TFT-Copy1

July 20, 2023

```
[19]: library(zoo)
      library(tidyverse)
      library(tsibble)
      library(remotes)
      library(dplyr)
      library(caret)
      library(tidymodels)
      library(tft)
      library(lubridate)
      library(timetk)
      library(tibbletime)
      library(luz)
      library(torch)
      set.seed(565)
[20]: options(torch.threshold_call_gc = 6000)
[21]: #read comodity prices and convert to xts object
      com_prices_data = read.csv("commodity-prices-2016.csv")
      com_prices_data = select(com_prices_data, Date, #Crude.Oil.
       ⇒petroleum, Aluminum, Bananas, Barley, Beef, Coal, Cocoa. beans, Coffee. Other. Mild.
       -Arabicas, Coffee. Robusta, Rapeseed. oil, Copper, Cotton, Fishmeal, Groundnuts.
       ⇔peanuts, Hides, Lamb, Lead, Soft. Logs, Hard. Logs, Maize. corn, Olive.
       \neg Oil, Oranges, Palm.oil, Poultry. chicken, Rice, Rubber, Fish.salmon, Hard.
       Sawnwood, Soft. Sawnwood, Shrimp, Soybean. Meal, Soybean. Oil, Soybeans, Sunflower.
       ⇔oil, Tea, Tin, Uranium,
                                Wheat)
      com_prices_data$Date = as.Date(com_prices_data$Date, format = "%Y-%m-%d")
[22]: require(curl) # To load from url for data.table's fread
      require(data.table) # High performance data frame
      require(zoo) # High performance data frame
      com_prices_data2 = as.data.table(com_prices_data)
      com_prices_data2[, Day := as.Date(Date)]
      setkey(com prices data2, Day)
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start = com_prices_data2$Date[1]
      end = tail(com_prices_data2$Date, 1)
      date.indexes = seq(
        from = start,
        to = end,
        by = "days"
      )
      com_prices_data_daily = com_prices_data2[J(date.indexes), roll = 31]
[23]: com_prices_data3 <- com_prices_data_daily %>%
        mutate(Week = lubridate::floor_date(Day, unit = "week")) %>%
        group_by(Week) %>%
        summarise(across(everything(), .fns = ~mean(.x, na.rm = TRUE)), .groups =__

¬"drop") %>%
      pivot longer(c(
      \#Crude.Oil.petroleum, Aluminum, Bananas, Barley, Beef, Coal, Cocoa.beans, Coffee.Other.
       Mild. Arabicas, Coffee. Robusta, Rapeseed. oil, Copper, Cotton, Fishmeal, Groundnuts.
       →peanuts, Hides, Lamb, Lead, Soft. Logs, Hard. Logs, Maize. corn, Olive.
       \hookrightarrow Oil, Oranges, Palm.oil, Poultry.chicken, Rice, Rubber, Fish.salmon, Hard.
       Sawnwood, Soft. Sawnwood, Shrimp, Soybean. Meal, Soybean. Oil, Soybeans, Sunflower.
       ⇔oil, Tea, Tin, Uranium,
          Wheat)) %>%
             rename("Price" = "value", "Commodity" = "name") %>%
      group by(Week, Commodity) %>%
      summarise(across(everything(), .fns = ~mean(.x, na.rm = TRUE)), .groups =___

¬"drop")
      tt_p = com_prices_data3
[24]: tt = tt p[,-(3:4)]
      colnames(tt)[1] = "date"
[25]: last_date <- max(tt$date)</pre>
      train <- tt %>% filter(date <= (last date - lubridate::weeks(48)))</pre>
      valid <- tt %>% filter(date > (last_date - lubridate::weeks(48)),
                                        date <= (last_date - lubridate::weeks(12)))</pre>
      test <- tt %>% filter(date > (last_date - lubridate::weeks(12)))
      rec <- recipe(Price ~ ., data = train) %>%
        step_mutate(
          date_time_since_begining = as.numeric(difftime(
            time1 = date,
            time2 = lubridate::ymd(min(tt$date)),
            units = "weeks"
          )),
          date_week = as.factor(lubridate::week(date)),
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date_month = as.factor(lubridate::month(date)),
          date_wday = as.factor(lubridate::wday(date))
        step_normalize(all_numeric_predictors())
[26]: spec <- tft_dataset_spec(rec, train) %>%
        spec_covariate_index(date) %>%
        spec_covariate_key(Commodity) %>%
        spec_covariate_known(starts_with("date_")) %>%
        spec_time_splits(lookback = 5*35, horizon = 12)
[27]: spec <- prep(spec)
      spec
     A prepared_tft_dataset_spec> with:
       lookback = 175 and horizon = 12.
      The number of possible slices is 1,650
       Covariates:
       `index`: date
       `keys`: Commodity
       `static`:
       `known`: date_time_since_begining, date_week, date_month, and date_wday
       `unknown`:
      Variables that are not specified in other types are considered `unknown`.
      Call `transform()` to apply this spec to a different dataset.
[28]: model <- temporal_fusion_transformer(
        spec,
        hidden_state_size = 8,
        learn rate = 1e-3,
        \#dropout = 0.5,
        num_attention_heads = 1,
        num_lstm_layers = 1
[29]: fitted <- model %>%
        fit(
          transform(spec),
          valid_data = transform(spec, new_data = valid),
          epochs = 3,
            callbacks = list(
            luz::luz_callback_keep_best_model(monitor = "valid_loss"),
            luz::luz_callback_early_stopping(
              monitor = "valid_loss",
```

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patience = 5,
              min_delta = 0.001
            )
          ),
          verbose = TRUE,
          dataloader_options = list(batch_size = 32, num_workers = 4)
        )
     Epoch 1/3
     Train metrics: Loss: 0.364 - q10: 0.1986 - q50: 0.4283 - q90: 0.465
     Valid metrics: Loss: 0.1327 - q10: 0.1142 - q50: 0.1897 - q90: 0.0941
     Epoch 2/3
     Train metrics: Loss: 0.2342 - q10: 0.1067 - q50: 0.313 - q90: 0.2827
     Valid metrics: Loss: 0.0861 - q10: 0.0747 - q50: 0.1279 - q90: 0.0558
     Epoch 3/3
     Train metrics: Loss: 0.1911 - q10: 0.0956 - q50: 0.2603 - q90: 0.2175
     Valid metrics: Loss: 0.1558 - q10: 0.0431 - q50: 0.3349 - q90: 0.0893
[30]: fitted %>%
        luz::evaluate(
          transform(spec, new_data = test, past_data = bind_rows(train, valid)),
          metrics = list(luz_metric_rmse(),luz_metric_mae())
        )
     A `luz_module_evaluation`
       Results
     loss: 0.1523
     rmse: 1.2303
     mae: 1.9638
[31]: fitted
     A `luz_module_fitted`
       Time
     • Total time: 10m 26.9s
     • Avg time per training epoch: 3m 11.3s
       Results
     Metrics observed in the last epoch.
      Training:
     loss: 0.1911
     q10: 0.0956
     q50: 0.2603
     q90: 0.2175
```

Model

```
An `nn_module` containing 9,852 parameters.
```

Modules

- preprocessing: reprocessing> #1,112 parameters
- context: <static_context> #1,709 parameters
- temporal_selection: <selection> #4,693 parameters
- locality_enhancement: <locality_enhancement_layer> #1,472 parameters
- temporal_attn: <nn_module> #310 parameters
- position_wise: <nn_module> #528 parameters
- output_layer: <nn_module> #27 parameters

date

Parameters

• .check: Float [1:1]

[32]: forecasts <- generics::forecast(fitted, past_data = bind_rows(train, valid))
as.data.frame(forecasts)

Commodity .pred lower .pred

.pred upper

		•	•	•	
A data.frame: 12×5	< date >	<chr $>$	<dbl></dbl>	<dbl $>$	<dbl></dbl>
	2015-11-15	Wheat	144.7985	204.6912	215.4223
	2015-11-22	Wheat	146.1455	199.8800	217.9047
	2015-11-29	Wheat	142.8414	196.7768	218.2290
	2015-12-06	Wheat	143.5570	201.6434	217.0840
	2015-12-13	Wheat	146.8060	202.9227	216.1608
	2015-12-20	Wheat	143.9311	201.7064	216.6562
	2015 - 12 - 27	Wheat	143.5825	197.1212	218.7276
	2016-01-03	Wheat	143.4986	196.3618	218.8069
	2016-01-10	Wheat	147.6533	197.7174	218.2329
	2016-01-17	Wheat	142.8465	195.0666	219.6921
	2016-01-24	Wheat	144.1247	200.9608	217.1580
	2016-01-31	Wheat	144.0747	194.7415	219.4793

```
[33]: options(repr.plot.width = 10, repr.plot.height =10)
tt %>%
    filter(date > lubridate::ymd("2015-01-01")) %>%
    full_join(forecasts) %>%
    #filter(Commodity == "Copper") %>%
    ggplot(aes(x = date, y = Price)) +
    geom_line() +
    geom_line(aes(y = .pred), color = "green") +
    geom_ribbon(aes(ymin = .pred_lower, ymax = .pred_upper), alpha = 0.3)
    #facet_wrap(~Commodity)
#theme(text = element_text(size = 10), element_line(size =1))
```

```
Joining with `by = join_by(date, Commodity)`
Warning message:
```

[&]quot;Removed 45 rows containing missing values (`geom_line()`)."

