii	2006.0	Sports	Nintendo	41.36	28.96
	1	1	Super Mario Bros.	NES	1985.0	Platform	Nintendo	29.08	3.58
	2	2	Mario Kart Wii	Wii	2008.0	Racing	Nintendo	15.68	12.76
	3	3	Wii Sports Resort	Wii	2009.0	Sports	Nintendo	15.61	10.93
	4	4	Pokemon Red/Pokemon Blue	GB	1996.0	Role- Playing	Nintendo	11.27	8.89
	4								•

```
df.describe() # melihat deskripsi data
Out[]:
                      index Year_of_Release
                                                 NA_Sales
                                                               EU_Sales
                                                                             JP_Sales
                                                                                        Other_Sa
         count 16928.000000
                                 16655.00000
                                             16928.000000 16928.000000 16928.000000
                                                                                      16928.000C
                                                  0.266390
                                                                             0.078170
                                                                                           0.0478
         mean
                 8463.500000
                                  2006.48532
                                                               0.146850
           std
                 4886.837014
                                     5.88289
                                                  0.814326
                                                               0.504585
                                                                             0.308756
                                                                                           0.1866
          min
                    0.000000
                                  1980.00000
                                                  0.000000
                                                               0.000000
                                                                             0.000000
                                                                                           0.0000
          25%
                 4231.750000
                                  2003.00000
                                                  0.000000
                                                               0.000000
                                                                             0.000000
                                                                                           0.000C
          50%
                 8463.500000
                                  2007.00000
                                                  0.080000
                                                               0.020000
                                                                             0.000000
                                                                                           0.0100
          75%
                12695.250000
                                  2010.00000
                                                  0.240000
                                                               0.110000
                                                                             0.040000
                                                                                           0.0400
          max 16927.000000
                                  2020.00000
                                                 41.360000
                                                              28.960000
                                                                            10.220000
                                                                                          10.5700
In [ ]: # Semua Kolom Data
         df.columns
Out[ ]: Index(['index', 'Name', 'Platform', 'Year_of_Release', 'Genre', 'Publisher',
                 'NA_Sales', 'EU_Sales', 'JP_Sales', 'Other_Sales', 'Global_Sales',
                'Critic_Score', 'Critic_Count', 'User_Score', 'User_Count', 'Developer',
                 'Rating'],
               dtype='object')
In [ ]: # Menghapus Kolom yang Tidak Diperlukan
         df = df[['Year of Release', 'Global Sales']]
         df.columns = ['ds', 'y']
         df['ds'] = pd.to_datetime(df['ds'], format='%Y')
        df.head()
Out[ ]:
                    ds
                           У
         0 2006-01-01 82.53
         1 1985-01-01 40.24
         2 2008-01-01 35.52
         3 2009-01-01 32.77
         4 1996-01-01 31.37
In [ ]: # Menampilkan Grafik Data
         df.set_index('ds').plot(figsize=(12, 8))
         plt.xlabel('Tahun')
         plt.ylabel('Global Sales')
         plt.title('Video Games Global Sales')
Out[ ]: Text(0.5, 1.0, 'Video Games Global Sales')
```

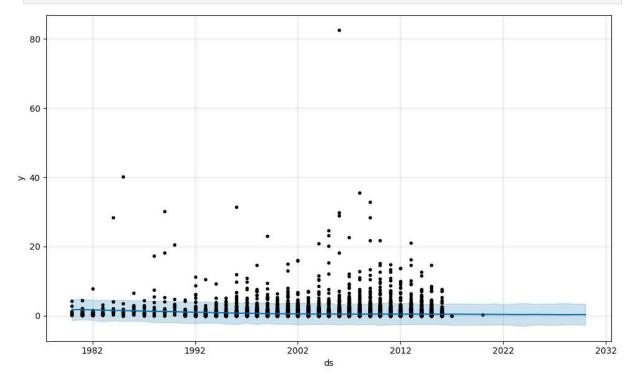


Mempelajari model data

```
In [ ]: # Train Model Prophet
        df = df.rename(columns={'Year_of_Release': 'ds', 'Global_Sales': 'y'})
        df.head()
Out[]:
                   ds
                          У
        0 2006-01-01 82.53
           1985-01-01 40.24
        2 2008-01-01 35.52
           2009-01-01 32.77
           1996-01-01 31.37
In [ ]: # Forcasting Data Pembelian Global Menggunakan Prophet
        model = Prophet(interval_width=0.95, daily_seasonality=True)
        model.fit(df)
        future = model.make_future_dataframe(periods=10,freq='Y') # membuat data prediksi 1
        forecast = model.predict(future)
        forecast[['ds', 'yhat', 'yhat_lower', 'yhat_upper']].tail()
```

```
Out[]:
                    ds
                            yhat yhat_lower yhat_upper
         44 2025-12-31 0.357469
                                   -2.586496
                                                3.452540
         45 2026-12-31 0.338431
                                   -2.723028
                                                3.496812
         46 2027-12-31 0.319282
                                   -2.526604
                                                3.631528
            2028-12-31 0.351579
                                   -2.563086
                                                3.453843
         48 2029-12-31 0.332669
                                   -2.665189
                                                3.445404
```

```
In []: # Visualisasi Data Global Sales
   if model is not None:
        model.plot(forecast, uncertainty=True)
        plt.show()
```



```
In [ ]: # Visualisasi Data Global Sales
   if model is not None:
        model.plot_components(forecast)
        plt.show()
```

