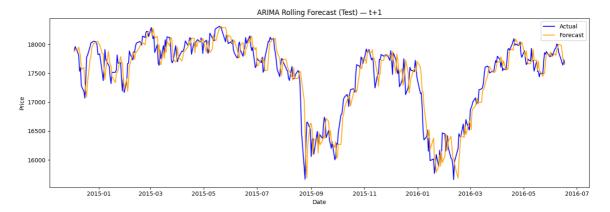
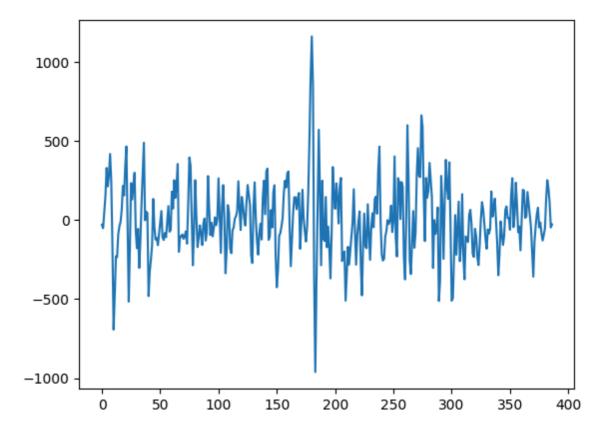
```
In [12]:
         import numpy as np
         import pandas as pd
         import joblib
         from tensorflow.keras.models import load_model
         from sklearn.metrics import mean_squared_error, mean_absolute_error, r2_sco
         import matplotlib.pyplot as plt
         from tcn import TCN
         from statsmodels.tsa.arima.model import ARIMA
         import os
In [2]: df_test = pd.read_csv("test_dataset.csv", parse_dates=["Date"])
         df_test = df_test.sort_values("Date").reset_index(drop=True)
In [3]: | df_train = pd.read_csv("train_dataset.csv", parse_dates=["Date"])
In [4]: | df_train.drop(columns=['Next_Close', 'Next_3_Close', 'Next_7_Close'], inpla
         ce=True)
         df_test.drop(columns=['Next_Close', 'Next_3_Close', 'Next_7_Close'], inplac
         e=True)
In [5]: df_test = df_test.iloc[:-10]
 In [6]: | df_test.shape
Out[6]: (388, 53)
```

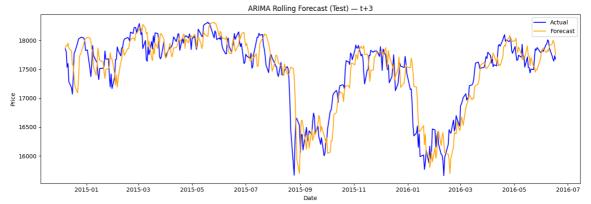
```
In [35]:
         def arima_true_rolling_test(df_train, df_test, forecast_horizon=1, order=
         (1, 1, 1)):
             print(f"\n 	≠ Rolling Forecast ARIMA TEST (t+{forecast_horizon}) with o
         rder {order}")
             close_series = df_train['Close'].tolist() + df_test['Close'].tolist()
             start_idx = len(df_train)
             history = close_series[:start_idx] # Only train data initially
             y true = []
             y_pred = []
             for t in range(len(df_test) - forecast_horizon):
                     model = ARIMA(history, order=order).fit()
                     forecast = model.forecast(steps=forecast horizon)
                     y_pred.append(forecast[-1])
                     y_true.append(close_series[start_idx + t + forecast_horizon])
                 except:
                     y_pred.append(np.nan)
                     y_true.append(np.nan)
                 history.append(close_series[start_idx + t]) # simulate real-time u
         pdate
             y_true = np.array(y_true)
             y_pred = np.array(y_pred)
             mask = ~np.isnan(y_pred)
             y_true = y_true[mask]
             y_pred = y_pred[mask]
             date_series = df_test['Date'].iloc[forecast_horizon: len(y_true) + fore
         cast_horizon].values
             r2 = r2_score(y_true, y_pred)
             rmse = np.sqrt(mean_squared_error(y_true, y_pred))
             mae = mean_absolute_error(y_true, y_pred)
             print(f"  ARIMA Test (t+{forecast_horizon}): R² = {r2:.4f}, RMSE = {r
         mse:.2f}, MAE = {mae:.2f}")
             plt.figure(figsize=(14, 5))
             plt.plot(date series, y true, label='Actual', color='blue')
             plt.plot(date_series, y_pred, label='Forecast', color='orange')
             plt.title(f"ARIMA Rolling Forecast (Test) - t+{forecast_horizon}")
             plt.xlabel("Date")
             plt.ylabel("Price")
             plt.legend()
             plt.tight_layout()
             plt.show()
             plt.plot(np.array(y pred) - np.array(y true), label='Prediction Error
         (t+h - forecast)')
             return {"horizon": forecast_horizon, "R2": r2, "RMSE": rmse, "MAE": ma
         e}
         test_results = []
         for horizon in [1, 3, 7]:
             res = arima_true_rolling_test(df_train, df_test, forecast_horizon=horiz
```

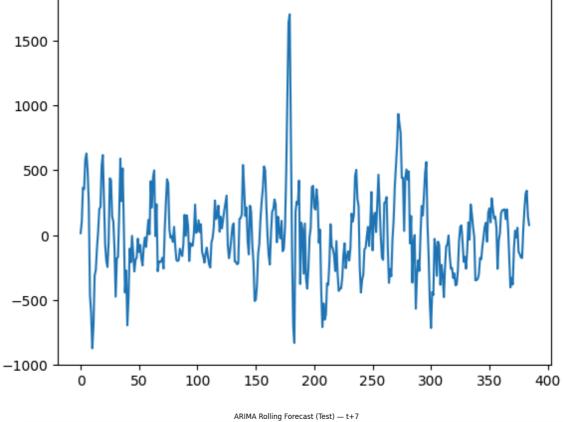
on, order=(1, 1, 1))
 test\_results.append(res)

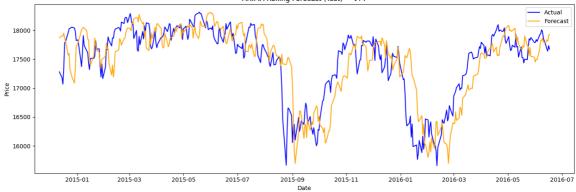
Rolling Forecast ARIMA TEST (t+1) with order (1, 1, 1)
ARIMA Test (t+1): R² = 0.8655, RMSE = 228.79, MAE = 170.45

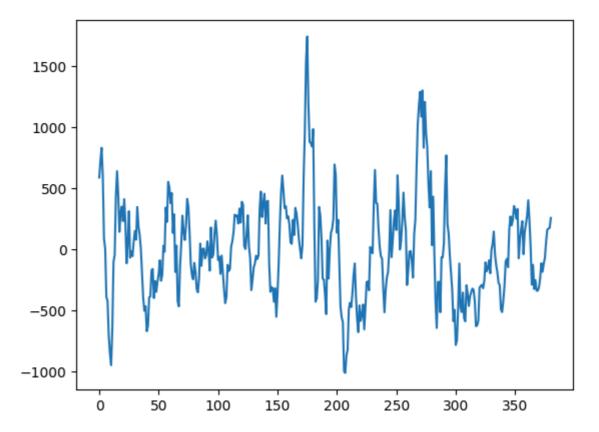












In [8]: df\_test.info()

<class 'pandas.core.frame.DataFrame'> RangeIndex: 388 entries, 0 to 387 Data columns (total 53 columns):

#	Column	Non-	-Null Count	Dtype
0	Date	388	non-null	datetime64[ns]
1	Open		non-null	float64
2	High		non-null	float64
3	Low	388	non-null	float64
4	Close	388	non-null	float64
5	Volume	388	non-null	int64
6	Adj Close	388	non-null	float64
7	Log_Returns	388	non-null	float64
8	Volatility_Log_10	388	non-null	float64
9	cl-op	388	non-null	float64
10	hi-lo	388	non-null	float64
11	Label	388	non-null	int64
12	<pre>vader_news_sentiment</pre>	388	non-null	float64
13	FinBERT_news_sentiment	388	non-null	float64
14	Smart_news_sentiment	388	non-null	float64
15	news_buying_intent	388	non-null	float64
16	news_selling_intent	388	non-null	float64
17	news_uncertainty_intent	388	non-null	float64
18	news_urgency_intent	388	non-null	float64
19	news_prediction_intent	388	non-null	float64
20	news_fear_intent	388	non-null	float64
21	news_greed_intent	388	non-null	float64
22	news_question_intent	388	non-null	float64
23	news_action_intent		non-null	float64
24	<pre>vader_reddit_sentiment</pre>		non-null	float64
25	FinBERT_reddit_sentiment		non-null	float64
26	Smart_reddit_sentiment		non-null	float64
27	reddit_buying_intent		non-null	float64
28	reddit_selling_intent		non-null	float64
29	reddit_uncertainty_intent		non-null	float64
30	reddit_urgency_intent		non-null	float64
31	reddit_prediction_intent		non-null	float64
32	reddit_fear_intent		non-null	float64
33	reddit_greed_intent		non-null	float64
34	reddit_question_intent		non-null	float64
35	reddit_action_intent		non-null	float64
36	Target		non-null	int64
37	pct_change		non-null	float64
38	finbert_final_sentiment		non-null	float64
39	total_buying_intent		non-null	float64
40	total_selling_intent		non-null	float64
41	total_uncertainty_intent		non-null	float64
42	total_urgency_intent		non-null	float64
43	total_prediction_intent		non-null	float64
44	total_fear_intent		non-null	float64
45	total_greed_intent		non-null	float64
46	total_question_intent		non-null	float64
47 48	total_action_intent		non-null	float64
48	sentiment_minus_uncertainty		non-null	float64
49 50	sentiment_minus_fear		non-null	float64
50	sentiment_minus_action		non-null	float64
51 52	sentiment_minus_urgency		non-null	float64
·				
<pre>dtypes: datetime64[ns](1), float64(49), int64(3) memory usage: 160.8 KB</pre>				

```
In [9]:
         df_test.isnull().sum()
Out[9]: Date
                                          0
                                          0
         0pen
                                          0
         High
         Low
                                          0
                                          0
         Close
         Volume
                                          0
         Adj Close
                                          0
         Log_Returns
                                          0
         Volatility_Log_10
                                          0
         cl-op
                                          0
         hi-lo
                                          0
         Label
                                          0
         vader_news_sentiment
                                          0
         FinBERT_news_sentiment
                                          0
         Smart_news_sentiment
                                          0
         news_buying_intent
                                          0
         news_selling_intent
                                          0
         news_uncertainty_intent
                                          0
         news_urgency_intent
                                          0
         news_prediction_intent
                                          0
         news_fear_intent
                                          0
         news_greed_intent
                                          0
         news question intent
                                          0
                                          0
         news_action_intent
         vader_reddit_sentiment
         FinBERT_reddit_sentiment
                                          0
         Smart_reddit_sentiment
                                          0
                                          0
         reddit_buying_intent
         reddit_selling_intent
         reddit_uncertainty_intent
                                          0
         reddit_urgency_intent
                                          0
                                          0
         reddit_prediction_intent
         reddit_fear_intent
                                          0
         reddit_greed_intent
                                          0
                                          0
         reddit_question_intent
         reddit_action_intent
                                          0
         Target
         pct_change
                                          0
         finbert_final_sentiment
                                          0
         total_buying_intent
                                          0
                                          0
         total_selling_intent
         total_uncertainty_intent
                                          0
         total_urgency_intent
                                          0
                                          0
         total_prediction_intent
         total_fear_intent
                                          0
         total_greed_intent
                                          0
         total_question_intent
         total_action_intent
                                          0
         sentiment_minus_uncertainty
                                          0
         sentiment_minus_fear
         sentiment_minus_action
                                          0
         sentiment minus urgency
                                          0
         sentiment_minus_prediction
         dtype: int64
```

```
In [10]: df_test.tail()
```

## Out[10]:

```
Open
                                 High
                                                           Close
                                                                     Volume
                                                                                Adj Close
     Date
                                               Low
    2016-
383
           17830.500000
                        17893.279297 17731.349609 17732.480469 101690000 17732.480469
    06-13
    2016-
384
           17710.769531 17733.919922 17595.789062 17674.820312
                                                                   93740000 17674.820312
    06-14
    2016-
385
           17703.650391 17762.960938 17629.009766 17640.169922
                                                                   94130000 17640.169922
    06-15
    2016-
           17602.230469 17754.910156 17471.289062 17733.099609
                                                                   91950000 17733.099609
    06-16
    2016-
           17733.439453 17733.439453 17602.779297 17675.160156 248680000 17675.160156
387
    06 - 17
```

## 5 rows × 53 columns

**\*** 

```
In [20]: df test.columns
Out[20]: Index(['Date', 'Open', 'High', 'Low', 'Close', 'Volume', 'Adj Close',
                  'Log_Returns', 'Volatility_Log_10', 'cl-op', 'hi-lo', 'Label',
                  'vader_news_sentiment', 'FinBERT_news_sentiment',
                  'Smart_news_sentiment', 'news_buying_intent', 'news_selling_inten
          t',
                 'news_uncertainty_intent', 'news_urgency_intent',
'news_prediction_intent', 'news_fear_intent', 'news_greed_intent',
                  'news_question_intent', 'news_action_intent', 'vader_reddit_sentime
          nt',
                  'FinBERT_reddit_sentiment', 'Smart_reddit_sentiment',
                  'reddit_buying_intent', 'reddit_selling_intent',
                  'reddit_uncertainty_intent', 'reddit_urgency_intent',
                  'reddit_prediction_intent', 'reddit_fear_intent', 'reddit_greed_int
          ent',
                  'reddit_question_intent', 'reddit_action_intent', 'Target',
                  'pct_change', 'finbert_final_sentiment', 'total_buying_intent',
                  'total_selling_intent', 'total_uncertainty_intent',
                  'total_urgency_intent', 'total_prediction_intent', 'total_fear_inte
          nt',
                  'total_greed_intent', 'total_question_intent', 'total_action_inten
          t',
                  'sentiment_minus_uncertainty', 'sentiment_minus_fear',
                  'sentiment_minus_action', 'sentiment_minus_urgency',
                  'sentiment minus prediction'],
                dtype='object')
```

```
In [25]: df_lstm = df_test.drop(columns=['Date', 'Label', 'Target']).copy()
```

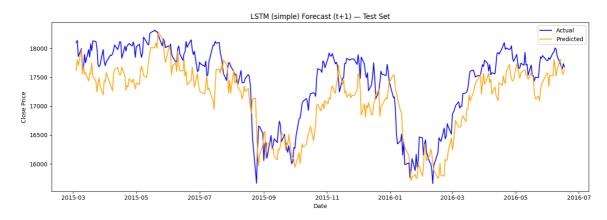
```
In [26]:
         df_lstm.columns
Out[26]: Index(['Open', 'High', 'Low', 'Close', 'Volume', 'Adj Close', 'Log_Return
         s',
                 'Volatility_Log_10', 'cl-op', 'hi-lo', 'vader_news_sentiment',
                 'FinBERT_news_sentiment', 'Smart_news_sentiment', 'news_buying_inte
         nt',
                 'news_selling_intent', 'news_uncertainty_intent', 'news_urgency_int
         ent',
                 'news_prediction_intent', 'news_fear_intent', 'news_greed_intent',
                 'news_question_intent', 'news_action_intent', 'vader_reddit_sentime
         nt',
                 'FinBERT_reddit_sentiment', 'Smart_reddit_sentiment',
                 'reddit_buying_intent', 'reddit_selling_intent',
                 'reddit_uncertainty_intent', 'reddit_urgency_intent',
                 'reddit_prediction_intent', 'reddit_fear_intent', 'reddit_greed int
         ent',
                 'reddit_question_intent', 'reddit_action_intent', 'pct_change',
                 'finbert_final_sentiment', 'total_buying_intent',
                 'total_selling_intent', 'total_uncertainty_intent',
                 'total_urgency_intent', 'total_prediction_intent', 'total_fear_inte
         nt',
                 'total_greed_intent', 'total_question_intent', 'total_action_inten
         t',
                 'sentiment_minus_uncertainty', 'sentiment_minus_fear',
                 'sentiment_minus_action', 'sentiment_minus_urgency',
                 'sentiment_minus_prediction'],
                dtype='object')
```

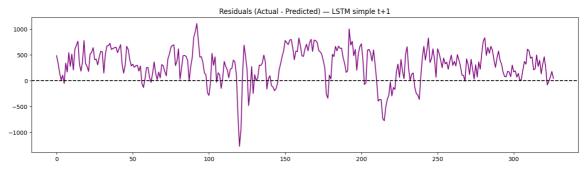
```
def load_and_test_lstm(model_dir, model_type='simple', forecast_horizon=1,
df_lstm=None, window_size=60):
    assert model_type in ['simple', 'stacked'], "model_type must be 'simpl
e' or 'stacked'"
    model_name = f"lstm_tplus{forecast_horizon}_{model_type}"
    model_path = os.path.join(model_dir, f"{model_name}.keras")
    scalerX_path = os.path.join(model_dir, f"{model_name}_scalerX.pkl")
    scalerY_path = os.path.join(model_dir, f"{model_name}_scalerY.pkl")
    if not all(os.path.exists(p) for p in [model_path, scalerX_path, scaler
Y_path]):
        return None
    print(f"\n    Testing {model name}")
    # Load model and scalers
    model = load_model(model_path)
    X_scaler = joblib.load(scalerX_path)
    y_scaler = joblib.load(scalerY_path)
    # Scale test data
    X_scaled = X_scaler.transform(df_lstm)
    y_scaled = y_scaler.transform(df_lstm['Close'].values.reshape(-1, 1))
    # Create test sequences
    X_{seq}, y_{seq} = [], []
    for i in range(window_size, len(X_scaled) - forecast_horizon):
        X_seq.append(X_scaled[i - window_size:i])
        y_seq.append(y_scaled[i + forecast_horizon])
    X_seq, y_seq = np.array(X_seq), np.array(y_seq)
    # Predict
    y pred scaled = model.predict(X seq)
    y_pred = y_scaler.inverse_transform(y_pred_scaled)
    y_true = y_scaler.inverse_transform(y_seq.reshape(-1, 1))
    date_series = df_test['Date'].iloc[window_size + forecast_horizon: wind
ow size + forecast horizon + len(y true)].values
    # Evaluation
    r2 = r2_score(y_true, y_pred)
    rmse = np.sqrt(mean_squared_error(y_true, y_pred))
    mae = mean_absolute_error(y_true, y_pred)
    print(f" | Test R<sup>2</sup>: {r2:.4f}, RMSE: {rmse:.2f}, MAE: {mae:.2f}")
    # Plot predictions
    plt.figure(figsize=(14, 5))
    plt.plot(date_series, y_true, label='Actual', color='blue')
    plt.plot(date_series, y_pred, label='Predicted', color='orange')
    plt.title(f"LSTM ({model_type}) Forecast (t+{forecast_horizon}) - Test
Set")
    plt.xlabel("Date")
    plt.ylabel("Close Price")
    plt.legend()
    plt.tight_layout()
    plt.show()
    # Plot residuals
    residuals = y_true.flatten() - y_pred.flatten()
```

```
plt.figure(figsize=(14, 4))
    plt.plot(residuals, color='purple')
    plt.axhline(0, linestyle='--', color='black')
    plt.title(f"Residuals (Actual - Predicted) - LSTM {model_type} t+{forec
ast_horizon}")
    plt.tight_layout()
    plt.show()
    return {"r2": r2, "rmse": rmse, "mae": mae}
model dir = "B:/DCU/Practicum/Proj/Models"
# Simple LSTM evaluation
for h in [1, 3, 7]:
    result = load_and_test_lstm(model_dir, model_type='simple', forecast_ho
rizon=h, df_lstm=df_lstm)
    if result is not None:
        print(result)
# Stacked LSTM evaluation
for h in [1, 3, 7]:
    result = load_and_test_lstm(model_dir, model_type='stacked', forecast_h
orizon=h, df_lstm=df_lstm)
    if result is not None:
        print(result)
```

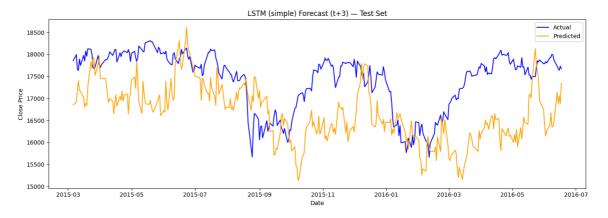
b:\DCU\Practicum\Proj\App\venv\_3\_11\Lib\site-packages\sklearn\utils\valida tion.py:2742: UserWarning: X has feature names, but MinMaxScaler was fitte d without feature names warnings.warn(

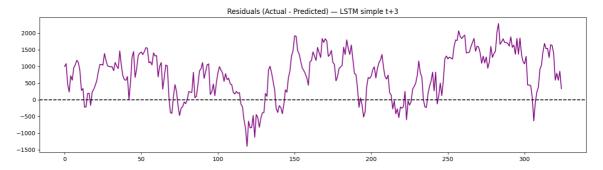
**11/11** — **0s** 31ms/step Test R<sup>2</sup>: 0.5364, RMSE: 444.72, MAE: 376.64





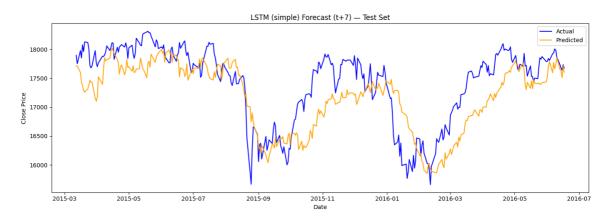
b:\DCU\Practicum\Proj\App\venv\_3\_11\Lib\site-packages\sklearn\utils\valida
tion.py:2742: UserWarning: X has feature names, but MinMaxScaler was fitte
d without feature names
warnings.warn(

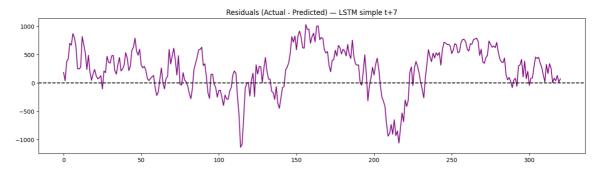




b:\DCU\Practicum\Proj\App\venv\_3\_11\Lib\site-packages\sklearn\utils\valida
tion.py:2742: UserWarning: X has feature names, but MinMaxScaler was fitte
d without feature names
warnings.warn(

11/11 — 0s 28ms/step Test R<sup>2</sup>: 0.4921, RMSE: 467.12, MAE: 385.83

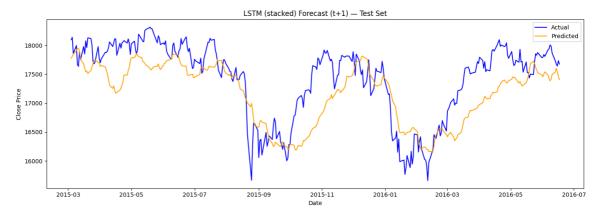


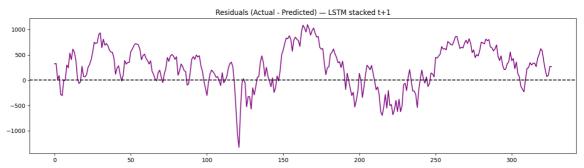


Testing lstm\_tplus1\_stacked

b:\DCU\Practicum\Proj\App\venv\_3\_11\Lib\site-packages\sklearn\utils\valida tion.py:2742: UserWarning: X has feature names, but MinMaxScaler was fitte d without feature names warnings.warn(

**11/11 1s** 34ms/step **1l** Test R<sup>2</sup>: 0.4652, RMSE: 477.63, MAE: 392.49

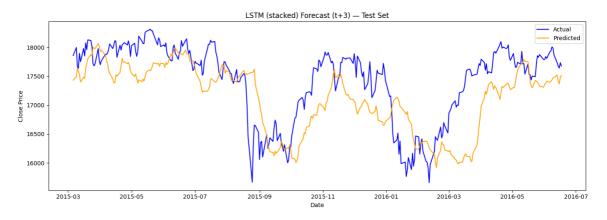


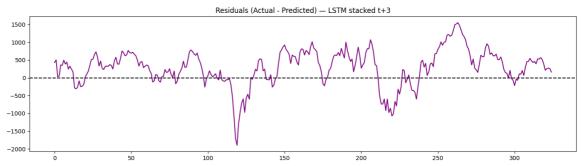


## Testing lstm\_tplus3\_stacked

b:\DCU\Practicum\Proj\App\venv\_3\_11\Lib\site-packages\sklearn\utils\valida
tion.py:2742: UserWarning: X has feature names, but MinMaxScaler was fitte
d without feature names
warnings.warn(

**11/11 1s** 33ms/step **Test** R<sup>2</sup>: 0.2065, RMSE: 581.59, MAE: 470.92

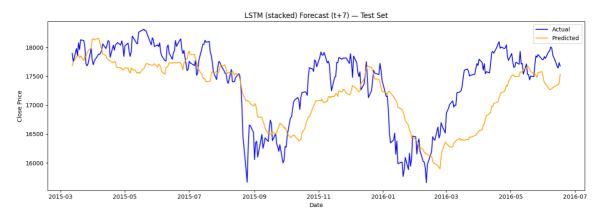


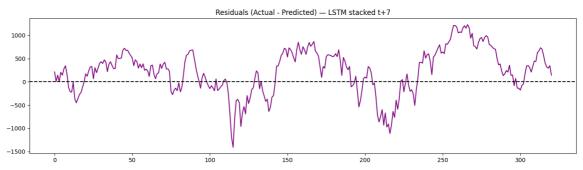


Testing lstm\_tplus7\_stacked

b:\DCU\Practicum\Proj\App\venv\_3\_11\Lib\site-packages\sklearn\utils\valida tion.py:2742: UserWarning: X has feature names, but MinMaxScaler was fitte d without feature names warnings.warn(

```
11/11 1s 34ms/step 1l Test R<sup>2</sup>: 0.3454, RMSE: 530.32, MAE: 437.21
```





```
In [30]: df_tcn.columns
```

```
Out[30]: Index(['Close', 'Log_Returns', 'vader_news_sentiment',
                  'FinBERT_news_sentiment', 'Smart_news_sentiment', 'news_buying_inte
          nt',
                  'news_selling_intent', 'news_uncertainty_intent', 'news_urgency_int
          ent',
                  'news_prediction_intent', 'news_fear_intent', 'news_greed_intent',
                  'news_question_intent', 'news_action_intent', 'vader_reddit_sentime
          nt',
                  'FinBERT reddit sentiment', 'Smart reddit sentiment',
                  'reddit_buying_intent', 'reddit_selling_intent',
                  'reddit_uncertainty_intent', 'reddit_urgency_intent',
                  'reddit_prediction_intent', 'reddit_fear_intent', 'reddit_greed_int
          ent',
                  'reddit_question_intent', 'reddit_action_intent',
'finbert_final_sentiment', 'total_selling_intent',
                  'total_uncertainty_intent', 'total_urgency_intent', 'total_fear_int
          ent',
                  'total_greed_intent', 'total_question_intent', 'total_action_inten
          t',
                  'sentiment_minus_uncertainty', 'sentiment_minus_fear',
                  'sentiment_minus_action', 'sentiment_minus_urgency',
                  'sentiment minus prediction'],
                dtype='object')
```

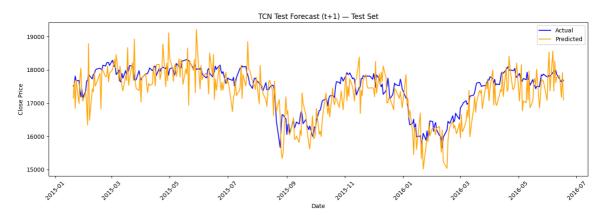
```
In [ ]: def test_tcn_logreturn_model(df_tcn, forecast_horizon=1, window_size=30):
            horizon}")
            model name = f"tcn logret tplus{forecast horizon}"
            base_path = "B:/DCU/Practicum/Proj/Models"
            # === Load model and scalers ===
            model = load_model(f"{base_path}/{model_name}.keras", custom_objects=
        {"TCN": TCN})
            X_scaler = joblib.load(f"{base_path}/{model_name}_scalerX.pkl")
            y_scaler = joblib.load(f"{base_path}/{model_name}_scalerY.pkl")
            # === Use df_tcn directly ===
            X_scaled = X_scaler.transform(df_tcn.values)
            y_scaled = y_scaler.transform(df_tcn['Log_Returns'].shift(-forecast_hor
        izon).dropna().values.reshape(-1, 1))
            # === Create sequences ===
            X_{seq}, y_{seq} = [], []
            valid_len = min(len(X_scaled), len(y_scaled))
            for i in range(window_size, valid_len):
                X_seq.append(X_scaled[i - window_size:i])
                y_seq.append(y_scaled[i])
            X_seq, y_seq = np.array(X_seq), np.array(y_seq)
            # === Predict
            y_pred_scaled = model.predict(X_seq)
            y_pred_log = y_scaler.inverse_transform(y_pred_scaled).flatten()
            y_true_log = y_scaler.inverse_transform(y_seq.reshape(-1, 1)).flatten()
            # === Get Close_t from df_tcn
            close_t = df_tcn['Close'].iloc[window_size - 1 : window_size - 1 + len
        (y_pred_log)].values
            # === Predicted & Actual Close at t+h
            y_pred_close = close_t * np.exp(y_pred_log)
            y_true_close = df_tcn['Close'].shift(-forecast_horizon).dropna().iloc[w
        indow_size:].values
            # === Dates for plotting
            date_series = df_test['Date'].iloc[window_size + forecast_horizon: wind
        ow_size + forecast_horizon + len(y_pred_close)]
            # === Metrics
            r2 = r2_score(y_true_close, y_pred_close)
            rmse = np.sqrt(mean squared error(y true close, y pred close))
            mae = mean_absolute_error(y_true_close, y_pred_close)
            print(f" TCN Test R²: {r2:.4f}, RMSE: {rmse:.2f}, MAE: {mae:.2f}")
            # === Save metrics
            metrics path = os.path.join(base path, f"{model name} test metrics.tx
            with open(metrics path, "w") as f:
                f.write(f"Test Forecast Horizon = t+{forecast_horizon}\n")
                f.write(f"Test R^2 = {r2:.4f}\n")
                f.write(f"Test RMSE = {rmse:.2f}\n")
                f.write(f"Test MAE = {mae:.2f}\n")
            # === Plot
```

```
plot_path = os.path.join(base_path, f"{model_name}_testplot.png")
    plt.figure(figsize=(14, 5))
    plt.plot(date_series, y_true_close, label='Actual', color='blue')
    plt.plot(date_series, y_pred_close, label='Predicted', color='orange')
    plt.title(f"TCN Test Forecast (t+{forecast_horizon}) - Test Set")
    plt.xlabel("Date")
    plt.ylabel("Close Price")
    plt.xticks(rotation=45)
    plt.legend()
    plt.tight_layout()
    plt.savefig(plot_path)
    plt.show()
    print(f" Saved test plot to {plot_path}")
    return {"r2": r2, "rmse": rmse, "mae": mae}
# Final TCN test loop
for h in [1, 3, 7]:
    test_tcn_logreturn_model(df_tcn=df_tcn, forecast_horizon=h)
```

→ Testing TCN model for Log Returns → Close at t+1

b:\DCU\Practicum\Proj\App\venv\_3\_11\Lib\site-packages\sklearn\utils\valida tion.py:2749: UserWarning: X does not have valid feature names, but MinMax Scaler was fitted with feature names warnings.warn(

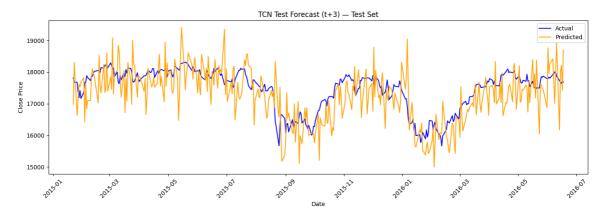
**12/12 1s** 33ms/step TCN Test R<sup>2</sup>: 0.3939, RMSE: 499.45, MAE: 404.61



- Testing TCN model for Log Returns → Close at t+3

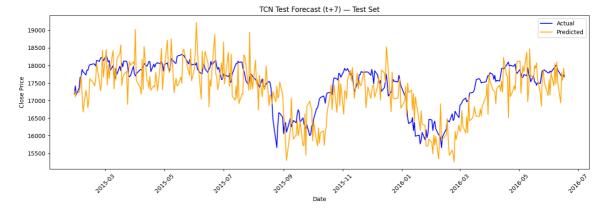
b:\DCU\Practicum\Proj\App\venv\_3\_11\Lib\site-packages\sklearn\utils\valida tion.py:2749: UserWarning: X does not have valid feature names, but MinMax Scaler was fitted with feature names warnings.warn(

**12/12** — **1s** 29ms/step TCN Test R<sup>2</sup>: -0.0241, RMSE: 651.03, MAE: 525.44



- Saved test plot to B:/DCU/Practicum/Proj/Models\tcn\_logret\_tplus3\_testplot.png
- Testing TCN model for Log Returns → Close at t+7

b:\DCU\Practicum\Proj\App\venv\_3\_11\Lib\site-packages\sklearn\utils\valida tion.py:2749: UserWarning: X does not have valid feature names, but MinMax Scaler was fitted with feature names warnings.warn(



Saved test plot to B:/DCU/Practicum/Proj/Models\tcn\_logret\_tplus7\_testplot.png