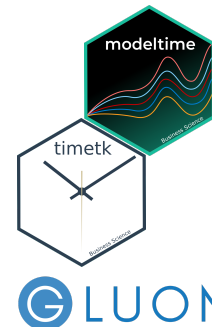


Forecasting with External Regressors Workflow

If you want to apply this workflow to a *real business analysis*, take the [High-Performance Forecasting Course](#) through Business Science University.

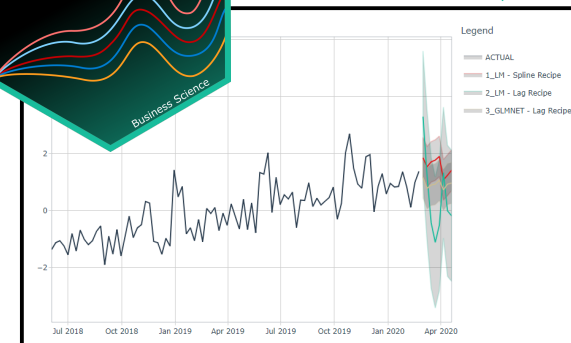
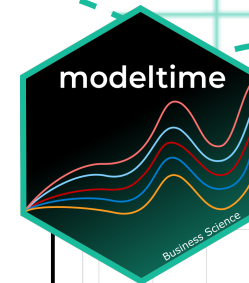
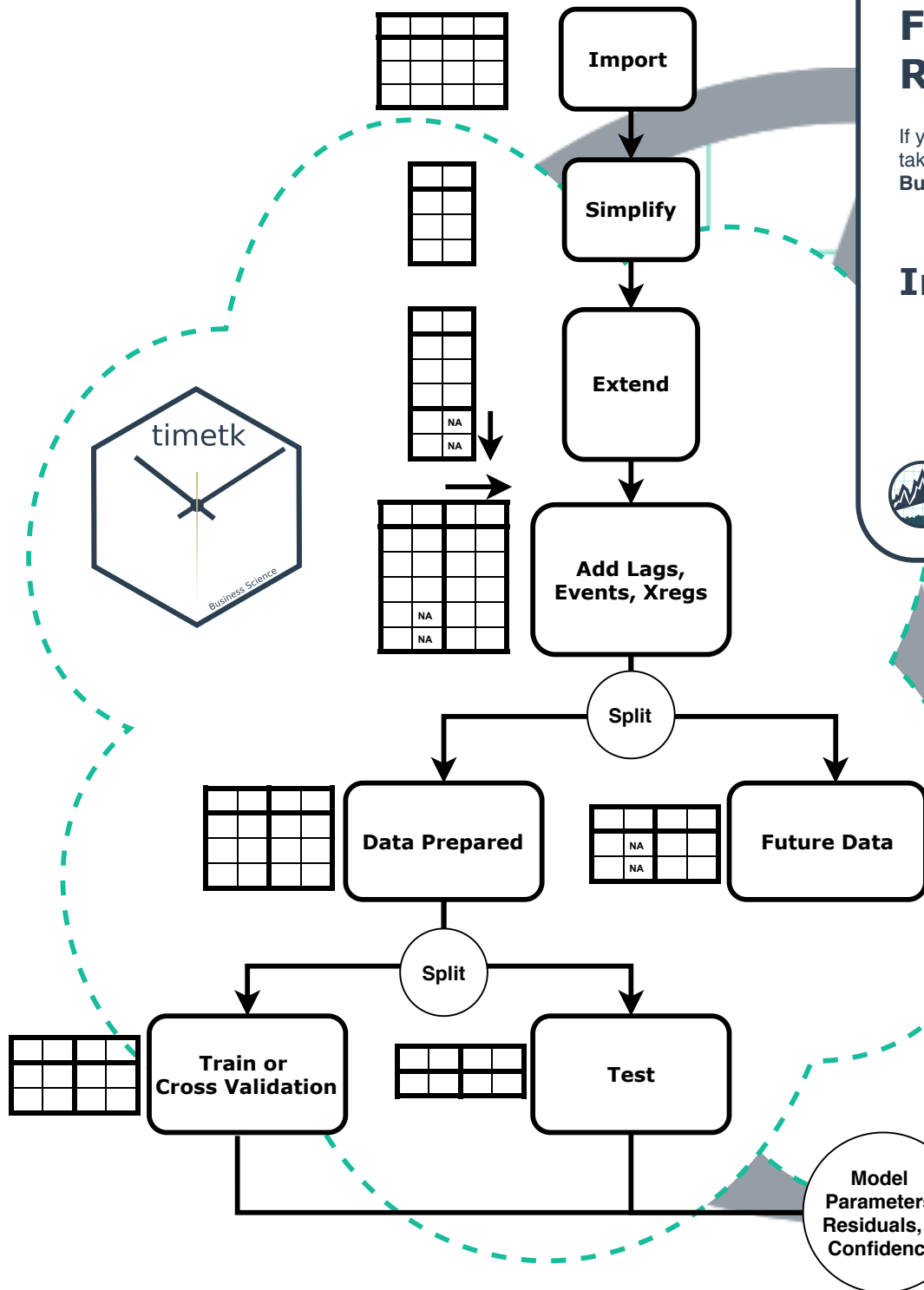


Important Resources

- **Timetk:** <https://business-science.github.io/timetk/>
- **Modeltime:** <https://business-science.github.io/modeltime/>
- **GluonTS:** <https://gluon-ts.mxnet.io/index.html>



Business Science University
university.business-science.io



MODELTIME Workflow

Create Modeltime Table
modeltime_table()

Calibrate
modeltime_calibrate()

Refit
modeltime_refit()

```
# Modeltime Table
# A tibble: 5 x 3
  .model_id .model .model_desc
  <int> <list> <chr>
1 1 <fit(+)> ARIMA(0,1,1)(0,1,1)[12]
2 2 <fit(+)> ARIMA(0,1,1)(0,1,1)[12] W/ XGBOOST ERRORS
3 3 <fit(+)> ETS(M,A,A)
4 4 <fit(+)> LM
5 5 <fit(+)> EARTH
```

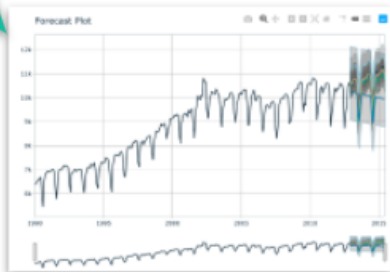
```
# Modeltime Table
# A tibble: 6 x 5
  .model_id .model .model_desc .type .calibration_data
  <int> <list> <chr> <chr> <list>
1 1 <fit(+)> ARIMA(0,1,1)(0,1,1)[12] Test <tibble [31 x 4]>
2 2 <fit(+)> ARIMA(0,1,1)(0,1,1)[12] W/ XGBOOST ERRORS Test <tibble [31 x 4]>
3 3 <fit(+)> ETS(M,A,A) Test <tibble [31 x 4]>
4 4 <fit(+)> PROPHET Test <tibble [31 x 4]>
5 5 <fit(+)> LM Test <tibble [31 x 4]>
6 6 <fit(+)> EARTH Test <tibble [31 x 4]>
```

```
# Modeltime Table
# A tibble: 6 x 5
  .model_id .model .model_desc .type .calibration_data
  <int> <list> <chr> <chr> <list>
1 1 <fit(+)> ARIMA(0,1,1)(0,1,1)[12] Test <tibble [31 x 4]>
2 2 <fit(+)> ARIMA(0,1,1)(0,1,1)[12] W/ XGBOOST ERRORS Test <tibble [31 x 4]>
3 3 <fit(+)> ETS(A,A,A) Test <tibble [31 x 4]>
4 4 <fit(+)> PROPHET Test <tibble [31 x 4]>
5 5 <fit(+)> LM Test <tibble [31 x 4]>
6 6 <fit(+)> EARTH Test <tibble [31 x 4]>
```

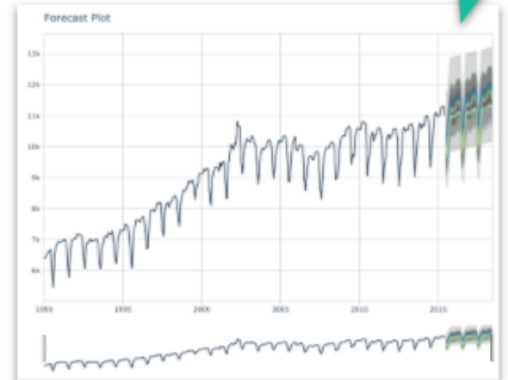
Forecast Test Set
modeltime_forecast()

Test Accuracy
modeltime_accuracy()

Forecast Future
modeltime_forecast()



Accuracy Table								
.model_id	.model_desc	.type	mae	mape	mase	smape	rmse	rsq
1	ARIMA(0,1,1)(0,1,1)[12]	Test	161.33	1.41	0.52	1.43	197.71	0.93
2	ARIMA(0,1,1)(0,1,1)[12] W/ XGBOOST ERRORS	Test	147.04	1.37	0.50	1.39	191.84	0.93
3	ETS(M,A,A)	Test	77.00	0.73	0.26	0.73	90.27	0.98
4	PROPHET	Test	177.51	1.70	0.61	1.70	234.65	0.88
5	LM	Test	629.12	6.01	2.15	5.81	657.19	0.91
6	EARTH	Test	709.83	6.59	2.42	6.86	782.82	0.55



plot_modeltime_forecast()

table_modeltime_accuracy()

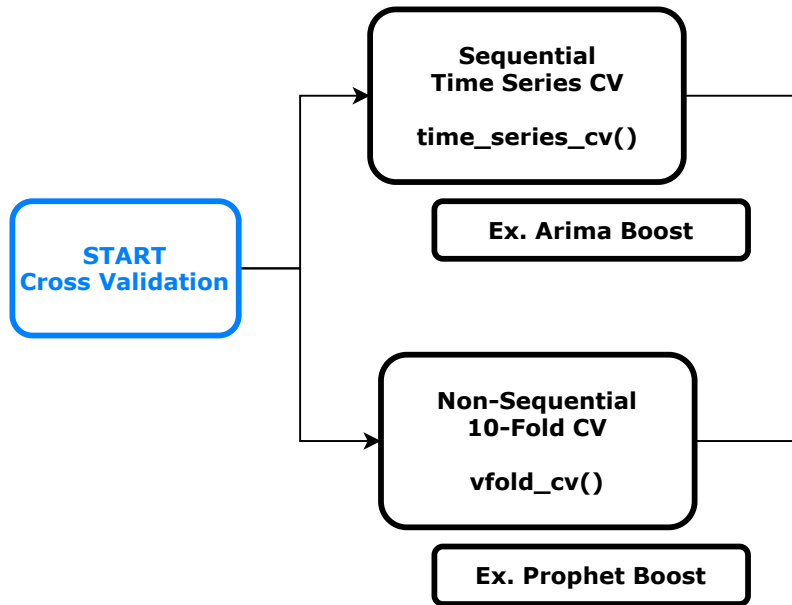
plot_modeltime_forecast()





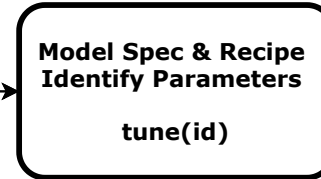
1. Cross Validation Plan

Can be used for more than 1 model



2. Identify Tuning Parameters

Specific to Each Model



3. Make Grid for Parameters

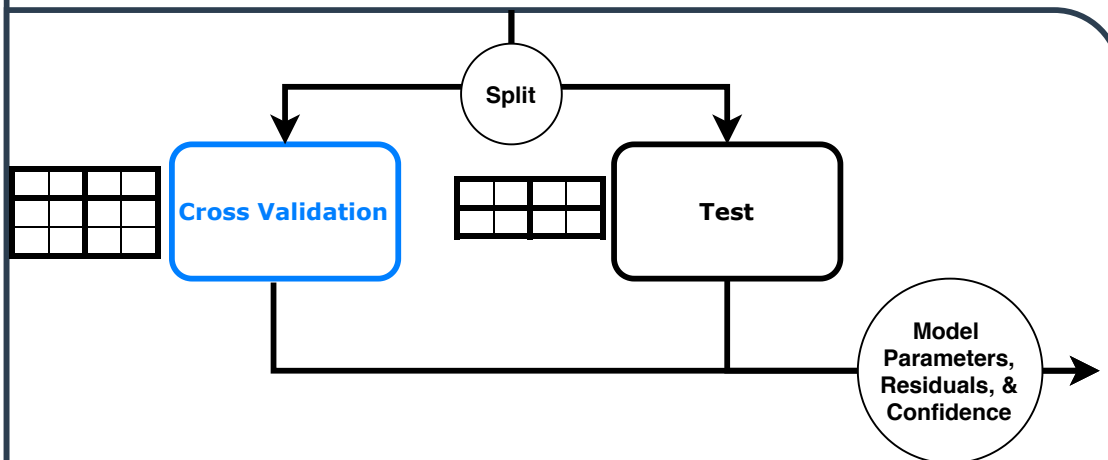
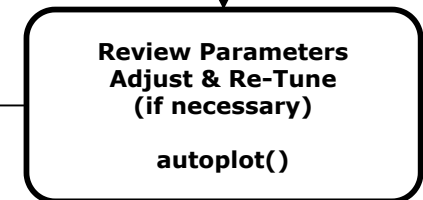
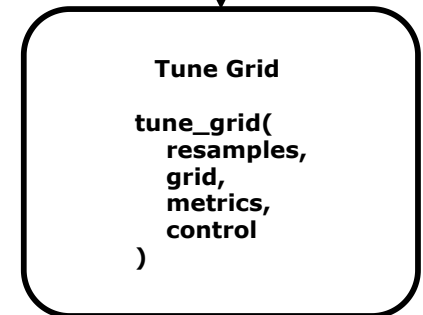
Specific to Each Model



4. Hyperparameter Tune

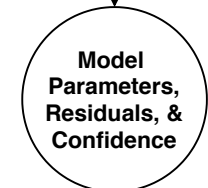
Specific to Each Model

Use Parallel Processing to Speed Up



Hyperparameter Tuning & Cross-Validation for Time Series

Cross-Validation and Hyperparameter Tuning are absolutely critical for Machine Learning algorithms.





MODELTIME ENSEMBLE

Multi-Level Stacking



Level 3: Weighted Stack

`ensemble_weighted()`
`ensemble_average()`

$$w1*m1 + w2*m2 + w3*m3 + \dots$$

Level 2: Stacking Algorithms

`ensemble_model_spec()`
`modeltime_fit_resamples()`

Linear Stack

Tree Stack

Level 1: Sub-Models

ARIMA

GLMNET

SVM

XGBoost

`arima_reg()`

`linear_reg()`

`svm_rbf()`

`boost_tree()`

