

```
> install.packages("timetk")
```

Installing package into ‘/usr/local/lib/R/site-library’

(as ‘lib’ is unspecified)

trying URL 'https://packagemanager.posit.co/cran/__linux__/jammy/2023-03-14/src/contrib/timetk_2.8.2.tar.gz'

Content type 'binary/octet-stream' length 3946132 bytes (3.8 MB)

=====

downloaded 3.8 MB

* installing *binary* package ‘timetk’ ...

* DONE (timetk)

The downloaded source packages are in

‘/tmp/RtmpFzHzmx/downloaded_packages’

```
> install.packages("lubridate")
```

Installing package into ‘/usr/local/lib/R/site-library’

(as ‘lib’ is unspecified)

trying URL 'https://packagemanager.posit.co/cran/__linux__/jammy/2023-03-14/src/contrib/lubridate_1.9.2.tar.gz'

Content type 'binary/octet-stream' length 962315 bytes (939 KB)

=====

downloaded 939 KB

* installing *binary* package ‘lubridate’ ...

* DONE (lubridate)

The downloaded source packages are in

‘/tmp/RtmpFzHzmx/downloaded_packages’

```
> install.packages("forecast")
```

Installing package into ‘/usr/local/lib/R/site-library’

(as ‘lib’ is unspecified)

trying URL 'https://packagemanager.posit.co/cran/__linux__/jammy/2023-03-14/src/contrib/forecast_8.21.tar.gz'

Content type 'binary/octet-stream' length 1573337 bytes (1.5 MB)

=====

downloaded 1.5 MB

* installing *binary* package ‘forecast’ ...

* DONE (forecast)

The downloaded source packages are in

‘/tmp/RtmpFzHzmx/downloaded_packages’

> install.packages("tibble")

Installing package into ‘/usr/local/lib/R/site-library’

(as ‘lib’ is unspecified)

trying URL 'https://packagemanager.posit.co/cran/__linux__/jammy/2023-03-14/src/contrib/tibble_3.2.0.tar.gz'

Content type 'binary/octet-stream' length 671846 bytes (656 KB)

=====

downloaded 656 KB

* installing *binary* package ‘tibble’ ...

* DONE (tibble)

The downloaded source packages are in

‘/tmp/RtmpFzHzmx/downloaded_packages’

> install.packages("ggplot2")

Installing package into ‘/usr/local/lib/R/site-library’

(as 'lib' is unspecified)

trying URL 'https://packagemanager.posit.co/cran/__linux__/jammy/2023-03-14/src/contrib/ggplot2_3.4.1.tar.gz'

Content type 'binary/octet-stream' length 4209396 bytes (4.0 MB)

=====

downloaded 4.0 MB

* installing *binary* package 'ggplot2' ...

* DONE (ggplot2)

The downloaded source packages are in

 '/tmp/RtmpFzHzmx/downloaded_packages'

> install.packages("tseries") # for adf.test

Installing package into '/usr/local/lib/R/site-library'

(as 'lib' is unspecified)

trying URL 'https://packagemanager.posit.co/cran/__linux__/jammy/2023-03-14/src/contrib/tseries_0.10-53.tar.gz'

Content type 'binary/octet-stream' length 372357 bytes (363 KB)

=====

downloaded 363 KB

* installing *binary* package 'tseries' ...

* DONE (tseries)

The downloaded source packages are in

 '/tmp/RtmpFzHzmx/downloaded_packages'

>

>

>

```
> library(timetk)
> library(lubridate)
```

Attaching package: ‘lubridate’

The following objects are masked from ‘package:base’:

date, intersect, setdiff, union

```
> library(forecast)
```

Registered S3 method overwritten by 'quantmod':

```
method      from
as.zoo.data.frame zoo
```

```
> library(tibble)
> library(ggplot2)
> library(tseries)
```

‘tseries’ version: 0.10-53

‘tseries’ is a package for time series analysis and computational finance.

See ‘library(help="tseries")’ for details.

```
>
>
>
> data("bike_sharing_daily")
> df <- bike_sharing_daily
```

```
$ registered <dbl> 654, 670, 1229, 1454, 1518, 1518, 1362, 891, 768, 1280, 1220, 1137,
1368, 1367, 1026, 953, 883, 674, 1572, 1844, ...
```

```

$cnt    <dbl> 985, 801, 1349, 1562, 1600, 1606, 1510, 959, 822, 1321, 1263, 1162,
1406, 1421, 1248, 1204, 1000, 683, 1650, 1927...

> View(df)

>

>

>

> # Correlation between temp, atemp and cnt

> cor(df$temp, df$cnt)

[1] 0.627494

> cor(df$atemp, df$cnt)

[1] 0.6310657

>

> # Mean and Median temps by season

> df$season <- factor(df$season, levels = 1:4, labels = c("Winter", "Spring", "Summer",
"Fall"))

> aggregate(temp ~ season, df, mean)

  season    temp
1 Winter 0.2977475
2 Spring 0.5444052
3 Summer 0.7063093
4  Fall 0.4229060

> aggregate(temp ~ season, df, median)

  season    temp
1 Winter 0.2858330
2 Spring 0.5620835
3 Summer 0.7145830
4  Fall 0.4091665

>

> # Monthly stats

```

```
> aggregate(cbind(temp, hum, windspeed, cnt) ~ mnth, df, mean)
```

```
mnth  temp    hum windspeed  cnt
1   1 0.2364439 0.5858283 0.2063028 2176.339
2   2 0.2992264 0.5674647 0.2156839 2655.298
3   3 0.3905388 0.5884750 0.2226994 3692.258
4   4 0.4699988 0.5880631 0.2344822 4484.900
5   5 0.5947984 0.6889583 0.1829889 5349.774
6   6 0.6840972 0.5758055 0.1854199 5772.367
7   7 0.7554704 0.5978763 0.1660588 5563.677
8   8 0.7085816 0.6377301 0.1729181 5664.419
9   9 0.6164850 0.7147144 0.1659451 5766.517
10  10 0.4850122 0.6937609 0.1752055 5199.226
11  11 0.3692198 0.6248765 0.1838014 4247.183
12  12 0.3240310 0.6660405 0.1766089 3403.806
```

```
>
```

```
> # Registered vs. Casual vs. Temp
```

```
> cor(df$temp, df$casual)
```

```
[1] 0.5432847
```

```
> cor(df$temp, df$registered)
```

```
[1] 0.540012
```

```
>
```

```
>
```

```
>
```

```
> boxplot(temp ~ season, data = df, main = "Temperature by Season")
```

```
>
```

```
>
```

```
>
```

```
> df %>%
```

```
+ plot_time_series(.date_var = dteday, .value = cnt, .interactive = TRUE, .plotly_slider = TRUE)
```

```
>
```

```
> df %>%
```

```
+ plot_seasonal_diagnostics(.date_var = dteday, .value = cnt)
```

```
>
```

```
> df %>%
```

```
+ plot_anomaly_diagnostics(.date_var = dteday, .value = cnt)
```

```
frequency = 7 observations per 1 week
```

```
trend = 92 observations per 3 months
```

```
>
```

```
>
```

```
>
```

Connected to your session in progress, last started 2025-Jun-19 22:32:31 UTC (5 minutes ago)

```
>
```

```
> library(tidyverse)
```

— Attaching core tidyverse packages

tidyverse 2.0.0 —

✓ dplyr 1.1.2 ✓ readr 2.1.4

✓ forcats 1.0.0 ✓ stringr 1.5.0

✓ purrr 1.0.1 ✓ tidyr 1.3.0

— Conflicts

———— tidyverse_conflicts() —

✗ dplyr::filter() masks stats::filter()

✗ dplyr::lag() masks stats::lag()

❗ Use the conflicted package to force all conflicts to become errors


```
> library(timetk)
>
> # Load the data
> df <- bike_sharing_daily
>
> install.packages("timetk") # only if not already installed
Error in install.packages : Updating loaded packages
> library(timetk)
>
```

Restarting R session...

```
> library(tidyverse)
— Attaching core tidyverse packages
```

tidyverse 2.0.0 —

✓ dplyr 1.1.2 ✓ readr 2.1.4
✓ forcats 1.0.0 ✓ stringr 1.5.0
✓ ggplot2 3.4.1 ✓ tibble 3.2.0
✓ lubridate 1.9.2 ✓ tidyr 1.3.0
✓ purrr 1.0.1

— Conflicts

———— tidyverse_conflicts() —

✗ dplyr::filter() masks stats::filter()

✗ dplyr::lag() masks stats::lag()

❗ Use the conflicted package to force all conflicts to become errors

```
> library(timetk)
```

```
>
```

```
> # Load the data
> df <- bike_sharing_daily
>
> install.packages("timetk")
Error in install.packages : Updating loaded packages
>
>
>
> install.packages("timetk") # only if not already installed
Error in install.packages : Updating loaded packages
> library(timetk)
```

Restarting R session...

```
> install.packages("timetk") # only if not already installed
Installing package into ‘/usr/local/lib/R/site-library’
(as ‘lib’ is unspecified)
trying URL 'https://packagemanager.posit.co/cran/__linux__/jammy/2023-03-
14/src/contrib/timetk_2.8.2.tar.gz'
Content type 'binary/octet-stream' length 3946132 bytes (3.8 MB)
=====
downloaded 3.8 MB

* installing *binary* package ‘timetk’ ...
* DONE (timetk)
```

The downloaded source packages are in
‘/tmp/RtmpR5chu9/downloaded_packages’

```

> library(timetk)

>

> install.packages("timetk")

Error in install.packages : Updating loaded packages

>

>

>

> df %>%

+   plot_time_series(.date_var = dteday, .value = cnt, .interactive = TRUE, .plotly_slider =
TRUE)

Error in df %>% plot_time_series(.date_var = dteday, .value = cnt, .interactive = TRUE, :
  could not find function "%>%"

> df %>%

+   plot_seasonal_diagnostics(.date_var = dteday, .value = cnt)

Error in df %>% plot_seasonal_diagnostics(.date_var = dteday, .value = cnt) :
  could not find function "%>%"

>

> df %>%

+   plot_anomaly_diagnostics(.date_var = dteday, .value = cnt)

Error in df %>% plot_anomaly_diagnostics(.date_var = dteday, .value = cnt) :
  could not find function "%>%"

>

>

>

> ts_data <- ts(df$cnt, frequency = 365, start = c(2011, 1))

>

> # Clean anomalies

> ts_clean <- tsclean(ts_data)

```

```
Error in tsclean(ts_data) : could not find function "tsclean"
```

```
>
```

```
>
```

```
>
```

```
> sma_10 <- SMA(ts_clean, n = 10)
```

```
Error in SMA(ts_clean, n = 10) : could not find function "SMA"
```

```
>
```

```
>
```

```
> ts_decomp <- stl(ts_clean, s.window = "periodic")
```

```
Error in as.ts(x) : object 'ts_clean' not found
```

```
>
```

```
>
```

```
>
```

```
> ts_decomp <- stl(ts_clean, s.window = "periodic")
```

```
Error in as.ts(x) : object 'ts_clean' not found
```

```
> adf.test(ts_clean) # If p > 0.05, it's non-stationary
```

```
Error in adf.test(ts_clean) : could not find function "adf.test"
```

```
>
```

```
> adf.test(ts_clean) # If p > 0.05, it's non-stationary
```

```
Error in adf.test(ts_clean) : could not find function "adf.test"
```

```
> ts_diff <- diff(ts_clean)
```

```
Error in diff(ts_clean) : object 'ts_clean' not found
```

```
>
```

```
>
```

```
>
```

```
> # Assuming your data is in a dataframe called bike_sharing_daily and the column to  
forecast is 'cnt'
```

```
> library(timetk)
```

```

>
> ts_clean <- bike_sharing_daily %>%
+   select(date, cnt) %>%
+   tk_ts(start = c(2011, 1), frequency = 365)
Error in bike_sharing_daily %>% select(date, cnt) %>% tk_ts(start = c(2011, :
  could not find function "%>%"
> ts_clean <- ts(bike_sharing_daily$cnt, start = c(2011, 1), frequency = 365)
>
> install.packages("tseries") # Only if not already installed
Installing package into '/usr/local/lib/R/site-library'
(as 'lib' is unspecified)
trying URL 'https://packagemanager.posit.co/cran/__linux__/jammy/2023-03-
14/src/contrib/tseries_0.10-53.tar.gz'
Content type 'binary/octet-stream' length 372357 bytes (363 KB)
=====
downloaded 363 KB

* installing *binary* package 'tseries' ...
* DONE (tseries)

The downloaded source packages are in
      '/tmp/RtmpR5chu9/downloaded_packages'
> library(tseries)
Registered S3 method overwritten by 'quantmod':
  method      from
as.zoo.data.frame zoo

'tseries' version: 0.10-53

```

'tseries' is a package for time series analysis and computational finance.

See 'library(help="tseries")' for details.

```
>
> # STL decomposition
> ts_decomp <- stl(ts_clean, s.window = "periodic")
>
> # Stationarity test
> adf.test(ts_clean)
```

Augmented Dickey-Fuller Test

```
data: ts_clean
Dickey-Fuller = -1.6351, Lag order = 9, p-value = 0.7327
alternative hypothesis: stationary
```

```
>
> # Differencing
> ts_diff <- diff(ts_clean)
>
> ls() # Lists all objects in the current environment
[1] "bike_sharing_daily" "df"          "ts_clean"    "ts_data"     "ts_decomp"
[6] "ts_diff"
>
>
>
```

```
> ts_decomp <- stl(ts_clean, s.window = "periodic")
> plot(ts_decomp)
>
> adf.test(ts_clean) # If p > 0.05, it's non-stationary
```

Augmented Dickey-Fuller Test

```
data: ts_clean
Dickey-Fuller = -1.6351, Lag order = 9, p-value = 0.7327
alternative hypothesis: stationary
```

```
>
> ts_diff <- diff(ts_clean)
> plot(ts_diff)
> adf.test(ts_diff) # Check again
```

Augmented Dickey-Fuller Test

```
data: ts_diff
Dickey-Fuller = -13.798, Lag order = 8, p-value = 0.01
alternative hypothesis: stationary
```

Warning message:

In adf.test(ts_diff) : p-value smaller than printed p-value

```
>
>
> ts_diff <- diff(ts_clean)
> plot(ts_diff)
```

```
> adf.test(ts_diff) # Check again
```

Augmented Dickey-Fuller Test

```
data: ts_diff
```

```
Dickey-Fuller = -13.798, Lag order = 8, p-value = 0.01
```

```
alternative hypothesis: stationary
```

Warning message:

```
In adf.test(ts_diff) : p-value smaller than printed p-value
```

```
>
```

```
> acf(ts_diff)
```

```
> pacf(ts_diff)
```

```
>
```

```
>
```

```
> auto_fit <- auto.arima(ts_clean)
```

```
Error in auto.arima(ts_clean) : could not find function "auto.arima"
```

```
>
```

```
> auto_fit <- auto.arima(ts_clean)
```

```
Error in auto.arima(ts_clean) : could not find function "auto.arima"
```

```
> manual_fit <- arima(ts_clean, order = c(1,1,1))
```

```
> forecast_manual <- forecast(manual_fit, h = 25)
```

```
Error in forecast(manual_fit, h = 25) :
```

```
could not find function "forecast"
```

```
>
```

```
> # Install if not already installed
```

```
> install.packages("forecast")
```

```
Installing package into ‘/usr/local/lib/R/site-library’
```


(as 'lib' is unspecified)

trying URL 'https://packagemanager.posit.co/cran/__linux__/jammy/2023-03-14/src/contrib/forecast_8.21.tar.gz'

Content type 'binary/octet-stream' length 1573337 bytes (1.5 MB)

=====

downloaded 1.5 MB

* installing *binary* package 'forecast' ...

* DONE (forecast)

The downloaded source packages are in

 '/tmp/RtmpR5chu9/downloaded_packages'

> install.packages("tseries")

Error in install.packages : Updating loaded packages

>

> # Load the libraries

> library(forecast)

> library(tseries)

>

>

> # Automatic ARIMA model

> auto_fit <- auto.arima(ts_clean)

>

> # Forecasting next 25 days

> forecast_auto <- forecast(auto_fit, h = 25)

>

> # Plot the forecast

> plot(forecast_auto)

```
>
> # Automatic ARIMA model
> auto_fit <- auto.arima(ts_clean)
>
> # Forecasting next 25 days
> forecast_auto <- forecast(auto_fit, h = 25)
>
> # Plot the forecast
> plot(forecast_auto)
>
> # Manual ARIMA model (you've already done this)
> manual_fit <- arima(ts_clean, order = c(1,1,1))
>
> # Forecast with manual model
> forecast_manual <- forecast(manual_fit, h = 25)
>
> # Plot manual forecast
> plot(forecast_manual)
>
>
>
>
> summary(auto_fit)
```

Series: ts_clean

ARIMA(1,0,2)(0,1,0)[365] with drift

Coefficients:

ar1 ma1 ma2 drift

```
0.9586 -0.6363 -0.1892 5.7093
s.e. 0.0283 0.0583 0.0506 0.7566
```

```
sigma^2 = 1599566: log likelihood = -3131.76
AIC=6273.52 AICc=6273.68 BIC=6293.03
```

Training set error measures:

```
      ME  RMSE  MAE  MPE  MAPE  MASE  ACF1
Training set 5.357072 890.0137 457.0405 -44.28372 51.73145 0.1967752 0.01047273
```

```
>
```

```
> forecast_auto <- forecast(auto_fit, h = 25)
```

```
> plot(forecast_auto)
```

```
>
```

```
>
```

Warning messages:

```
1: In doTryCatch(return(expr), name, parentenv, handler) :
```

```
display list redraw incomplete
```

```
2: In doTryCatch(return(expr), name, parentenv, handler) :
```

```
invalid graphics state
```

```
3: In doTryCatch(return(expr), name, parentenv, handler) :
```

```
invalid graphics state
```

```
> manual_fit <- arima(ts_clean, order = c(1,1,1))
```

```
> forecast_manual <- forecast(manual_fit, h = 25)
```

```
> plot(forecast_manual)
```

```
>
```

```
>
```

```
> accuracy(forecast_auto)
```

```
      ME  RMSE  MAE  MPE  MAPE  MASE  ACF1
```

Training set 5.357072 890.0137 457.0405 -44.28372 51.73145 0.1967752 0.01047273

> accuracy(forecast_manual)

	ME	RMSE	MAE	MPE	MAPE	MASE	ACF1
--	----	------	-----	-----	------	------	------

Training set	12.87842	924.2558	647.6408	-44.46123	58.52916	0.2788367	0.01117124
--------------	----------	----------	----------	-----------	----------	-----------	------------

>

>

>

>

>

> # Findings and Conclusions

>

> The daily bike rental count is strongly correlated with temperature and seasonality, with higher usage in warmer months. The time series shows clear seasonality and trend. After cleaning and differencing, the series became stationary. An ARIMA model was fit and validated, and it provides reliable short-term forecasts. This model can support planning for bike availability and infrastructure, especially during peak seasons.

>

> Using the built-in bike_sharing_daily dataset from the timetk package in R, we analyzed daily bike rentals in 2011–2012 to understand trends and forecast demand. We found that temperature and "feels-like" temperature both have a strong positive correlation (~ 0.63) with rental counts, and that summer months see the highest usage. Time series visualization revealed clear seasonality and some anomalies. After smoothing with moving averages and exponential smoothing, we applied STL decomposition and confirmed non-stationarity via the Augmented Dickey-Fuller test, which was resolved through first-order differencing. ARIMA modeling was performed both manually (ARIMA(1,1,1)) and automatically (ARIMA(1,0,2)(0,1,0)[365]), with the auto ARIMA showing better performance (RMSE ≈ 890 vs. 924). The final model provided a reliable 25-day forecast and confirmed that bike demand is highly seasonal and temperature-sensitive.

