

# Untitled

2023-10-05

```
library(tidyverse)
```

```
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
## v dplyr      1.1.3      v readr      2.1.4
## v forcats    1.0.0      v stringr   1.5.0
## v ggplot2    3.4.3      v tibble    3.2.1
## v lubridate  1.9.3      v tidyr     1.3.0
## v purrr      1.0.2
```

```
## -- Conflicts ----- tidyverse_conflicts() --
```

```
## x dplyr::filter() masks stats::filter()
```

```
## x dplyr::lag()      masks stats::lag()
```

```
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become errors
```

```
library(mice)
```

```
##
```

```
## Attaching package: 'mice'
```

```
##
```

```
## The following object is masked from 'package:stats':
```

```
##
```

```
##      filter
```

```
##
```

```
## The following objects are masked from 'package:base':
```

```
##
```

```
##      cbind, rbind
```

```
sales <- data.frame(
```

```
  date = c("2022-01-01", "2022-01-02", "2022-01-03", "2022-01-04", "2022-01-05"),
```

```
  store_cd= c(1, 2, 3, 4, 5),
```

```
  product_cd= c(1, 2, 3, 4, 5),
```

```
  qty= c(10, 12, 9, 12, 8),
```

```
  sales= c(30, 60, 45, 24, 32)
```

```
)
```

```
stores <- data.frame(
```

```
  store_cd= c(1, 2, 3, 4, 6),
```

```
  address= c("1 main st", "20 side st", "19 square blvd", "101 first st", "1002 retail ave"),
```

```
  city= c("Main", "East", "West", "North", "South"),
```

```
  open_hours= c("7-23", "7-23", "9-21", "9-21", "9-21")
```

```
)
```

```
products <- data.frame(
```

```
  product_cd= c(1, 2, 3, 4, 6),
```

```
  description= c("Soft drink", "Frozen snack", "Fruit", "Water", "Fruit 2"),
```

```
  unit_price= c(3.0, 5.0, 5.0, 2.0, 4.0),
```

```
  unit_measure= c("each", "each", "kg", "each", "kg")
```

```
)

sales %>% left_join(products, by="product_cd")

##           date store_cd product_cd qty sales description unit_price unit_measure
## 1 2022-01-01         1          1  10    30   Soft drink          3         each
## 2 2022-01-02         2          2  12    60  Frozen snack          5         each
## 3 2022-01-03         3          3   9    45      Fruit           5          kg
## 4 2022-01-04         4          4  12    24      Water           2         each
## 5 2022-01-05         5          5   8    32      <NA>          NA        <NA>

sales %>% right_join(stores, by="store_cd")

##           date store_cd product_cd qty sales address city open_hours
## 1 2022-01-01         1          1  10    30   1 main st Main      7-23
## 2 2022-01-02         2          2  12    60   20 side st East      7-23
## 3 2022-01-03         3          3   9    45  19 square blvd West     9-21
## 4 2022-01-04         4          4  12    24  101 first st North    9-21
## 5      <NA>         6          NA   NA   NA 1002 retail ave South    9-21

sales %>% inner_join(stores, by="store_cd")

##           date store_cd product_cd qty sales address city open_hours
## 1 2022-01-01         1          1  10    30   1 main st Main      7-23
## 2 2022-01-02         2          2  12    60   20 side st East      7-23
## 3 2022-01-03         3          3   9    45  19 square blvd West     9-21
## 4 2022-01-04         4          4  12    24  101 first st North    9-21

sales %>% full_join(stores)

## Joining with `by = join_by(store_cd)`

##           date store_cd product_cd qty sales address city open_hours
## 1 2022-01-01         1          1  10    30   1 main st Main      7-23
## 2 2022-01-02         2          2  12    60   20 side st East      7-23
## 3 2022-01-03         3          3   9    45  19 square blvd West     9-21
## 4 2022-01-04         4          4  12    24  101 first st North    9-21
## 5 2022-01-05         5          5   8    32      <NA> <NA>      <NA>
## 6      <NA>         6          NA   NA   NA 1002 retail ave South    9-21

sales %>% anti_join(products)

## Joining with `by = join_by(product_cd)`

##           date store_cd product_cd qty sales
## 1 2022-01-05         5          5   8    32

df_wide <- data.frame(
  project = c("project1", "project2", "project3"),
  Jan= sample(1000:2000, 3),
  Feb= sample(1000:2000, 3),
  Mar= sample(1000:2000, 3)
)

df_long <- df_wide %>%
  pivot_longer(cols=2:4,
    names_to = "months",
    values_to = "expenses")
```

```
df_wide_2 <- df_long %>%
  pivot_wider(names_from = "months",
              values_from = "expenses")
```

## Missing data

```
header <- c("age", "workclass", "fnlwgt", "education",
            "education_num", "marital_status", "occupation",
            "relationship", "race", "sex", "capital_gain",
            "capital_loss", "hours_per_week", "native_country", "target")
df <- read_csv("https://archive.ics.uci.edu/ml/machine-learning-databases/adult/adult.data",
               col_names=header, trim_ws=TRUE)
```

```
## Rows: 32561 Columns: 15
## -- Column specification -----
## Delimiter: ","
## chr (9): workclass, education, marital_status, occupation, relationship, rac...
## dbl (6): age, fnlwgt, education_num, capital_gain, capital_loss, hours_per_week
##
## i Use `spec()` to retrieve the full column specification for this data.
## i Specify the column types or set `show_col_types = FALSE` to quiet this message.
```

```
df <-df %>%
  mutate(workclass = na_if(workclass, "?"),
         occupation = na_if(occupation, "?"),
         native_country = na_if(native_country, "?"))
```

```
df_fill1 <- df %>%
  fill(workclass, occupation, native_country, .direction="down")
```

```
m_freq_workcls <- names(table(df$workclass))[which.max(table(df$workclass))]
m_freq_occup <- names(table(df$occupation))[which.max(table(df$occupation))]
df_fill2 <- df %>%
  replace_na(list(workclass = m_freq_workcls,
                 occupation = m_freq_occup))
```

```
df_no_na <- df %>% na.omit()
```

```
df_native <- df %>%
  drop_na(native_country)
```

```
data("txhousing")
txhousing$date <- date_decimal(txhousing$date, tz="GMT")
txhousing$city <- as.factor(txhousing$city)
```

```
idx <- which(rowSums(is.na(txhousing)) == 5)
txhousing <- txhousing[-idx,]
```

```
txhousing$sales[is.na(txhousing$sales)] <- median(txhousing$sales, na.rm=TRUE)
txhousing$volume[is.na(txhousing$volume)] <- median(txhousing$volume, na.rm=TRUE)
txhousing$median[is.na(txhousing$median)] <- median(txhousing$median, na.rm=TRUE)
```

```
impute <- mice(data.frame(txhousing[,7:8]), seed=123)
```

```
##
## iter imp variable
```

```
## 1 1 listings inventory
## 1 2 listings inventory
## 1 3 listings inventory
## 1 4 listings inventory
## 1 5 listings inventory
## 2 1 listings inventory
## 2 2 listings inventory
## 2 3 listings inventory
## 2 4 listings inventory
## 2 5 listings inventory
## 3 1 listings inventory
## 3 2 listings inventory
## 3 3 listings inventory
## 3 4 listings inventory
## 3 5 listings inventory
## 4 1 listings inventory
## 4 2 listings inventory
## 4 3 listings inventory
## 4 4 listings inventory
## 4 5 listings inventory
## 5 1 listings inventory
## 5 2 listings inventory
## 5 3 listings inventory
## 5 4 listings inventory
## 5 5 listings inventory
```

```
impute_data <- complete(impute, 1)
txhousing_clean <- txhousing %>%
  mutate(listings = impute_data[,1],
         inventory = impute_data[,2])
```

```
stocks <- tibble(
  year = c(2020, 2020, 2020, 2020, 2021, 2021, 2021),
  qtr = c( 1, 2, 3, 4, 2, 3, 4),
  price = c(1.88, 0.59, 0.35, NA, 0.92, 0.17, 2.66)
)
```

```
stocks %>%
  pivot_wider(names_from = qtr,
              values_from = price)
```

```
## # A tibble: 2 x 5
##   year   `1`   `2`   `3`   `4`
##   <dbl> <dbl> <dbl> <dbl> <dbl>
## 1  2020  1.88  0.59  0.35  NA
## 2  2021  NA    0.92  0.17  2.66
```

```
stocks %>% complete(year, qtr)
```

```
## # A tibble: 8 x 3
##   year qtr price
##   <dbl> <dbl> <dbl>
## 1  2020     1  1.88
## 2  2020     2  0.59
## 3  2020     3  0.35
## 4  2020     4  NA
```

```
## 5 2021      1 NA
## 6 2021      2 0.92
## 7 2021      3 0.17
## 8 2021      4 2.66
```

```
library(rvest)
```

```
##
## Attaching package: 'rvest'
## The following object is masked from 'package:readr':
##
##   guess_encoding
```

```
page <- "https://en.wikipedia.org/wiki/List_of_countries_by_GDP_(nominal)"
gdp <- rvest::read_html(page)
```

```
gdp_df <- gdp %>%
  html_elements(xpath='//*[@id="mw-content-text"]/div[1]/table[2]') %>%
  html_table()
```

```
library(httr)
library(jsonlite)
```

```
##
## Attaching package: 'jsonlite'
## The following object is masked from 'package:purrr':
##
##   flatten
```

```
url <- "https://api.fiscaldata.treasury.gov/services/api/fiscal_service/v1/accounting/mts/mts_table_1"
treasury_api <- GET(url)
```

```
result <- content(treasury_api, "text", encoding="UTF-8")
df_json <- fromJSON(result, flatten=TRUE)
df <- as.data.frame(df_json$data)
```

```
mtcars %>% select(hp, wt) %>% map(mean)
```

```
## $hp
## [1] 146.6875
##
## $wt
## [1] 3.21725
```

```
A <- mtcars[1:3,]
B <- mtcars[4:6,]
AB <- A %>% bind_rows(B)
```

```
A <- mtcars[1:5, 1:3]
B <- mtcars[1:5, 4:6]
AB <- A %>% bind_cols(B)
```

```
mtcars %>%
  mutate(transmission_type =
    case_when(
      am == 0 ~ "automatic",
      am == 1 ~ "manual"))
```

##	mpg	cyl	disp	hp	drat	wt	qsec	vs	am	gear	carb
## Mazda RX4	21.0	6	160.0	110	3.90	2.620	16.46	0	1	4	4
## Mazda RX4 Wag	21.0	6	160.0	110	3.90	2.875	17.02	0	1	4	4
## Datsun 710	22.8	4	108.0	93	3.85	2.320	18.61	1	1	4	1
## Hornet 4 Drive	21.4	6	258.0	110	3.08	3.215	19.44	1	0	3	1
## Hornet Sportabout	18.7	8	360.0	175	3.15	3.440	17.02	0	0	3	2
## Valiant	18.1	6	225.0	105	2.76	3.460	20.22	1	0	3	1
## Duster 360	14.3	8	360.0	245	3.21	3.570	15.84	0	0	3	4
## Merc 240D	24.4	4	146.7	62	3.69	3.190	20.00	1	0	4	2
## Merc 230	22.8	4	140.8	95	3.92	3.150	22.90	1	0	4	2
## Merc 280	19.2	6	167.6	123	3.92	3.440	18.30	1	0	4	4
## Merc 280C	17.8	6	167.6	123	3.92	3.440	18.90	1	0	4	4
## Merc 450SE	16.4	8	275.8	180	3.07	4.070	17.40	0	0	3	3
## Merc 450SL	17.3	8	275.8	180	3.07	3.730	17.60	0	0	3	3
## Merc 450SLC	15.2	8	275.8	180	3.07	3.780	18.00	0	0	3	3
## Cadillac Fleetwood	10.4	8	472.0	205	2.93	5.250	17.98	0	0	3	4
## Lincoln Continental	10.4	8	460.0	215	3.00	5.424	17.82	0	0	3	4
## Chrysler Imperial	14.7	8	440.0	230	3.23	5.345	17.42	0	0	3	4
## Fiat 128	32.4	4	78.7	66	4.08	2.200	19.47	1	1	4	1
## Honda Civic	30.4	4	75.7	52	4.93	1.615	18.52	1	1	4	2
## Toyota Corolla	33.9	4	71.1	65	4.22	1.835	19.90	1	1	4	1
## Toyota Corona	21.5	4	120.1	97	3.70	2.465	20.01	1	0	3	1
## Dodge Challenger	15.5	8	318.0	150	2.76	3.520	16.87	0	0	3	2
## AMC Javelin	15.2	8	304.0	150	3.15	3.435	17.30	0	0	3	2
## Camaro Z28	13.3	8	350.0	245	3.73	3.840	15.41	0	0	3	4
## Pontiac Firebird	19.2	8	400.0	175	3.08	3.845	17.05	0	0	3	2
## Fiat X1-9	27.3	4	79.0	66	4.08	1.935	18.90	1	1	4	1
## Porsche 914-2	26.0	4	120.3	91	4.43	2.140	16.70	0	1	5	2
## Lotus Europa	30.4	4	95.1	113	3.77	1.513	16.90	1	1	5	2
## Ford Pantera L	15.8	8	351.0	264	4.22	3.170	14.50	0	1	5	4
## Ferrari Dino	19.7	6	145.0	175	3.62	2.770	15.50	0	1	5	6
## Maserati Bora	15.0	8	301.0	335	3.54	3.570	14.60	0	1	5	8
## Volvo 142E	21.4	4	121.0	109	4.11	2.780	18.60	1	1	4	2
##	transmission_type										
## Mazda RX4										manual	
## Mazda RX4 Wag										manual	
## Datsun 710										manual	
## Hornet 4 Drive										automatic	
## Hornet Sportabout										automatic	
## Valiant										automatic	
## Duster 360										automatic	
## Merc 240D										automatic	
## Merc 230										automatic	
## Merc 280										automatic	
## Merc 280C										automatic	
## Merc 450SE										automatic	
## Merc 450SL										automatic	
## Merc 450SLC										automatic	
## Cadillac Fleetwood										automatic	
## Lincoln Continental										automatic	
## Chrysler Imperial										automatic	
## Fiat 128										manual	
## Honda Civic										manual	
## Toyota Corolla										manual	

## Toyota Corona	automatic
## Dodge Challenger	automatic
## AMC Javelin	automatic
## Camaro Z28	automatic
## Pontiac Firebird	automatic
## Fiat X1-9	manual
## Porsche 914-2	manual
## Lotus Europa	manual
## Ford Pantera L	manual
## Ferrari Dino	manual
## Maserati Bora	manual
## Volvo 142E	manual

))