# Package ‘AirMonitor’

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**Type** Package  

**Version** 0.4.0  

**Title** Air Quality Data Analysis  

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**Description** Utilities for working with hourly air quality monitoring data with a focus on small particulates (PM2.5). A compact data model is structured as a list with two dataframes. A 'meta' dataframe contains spatial and measuring device metadata associated with deployments at known locations. A 'data' dataframe contains a 'datetime' column followed by columns of measurements associated with each `device-deployment`. Algorithms to calculate NowCast and the associated Air Quality Index (AQI) are defined at the US Environmental Projection Agency AirNow program: &lt;https://www.airnow.gov/sites/default/files/2020-05/aqi-technical-assistance-document-sept2018.pdf&gt;.

**License** GPL-3  

**URL** https://github.com/MazamaScience/AirMonitor,  

https://mazamascience.github.io/AirMonitor/  

**BugReports** https://github.com/MazamaScience/AirMonitor/issues  

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```
addAQILegend

Description

This function is a convenience wrapper around graphics::legend(). It will show the AQI colors and names by default if col and legend are not specified.

AQI categories are arranged with lower levels at the bottom of the legend to match the arrangement in the plot. This is different from the default "reading order" so you may wish to reverse the order of user supplied arguments with rev().

Add an AQI legend to a map
addAQILines

Usage

addAQILegend(
  x = "topright",
  y = NULL,
  pollutant = c("PM2.5", "CO", "OZONE", "PM10", "AQI"),
  palette = c("EPA", "subdued", "deuteranopia"),
  languageCode = c("eng", "spa"),
  NAAQS = c("PM2.5", "PM2.5_2024"),
  ...)

Arguments

  x   x Coordinate passed on to the legend() command.
  y   y Coordinate passed on to the legend() command.
  pollutant   EPA AQS criteria pollutant.
  palette   Named color palette to use for AQI categories.
  NAAQS   Version of NAAQS levels to use. See Note.
  ...   Additional arguments to be passed to legend().

Value

A list with components rect and text is returned invisibly. (See legend.)

Note

On February 7, 2024, EPA strengthened the National Ambient Air Quality Standards for Particulate Matter (PM NAAQS) to protect millions of Americans from harmful and costly health impacts, such as heart attacks and premature death. Particle or soot pollution is one of the most dangerous forms of air pollution, and an extensive body of science links it to a range of serious and sometimes deadly illnesses. EPA is setting the level of the primary (health-based) annual PM2.5 standard at 9.0 micrograms per cubic meter to provide increased public health protection, consistent with the available health science. See PM NAAQS update.

addAQILines

Add AQI lines to a plot

Description

Draws AQI lines across a plot at the levels appropriate for the monitor_timeseriesPlot function uses this function internally when specifying addAQI = TRUE. pollutant.
addAQIStackedBar

Usage

addAQIStackedBar(
  pollutant = c("PM2.5", "CO", "OZONE", "PM10", "AQI"),
  palette = c("EPA", "subdued", "deuteranopia"),
  NAAQS = c("PM2.5", "PM2.5_2024"),
  ...)

Arguments

pollutant  EPA AQS criteria pollutant.
palette    Named color palette to use for AQI categories.
NAAQS      Version of NAAQS levels to use. See Note.
...         additional arguments to be passed to abline()

Value

No return value, called to add lines to a time series plot.

Note

On February 7, 2024, EPA strengthened the National Ambient Air Quality Standards for Particulate Matter (PM NAAQS) to protect millions of Americans from harmful and costly health impacts, such as heart attacks and premature death. Particle or soot pollution is one of the most dangerous forms of air pollution, and an extensive body of science links it to a range of serious and sometimes deadly illnesses. EPA is setting the level of the primary (health-based) annual PM2.5 standard at 9.0 micrograms per cubic meter to provide increased public health protection, consistent with the available health science. See PM NAAQS update.
Arguments

- pollutant: EPA AQS criteria pollutant.
- palette: Named color palette to use for AQI categories.
- width: Width of the bar as a fraction of the width of the plot area.
- height: Height of the bar as a fraction of the height of the plot area.
- pos: Position of the stacked bar relative to the plot.
- NAAQS: Version of NAAQS levels to use. See Note.

Value

No return value, called to add color bars to a time series plot.

Note

On February 7, 2024, EPA strengthened the National Ambient Air Quality Standards for Particulate Matter (PM NAAQS) to protect millions of Americans from harmful and costly health impacts, such as heart attacks and premature death. Particle or soot pollution is one of the most dangerous forms of air pollution, and an extensive body of science links it to a range of serious and sometimes deadly illnesses. EPA is setting the level of the primary (health-based) annual PM2.5 standard at 9.0 micrograms per cubic meter to provide increased public health protection, consistent with the available health science. See PM NAAQS update.

addShadedNight

Add nighttime shading to a timeseries plot

Description

Draw shading rectangles on a plot to indicate nighttime hours. The `monitor_timeseriesPlot` function uses this function internally when specifying `shadedNight = TRUE`.

Usage

```r
addShadedNight(timeInfo, col = adjustcolor("black", 0.1))
```

Arguments

- timeInfo: dataframe as returned by `MazamaTimeSeries::monitor_timeInfo()`
- col: Color used to shade nights.

Value

No return value, called to add day/night shading to a timeseries plot.
**AirFire_S3_archiveBaseUrl**

*USFS maintained archive base URL*

---

**Description**

The US Forest Service AirFire group maintains an archive of processed monitoring data. The base URL for this archive is used as the default in all `~load()` functions.


**Usage**

AirFire_S3_archiveBaseUrl

**Format**

A url

**Details**

AirFire_S3_archiveBaseUrl

---

**AirMonitor**

*Air Quality Data Analysis*

---

**Description**

Utilities for working with hourly air quality monitoring data with a focus on small particulates (PM2.5). A compact data model is structured as a list with two dataframes. A 'meta' dataframe contains spatial and measuring device metadata associated with deployments at known locations. A 'data' dataframe contains a 'datetime' column followed by columns of measurements associated with each "device-deployment".
Load annual AirNow monitoring data

Description

Loads pre-generated .rda files containing hourly AirNow data.

If `archiveDataDir` is defined, data will be loaded from this local archive. Otherwise, data will be loaded from the monitoring data repository maintained by the USFS AirFire team.

The files loaded by this function contain a single year's worth of data

For the most recent data in the last 10 days, use `airnow_loadLatest()`.

For daily updates covering the most recent 45 days, use `airnow_loadDaily()`.

For archival data for a specific month, use `airnow_loadMonthly()`.

Pre-processed AirNow exists for the following parameters:

1. PM2.5

Usage

```r
airnow_loadAnnual(
  year = NULL,
  archiveBaseUrl = paste0("https://airfire-data-exports.s3.us-west-2.amazonaws.com/",
                          "monitoring/v2"),
  archiveBaseDir = NULL,
  QC_negativeValues = c("zero", "na", "ignore"),
  parameterName = "PM2.5"
)
```

Arguments

- `year`: Year [YYYY].
- `archiveBaseUrl`: Base URL for monitoring v2 data files.
- `archiveBaseDir`: Local base directory for monitoring v2 data files.
- `QC_negativeValues`: Type of QC to apply to negative values.
- `parameterName`: One of the EPA AQS criteria parameter names.

Value

A `mts_monitor` object with AirNow data. (A list with meta and data dataframes.)

See Also

- `airnow_loadDaily`
- `airnow_loadLatest`
- `airnow_loadMonthly`
Examples

```r
## Not run:
library(AirMonitor)

# Fail gracefully if any resources are not available
try(

# See https://en.wikipedia.org/wiki/2017_Montana_wildfires

# Daily Barplot of Montana wildfires
airnow_loadAnnual(2017) \ 
  monitor_filter(stateCode == "MT") \ 
  monitor_filterDate(20170701, 20170930, timezone = "America/Denver") \ 
  monitor_dailyStatistic() \ 
  monitor_timeseriesPlot( 
    ylim = c(0, 300), 
    xpd = NA, 
    addAQI = TRUE, 
    main = "Montana 2017 -- AirNow Daily Average PM2.5"
  )
), silent = FALSE)

## End(Not run)
```

---

**amrnow_loadDaily**

 load daily AirNow monitoring data

---

**Description**

 Loads pre-generated .rda files containing hourly AirNow data.

 If `archiveDataDir` is defined, data will be loaded from this local archive. Otherwise, data will be loaded from the monitoring data repository maintained by the USFS AirFire team.

 The files loaded by this function are updated once per day and contain data for the previous 45 days.

 For the most recent data in the last 10 days, use `airnow_loadLatest()`.

 For data extended more than 45 days into the past, use `airnow_loadAnnual()`.

 Pre-processed AirNow exists for the following parameters:

 1. PM2.5
 2. PM2.5_nowcast

**Usage**

```r
airnow_loadDaily(
  archiveBaseUrl = paste0("https://airfire-data-exports.s3.us-west-2.amazonaws.com/", 
         "monitoring/v2"),
  archiveBaseDir = NULL,
```
QC_negativeValues = c("zero", "na", "ignore"),
  parameterName = "PM2.5"
)

Arguments

archiveBaseUrl Base URL for monitoring v2 data files.
archiveBaseDir Local base directory for monitoring v2 data files.
QC_negativeValues Type of QC to apply to negative values.
parameterName One of the EPA AQS criteria parameter names.

Value

A mts_monitor object with AirNow data. (A list with meta and data dataframes.)

See Also

airnow_loadAnnual
airnow_loadLatest
airnow_loadMonthly

Examples

## Not run:
library(AirMonitor)

# Fail gracefully if any resources are not available
try{{

  airnow_loadDaily() \
  monitor_filter(stateCode == "WA") \
  monitor_leaflet()
}
}, silent = FALSE)

## End(Not run)
**Description**

Loads pre-generated .rda files containing the most recent AirNow data.

If `archiveDataDir` is defined, data will be loaded from this local archive. Otherwise, data will be loaded from the monitoring data repository maintained by the USFS AirFire team.

The files loaded by this function are updated multiple times an hour and contain data for the previous 10 days.

For daily updates covering the most recent 45 days, use `airnow_loadDaily()`.

For data extended more than 45 days into the past, use `airnow_loadAnnual()`.

Pre-processed AirNow exists for the following parameters:

1. PM2.5
2. PM2.5_nowcast

**Usage**

```r
airnow_loadLatest(
  archiveBaseUrl = paste0("https://airfire-data-exports.s3.us-west-2.amazonaws.com/", "monitoring/v2"),
  archiveBaseDir = NULL,
  QC_negativeValues = c("zero", "na", "ignore"),
  parameterName = "PM2.5"
)
```

**Arguments**

- `archiveBaseUrl`: Base URL for monitoring v2 data files.
- `archiveBaseDir`: Local base directory for monitoring v2 data files.
- `QC_negativeValues`: Type of QC to apply to negative values.
- `parameterName`: One of the EPA AQS criteria parameter names.

**Value**

A `mts_monitor` object with AirNow data. (A list with `meta` and `data` dataframes.)

**See Also**

- `airnow_loadAnnual`
- `airnow_loadDaily`
- `airnow_loadMonthly`
Examples

```r
## Not run:
library(AirMonitor)

# Fail gracefully if any resources are not available
try({
  airnow_loadLatest() \n  monitor_filter(stateCode == "WA") \n  monitor_leaflet()
}, silent = FALSE)
## End(Not run)
```

##### airnow_loadMonthly

**Load monthly AirNow monitoring data**

**Description**

Loads pre-generated .rda files containing hourly AirNow data.

If `archiveDataDir` is defined, data will be loaded from this local archive. Otherwise, data will be loaded from the monitoring data repository maintained by the USFS AirFire team.

The files loaded by this function contain a single month’s worth of data.

For the most recent data in the last 10 days, use `airnow_loadLatest()`.

For daily updates covering the most recent 45 days, use `airnow_loadDaily()`.

For data extended more than 45 days into the past, use `airnow_loadAnnual()`.

Pre-processed AirNow exists for the following parameters:

1. PM2.5

**Usage**

```r
airnow_loadMonthly(
  monthStamp = NULL,
  archiveBaseUrl = paste0("https://airfire-data-exports.s3.us-west-2.amazonaws.com/",
                          "monitoring/v2"),
  archiveBaseDir = NULL,
  QC_negativeValues = c("zero", "na", "ignore"),
  parameterName = "PM2.5"
)
```
**airsis_loadAnnual**

*Load annual AIRSIS monitoring data*

### Arguments

- **monthStamp**
  - Year-month [YYYYmm].
- **archiveBaseUrl**
  - Base URL for monitoring v2 data files.
- **archiveBaseDir**
  - Local base directory for monitoring v2 data files.
- **QC_negativeValues**
  - Type of QC to apply to negative values.
- **parameterName**
  - One of the EPA AQS criteria parameter names.

### Value

A *mts_monitor* object with AirNow data. (A list with meta and data dataframes.)

### Description

Loads pre-generated .rda files containing annual AIRSIS data.

If `archiveDataDir` is defined, data will be loaded from this local archive. Otherwise, data will be loaded from the monitoring data repository maintained by the USFS AirFire team.

Current year files loaded by this function are updated once per week.

For the most recent data in the last 10 days, use `airsis_loadLatest()`.

For daily updates covering the most recent 45 days, use `airsis_loadDaily()`.

### Usage

```r
airsis_loadAnnual(
  year = NULL,
  archiveBaseUrl = paste0("https://airfire-data-exports.s3.us-west-2.amazonaws.com/", 
                           "monitoring/v2"),
  archiveBaseDir = NULL,
  QC_negativeValues = c("zero", "na", "ignore"),
  QC_removeSuspectData = TRUE
)
```

### Arguments

- **year**
  - Year [YYYY].
- **archiveBaseUrl**
  - Base URL for monitoring v2 data files.
- **archiveBaseDir**
  - Local base directory for monitoring v2 data files.
- **QC_negativeValues**
  - Type of QC to apply to negative values.
- **QC_removeSuspectData**
  - Removes monitors determined to be misbehaving.
Value

A mts_monitor object with AIRSIS data. (A list with meta and data dataframes.)

Note

Some older AIRSIS timeseries contain only values of 0, 1000, 2000, 3000, ... ug/m3. Data from these deployments pass instrument-level QC checks but these timeseries generally do not represent valid data and should be removed. With QC_removeSuspectData = TRUE (the default), data is checked and periods reporting only values of 0:10 * 1000 ug/m3 are invalidated. Only those personally familiar with the individual instrument deployments should work with the "suspect" data.

See Also

airsis_loadDaily
airsis_loadLatest

Examples

## Not run:
library(AirMonitor)

# Fail gracefully if any resources are not available
try({

# See https://en.wikipedia.org/wiki/Camp_Fire_(2018)

# AIRSIS monitors during the Camp Fire
airsis_loadAnnual(2018) \ 
  monitor_filter(stateCode == "CA") \ 
  monitor_filterDate(20181101, 20181201) \ 
  monitor_dropEmpty() \ 
  monitor_leaflet()
}, silent = FALSE)

## End(Not run)
For the most recent data in the last 10 days, use `airsis_loadLatest()`.
For data extended more than 45 days into the past, use `airsis_loadAnnual()`.

Usage

```r
airsis_loadDaily(
  archiveBaseUrl = paste0("https://airfire-data-exports.s3.us-west-2.amazonaws.com/", "monitoring/v2"),
  archiveBaseDir = NULL,
  QC_negativeValues = c("zero", "na", "ignore"),
  QC_removeSuspectData = TRUE
)
```

Arguments

- `archiveBaseUrl` Base URL for monitoring v2 data files.
- `archiveBaseDir` Local base directory for monitoring v2 data files.
- `QC_negativeValues` Type of QC to apply to negative values.
- `QC_removeSuspectData` Removes monitors determined to be misbehaving.

Value

A `mts_monitor` object with AIRSIS data. (A list with `meta` and `data` dataframes.)

Note

Some older AIRSIS timeseries contain only values of 0, 1000, 2000, 3000, ... ug/m3. Data from these deployments pass instrument-level QC checks but these timeseries generally do not represent valid data and should be removed. With `QC_removeSuspectData = TRUE` (the default), data is checked and periods reporting only values of 0:10 * 1000 ug/m3 are invalidated.

Only those personally familiar with the individual instrument deployments should work with the "suspect" data.

See Also

- `airsis_loadAnnual`
- `airsis_loadLatest`

Examples

```r
## Not run:
library(AirMonitor)

# Fail gracefully if any resources are not available
try({
  airsis_loadDaily()
})
```
airsis_loadLatest

Load most recent AIRSIS monitoring data

Description

Loads pre-generated .rda files containing the most recent AIRSIS data.

If `archiveDataDir` is defined, data will be loaded from this local archive. Otherwise, data will be loaded from the monitoring data repository maintained by the USFS AirFire team.

The files loaded by this function are updated multiple times an hour and contain data for the previous 10 days.

For daily updates covering the most recent 45 days, use `airsis_loadDaily()`.

For data extended more than 45 days into the past, use `airsis_loadAnnual()`.

Usage

`airsis_loadLatest(
  archiveBaseUrl = paste0("https://airfire-data-exports.s3.us-west-2.amazonaws.com/",
                         "monitoring/v2"),
  archiveBaseDir = NULL,
  QC_negativeValues = c("zero", "na", "ignore"),
  QC_removeSuspectData = TRUE
)
`

Arguments

- `archiveBaseUrl` Base URL for monitoring v2 data files.
- `archiveBaseDir` Local base directory for monitoring v2 data files.
- `QC_negativeValues` Type of QC to apply to negative values.
- `QC_removeSuspectData` Removes monitors determined to be misbehaving.

Value

A `mts_monitor` object with AIRSIS data. (A list with `meta` and `data` dataframes.)
Note
Some older AIRSIS timeseries contain only values of 0, 1000, 2000, 3000, ... ug/m3. Data from these deployments pass instrument-level QC checks but these timeseries generally do not represent valid data and should be removed. With QC_removeSuspectData = TRUE (the default), data is checked and periods reporting only values of 0:10 * 1000 ug/m3 are invalidated.
Only those personally familiar with the individual instrument deployments should work with the "suspect" data.

See Also
airsis_loadAnnual
airsis_loadDaily

Examples
## Not run:
library(AirMonitor)
# Fail gracefully if any resources are not available
try({
  aqis_loadLatest()
  monitor_filter(stateCode == "CA")
  monitor_leaflet()
}, silent = FALSE)
## End(Not run)

aqiCategories

Generate AQI categories

Description
This function converts hourly PM2.5 measurements into AQI category levels. These levels can then be converted to colors or names using the arrays found in US_AQI.

Usage
aqiCategories(
  x,
  pollutant = c("PM2.5", "AQI", "CO", "NO", "OZONE", "PM10", "SO2"),
  NAAQS = c("PM2.5", "PM2.5_2024"),
  conversionArray = NULL
)
Arguments

- **x**: Vector or matrix of PM2.5 values or an `mts_monitor` object.
- **pollutant**: EPA AQS criteria pollutant.
- **NAAQS**: Version of NAAQS levels to use. See Note.
- **conversionArray**: Array of six text or other values to return instead of integers.

Details

By default, return values will be integers in the range 1:6 or NA. The `conversionArray` parameter can be used to convert these integers into whatever is specified in the first six elements of `conversionArray`. A typical usage would be:  
```
conversionArray = US_AQI$names_eng.
```

Value

A vector or matrix of AQI category indices in the range 1:6.

Note

On February 7, 2024, EPA strengthened the National Ambient Air Quality Standards for Particulate Matter (PM NAAQS) to protect millions of Americans from harmful and costly health impacts, such as heart attacks and premature death. Particle or soot pollution is one of the most dangerous forms of air pollution, and an extensive body of science links it to a range of serious and sometimes deadly illnesses. EPA is setting the level of the primary (health-based) annual PM2.5 standard at 9.0 micrograms per cubic meter to provide increased public health protection, consistent with the available health science. See PM NAAQS update.

See Also

- `aqiColors`

Examples

```r
library(AirMonitor)

# Lane County, Oregon AQSIDs all begin with "41039"
LaneCounty <- NW_Megafires %>%
  monitor_filter(stringr::str_detect(AQSID, "^41039")) %>%
  monitor_filterDate(20150822, 20150823)

LaneCounty %>%
aqiCategories()

LaneCounty %>%
aqiCategories(conversionArray = US_AQI$names_eng)
```
### aqiColors

**Generate AQI colors**

**Description**

This function uses the `leaflet::colorBin()` function to return a vector or matrix of colors derived from data values.

**Usage**

```r
aqiColors(
  x,
  pollutant = c("PM2.5", "AQI", "CO", "NO", "OZONE", "PM10", "SO2"),
  palette = c("EPA", "subdued", "deuteranopia"),
  na.color = NA,
  NAAQS = c("PM2.5", "PM2.5_2024")
)
```

**Arguments**

- `x` Vector or matrix of PM2.5 values or an `mts_monitor` object.
- `pollutant` EPA AQS criteria pollutant.
- `palette` Named color palette to use for AQI categories.
- `na.color` Color assigned to missing values.
- `NAAQS` Version of NAAQS levels to use. See Note.

**Value**

A vector or matrix of AQI colors to be used in maps and plots.

**Note**

On February 7, 2024, EPA strengthened the National Ambient Air Quality Standards for Particulate Matter (PM NAAQS) to protect millions of Americans from harmful and costly health impacts, such as heart attacks and premature death. Particle or soot pollution is one of the most dangerous forms of air pollution, and an extensive body of science links it to a range of serious and sometimes deadly illnesses. EPA is setting the level of the primary (health-based) annual PM2.5 standard at 9.0 micrograms per cubic meter to provide increased public health protection, consistent with the available health science. See PM NAAQS update.

**See Also**

`aqiCategories`
Examples

```r
library(AirMonitor)

# Fancy plot based on pm2.5 values
pm2.5 <- Carmel_Valley$data[,2]
Carmel_Valley %>%
  monitor_timeseriesPlot(
    shadedNight = TRUE,
    pch = 16,
    cex = pmax(pm2.5 / 100, 0.5),
    col = aqiColors(pm2.5),
    opacity = 0.8
  )
```

Camp_Fire

Camp Fire example dataset

Description

The Camp_Fire dataset provides a quickly loadable version of a `mts_monitor` object for practicing and code examples.

Usage

Camp_Fire

Format

A `mts_monitor` object with 360 rows and 134 columns of data.

Details

The 2018 Camp Fire was the deadliest and most destructive wildfire in California's history, and the most expensive natural disaster in the world in 2018 in terms of insured losses. The fire caused at least 85 civilian fatalities and injured 12 civilians and five firefighters. It covered an area of 153,336 acres and destroyed more than 18,000 structures, most with the first 4 hours. Smoke from the fire resulted in the worst air pollution ever for the San Francisco Bay Area and Sacramento Valley.

This dataset was generated on 2022-10-12 by running:

```r
library(AirMonitor)

Camp_Fire <-
  monitor_loadAnnual(2018) %>%
  monitor_filter(stateCode == 'CA') %>%
  monitor_filterDate(
   startdate = 20181108,
   enddate = 20181123,
  )
```
Carmel_Valley

```r
timezone = "America/Los_Angeles"
) %>%
monitor_dropEmpty()

save(Camp_Fire, file = "data/Camp_Fire.rda")
```

---

Carmel_Valley  
*Carmel Valley example dataset*

### Description

The Carmel_Valley dataset provides a quickly loadable version of a `mts_monitor` object for practicing and code examples.

### Usage

```r
Carmel_Valley
```

### Format

A `mts_monitor` object with 576 rows and 2 columns of data.

### Details

In August of 2016, the Soberanes fire in California burned along the Big Sur coast. At the time, it was the most expensive wildfire in US history. This dataset contains PM2.5 monitoring data for the monitor in Carmel Valley which shows heavy smoke as well as strong diurnal cycles associated with sea breezes. Data are stored as a `mts_monitor` object and are used in some examples in the package documentation.

This dataset was generated on 2022-10-12 by running:

```r
library(AirMonitor)

Carmel_Valley <-
  airnow_loadAnnual(2016) %>%
  monitor_filterMeta(deviceDeploymentID == "a9572a904a4ed46d_840060530002") %>%
  monitor_filterDate(20160722, 20160815)

save(Carmel_Valley, file = "data/Carmel_Valley.rda")
```
CONUS

CONUS state codes

Description

State codes for the 48 contiguous states +DC that make up the CONtinental US.


Usage

CONUS

Format

A vector with 49 elements

Details

CONUS state codes

coreMetadataNames

Names of standard metadata columns

Description

Vector of names of the required monitor$meta columns. These represent metadata columns that must exist in every valid mts_monitor object. Any number of additional columns may also be present.

Usage

coreMetadataNames

Format

A vector of character strings

Details

coreMetadataNames

Examples

print(coreMetadataNames, width = 80)
epa_aqs_loadAnnual  

Load annual AirNow monitoring data

Description

Loads pre-generated .rda files containing hourly AirNow data.

If archiveDataDir is defined, data will be loaded from this local archive. Otherwise, data will be loaded from the monitoring data repository maintained by the USFS AirFire team.

The files loaded by this function contain a single year’s worth of data.

Pre-processed AirNow exists for the following parameter codes:

1. 88101 – PM2.5 FRM/FEM Mass
2. 88502 – PM2.5 non FRM/FEM Mass

Specifying parameterCode = "PM2.5" will merge records from both sources.

Usage

epa_aqs_loadAnnual(
  year = NULL,
  archiveBaseUrl = paste0("https://airfire-data-exports.s3.us-west-2.amazonaws.com/",
                         "monitoring/v2"),
  archiveBaseDir = NULL,
  QC_negativeValues = c("zero", "na", "ignore"),
  parameterCode = c("PM2.5", "88101", "88502")
)

Arguments

year  

Year [YYYY].

archiveBaseUrl  

Base URL for monitoring v2 data files.

archiveBaseDir  

Local base directory for monitoring v2 data files.

QC_negativeValues  

Type of QC to apply to negative values.

parameterCode  

One of the EPA AQS criteria parameter codes.

Value

A mts_monitor object with EPA AQS data. (A list with meta and data dataframes.)
Examples

```r
## Not run:
library(AirMonitor)

# Fail gracefully if any resources are not available
try({

# See https://en.wikipedia.org/wiki/2017_Montana_wildfires

# Daily Barplot of Montana wildfires
epa_aqs_loadAnnual(2015) \n  monitor_filter(stateCode == "WA") \n  monitor_filterDate(20150724, 20150907) \n  monitor_dailyStatistic() \n  monitor_timeseriesPlot(
    main = "Washington 2015 -- AirNow Daily Average PM2.5"
  )
}, silent = FALSE)

## End(Not run)
```

**monitor_aqi**

*Calculate hourly NowCast-based AQI values*

Description

Nowcast and AQI algorithms are applied to the data in the monitor object. A modified `mts_monitor` object is returned where values have been replaced with their Air Quality Index equivalents. See `monitor_nowcast`.

Usage

```r
monitor_aqi(
  monitor, 
  version = c("pm", "pmAsian", "ozone"),
  includeShortTerm = FALSE,
  NAAQS = c("PM2.5", "PM2.5_2024")
)
```

Arguments

- **monitor**: `mts_monitor` object.
- **version**: Name of the type of nowcast algorithm to be used.
- **includeShortTerm**: Logical specifying whether to calculate preliminary NowCast values starting with the 2nd hour.
- **NAAQS**: Version of NAAQS levels to use. See Note.
A modified mts_monitor object containing AQI values. (A list with meta and data dataframes.)

On February 7, 2024, EPA strengthened the National Ambient Air Quality Standards for Particulate Matter (PM NAAQS) to protect millions of Americans from harmful and costly health impacts, such as heart attacks and premature death. Particle or soot pollution is one of the most dangerous forms of air pollution, and an extensive body of science links it to a range of serious and sometimes deadly illnesses. EPA is setting the level of the primary (health-based) annual PM2.5 standard at 9.0 micrograms per cubic meter to provide increased public health protection, consistent with the available health science. See PM NAAQS update.

References

https://en.wikipedia.org/wiki/Nowcast_(Air_Quality_Index)
https://www.airnow.gov/aqi/aqi-basics/

Description

The variable(s) in ... are used to specify columns of monitor$meta to use for ordering. Under the hood, this function uses arrange on monitor$meta and then reorders monitor$data to match.

Usage

monitor_arrange(monitor, ...)

Arguments

monitor mts_monitor object.
...
variables in mts$meta.

Value

A reordered version of the incoming mts time series object. (A list with meta and data dataframes.)

See Also

monitor_select
Examples

library(AirMonitor)

Camp_Fire$meta$elevation[1:10]

byElevation <-
  Camp_Fire %>%
  monitor_arrange(elevation)

byElevation$meta$elevation[1:10]

---

monitor_bestTimezone  \textit{Return the most common timezone}

Description

Evaluates all timezones in monitor and returns the most common one. In the case of a tie, the alphabetically first one is returned.

Usage

\texttt{monitor_bestTimezone(monitor = NULL)}

Arguments

monitor  \textit{mts_monitor} object.

Value

A valid base::OlsonNames() timezone.

---

monitor_check  \textit{Check an mts_monitor object for validity.}

Description

Checks on the validity of an \textit{mts_monitor} object. If any test fails, this function will stop with a warning message.

Usage

\texttt{monitor_check(monitor)}

Arguments

monitor  \textit{mts_monitor} object.
**monitorCollapse**

**Value**

In invisibly returns `TRUE` if `mts_monitor` has the correct structure. Stops with an error message otherwise.

---

**Description**

Collapses data from all time series in a `mts_monitor` into a single-time series `mts_monitor` object using the function provided in the `FUN` argument. The single-time series result will be located at the mean longitude and latitude unless longitude and latitude parameters are specified.

Any columns of `monitor$meta` that are constant across all records will be retained in the returned `mts_monitor` meta dataframe.

The core metadata associated with this location (e.g. `countryCode`, `stateCode`, `timezone`, ...) will be determined from the most common (or average) value found in `monitor$meta`. This will be a reasonable assumption for the vast majority of intended use cases where data from multiple instruments in close proximity are averaged together.

**Usage**

```r
monitorCollapse(
  monitor,
  longitude = NULL,
  latitude = NULL,
  deviceID = "generatedID",
  FUN = mean,
  na.rm = TRUE,
  ...
)
```

**Arguments**

- `monitor`: `mts_monitor` object.
- `longitude`: Longitude of the collapsed time series.
- `latitude`: Latitude of the collapsed time series.
- `deviceID`: Device identifier for the collapsed time series.
- `FUN`: Function used to collapse multiple time series.
- `na.rm`: Logical specifying whether NA values should be ignored when `FUN` is applied.
- `...`: Additional arguments to be passed on to the `apply()` function.

**Value**

A `mts_monitor` object representing a single time series. (A list with `meta` and data dataframes.)
Note

After FUN is applied, values of +/-Inf and NaN are converted to NA. This is a convenience for the common case where FUN = min/max or FUN = mean and some of the time steps have all missing values. See the R documentation for min for an explanation.

Examples

```r
library(AirMonitor)

# Lane County, Oregon AQSIDs all begin with "41039"
LaneCounty <-
  NW_Megafires %>%
  monitor_filter(stringr::str_detect(AQSID, '^41039$')) %>%
  monitor_filterDate(20150821, 20150828)

# Get min/max for all monitors
LaneCounty_min <- monitor_collapse(LaneCounty, deviceID = 'LaneCounty_min', FUN = min)
LaneCounty_max <- monitor_collapse(LaneCounty, deviceID = 'LaneCounty_max', FUN = max)

# Create plot
monitor_timeseriesPlot(
  LaneCounty,
  shadedNight = TRUE,
  main = "Lane County Range of PM2.5 Values"
)

# Add min/max lines
monitor_timeseriesPlot(LaneCounty_max, col = 'red', type = 's', add = TRUE)
monitor_timeseriesPlot(LaneCounty_min, col = 'blue', type = 's', add = TRUE)
```

---

**monitor_combine**

Combine multiple mts_monitor objects

**Description**

Create a combined mts_monitor from any number of mts_monitor objects or from a list of mts_monitor objects. The resulting mts_monitor object will contain all deviceDeploymentIDs found in any incoming mts_monitor and will have a regular time axis covering the entire range of incoming data.

If incoming time ranges are temporally non-contiguous, the resulting mts_monitor will have gaps filled with NA values.

An error is generated if the incoming mts_monitor objects have non-identical metadata for the same deviceDeploymentID unless replaceMeta = TRUE.
Usage

```r
monitor_combine(
  ..., replaceMeta = FALSE,
  overlapStrategy = c("replace all", "replace na")
)
```

Arguments

- `...`: Any number of valid `mts_monitor` objects or a list of objects.
- `replaceMeta`: Logical specifying whether to allow replacement of metadata associated when duplicate `deviceDeploymentIDs` are encountered.
- `overlapStrategy`: Strategy to use when data found in time series overlaps.

Value

A combined `mts_monitor` object. (A list with `meta` and `data` dataframes.)

Note

Data are combined with a "later is better" sensibility where any data overlaps exist. Incoming `mts_monitor` objects are ordered based on the time stamp of their last record. Any data records found in a "later" `mts_monitor` will overwrite data associated with an "earlier" `mts_monitor`.

With `overlapStrategy = "replace all"`, any data records found in "later" `mts_monitor` objects are preferentially retained before the "shared" data are finally reordered by ascending `datetime`.

With `overlapStrategy = "replace missing"`, only missing values in "earlier" `mts_monitor` objects are replaced with data records from "later" time series.

Examples

```r
library(AirMonitor)

# Two monitors near Pendelton, Oregon
#
# Use the interactive map to get the `deviceDeploymentIDs`
# NW_Megafires %>% monitor_leaflet()

Pendleton_West <-
  NW_Megafires %>%
  monitor_select("f187226671d1109a_410590121_03") %>%
  monitor_filterDateTime(2015082300, 2015082305)

Pendleton_East <-
  NW_Megafires %>%
  monitor_select("6c906c6d1cf46b53_410597002_02") %>%
  monitor_filterDateTime(2015082300, 2015082305)

monitor_combine(Pendleton_West, Pendleton_East) %>%
```
monitor_dailyBarplot

Description

Creates a daily barplot of data from a mts_monitor object.
Reasonable defaults are chosen for annotations and plot characteristics. Users can override any defaults by passing in parameters accepted by graphics::barplot.

Usage

monitor_dailyBarplot(
    monitor = NULL,
    id = NULL,
    add = FALSE,
    addAQI = FALSE,
    palette = c("EPA", "subdued", "deuteranopia"),
    opacity = NULL,
    minHours = 18,
    dayBoundary = c("clock", "LST"),
    NAAQS = c("PM2.5", "PM2.5_2024"),
    ...
)

Arguments

monitor mts_monitor object.
id deviceDeploymentID for a single time series found in monitor. (Optional if monitor contains only a single time series.)
add Logical specifying whether to add to the current plot.
addAQI Logical specifying whether to add visual AQI decorations.
palette Named color palette to use when adding AQI decorations.
opacity Opacity to use for bars.
minHours Minimum number of valid hourly records per day required to calculate statistics. Days with fewer valid records will be assigned NA.
dayBoundary Treatment of daylight savings time: "clock" uses daylight savings time as defined in the local timezone, "LST" uses "local standard time" all year round.
NAAQS Version of NAAQS levels to use. See Note.
... Additional arguments to be passed to graphics::barplot().
**monitor_dailyStatistic**

Create daily statistics for each monitor in an mts_monitor object

**Description**

Daily statistics are calculated for each time series in monitor$data using FUN and any arguments passed in.

Because the returned mts_monitor object is defined on a daily axis in a specific time zone, it is important that the incoming monitor contain timeseries associated with a single time zone.

**Value**

No return value. This function is called to draw an air quality daily average plot on the active graphics device.

**Note**

The underlying axis for this plot is not a time axis so you cannot use this function to "add" bars on top of a monitor_timeseriesPlot(). See the [AirMonitorPlots](https://example.com) package for more flexibility in plotting.

On February 7, 2024, EPA strengthened the National Ambient Air Quality Standards for Particulate Matter (PM NAAQS) to protect millions of Americans from harmful and costly health impacts, such as heart attacks and premature death. Particle or soot pollution is one of the most dangerous forms of air pollution, and an extensive body of science links it to a range of serious and sometimes deadly illnesses. EPA is setting the level of the primary (health-based) annual PM2.5 standard at 9.0 micrograms per cubic meter to provide increased public health protection, consistent with the available health science. See [PM NAAQS update](https://example.com).

**Examples**

```r
library(AirMonitor)

layout(matrix(seq(2)))

Carmel_Valley %>% monitor_dailyBarplot()
title("(pre-2024 PM NAAQS)", line = 0)

Carmel_Valley %>% monitor_dailyBarplot(NAAQS = "PM2.5_2024")
title("(updated PM NAAQS)", line = 0)

layout(1)
```
monitor_dailyStatistic

Usage

```r
monitor_dailyStatistic(
  monitor = NULL,
  FUN = mean,
  na.rm = TRUE,
  minHours = 18,
  dayBoundary = c("clock", "LST"),
  ...
)
```

Arguments

- `monitor` *mts_monitor* object.
- `FUN` Function used to create daily statistics.
- `na.rm` Value passed on to `FUN`. If `FUN` does not use `na.rm`, this should be set to `NULL`.
- `minHours` Minimum number of valid hourly records per day required to calculate statistics. Days with fewer valid records will be assigned `NA`.
- `dayBoundary` Treatment of daylight savings time: "clock" uses daylight savings time as defined in the local timezone, "LST" uses "local standard time" all year round.
- `...` Additional arguments to be passed to `FUN`.

Value

A *mts_monitor* object containing daily statistical summaries. (A list with *meta* and *data* dataframes.)

Note

When `dayBoundary = "clock"`, the returned `monitor$data$datetime` time axis will be defined in the local timezone (not "UTC") with days defined by midnight as it appears on a clock in that timezone. The transition from DST to standard time will result in a 23 hour day and standard to DST in a 25 hour day.

When `dayBoundary = "LST"`, the returned `monitor$data$datetime` time axis will be defined in "UTC" with times as they appear in standard time in the local timezone. These days will be one hour off from clock time during DST but every day will consist of 24 hours.

Examples

```r
library(AirMonitor)

Carmel_Valley %>%
  monitor_dailyStatistic(max) %>%
  monitor_getData()

Carmel_Valley %>%
  monitor_dailyStatistic(min) %>%
  monitor_getData()
```
monitor_dailyThreshold

*Daily counts of values at or above a threshold*

**Description**

Calculates the number of hours per day each time series in `monitor` was at or above a given threshold.

Because the returned `mts_monitor` object is defined on a daily axis in a specific time zone, it is important that the incoming `monitor` contain only timeseries within a single time zone.

**Usage**

```r
monitor_dailyThreshold(
  monitor = NULL,
  threshold = NULL,
  na.rm = TRUE,
  minHours = 18,
  dayBoundary = c("clock", "LST"),
  NAAQS = c("PM2.5", "PM2.5_2024")
)
```

**Arguments**

- `monitor`: `mts_monitor` object.
- `threshold`: AQI level name (e.g. "unhealthy") or numerical threshold at and above which a measurement is counted.
- `na.rm`: Logical value indicating whether NA values should be ignored.
- `minHours`: Minimum number of valid hourly records per day required to calculate statistics. Days with fewer valid records will be assigned NA.
- `dayBoundary`: Treatment of daylight savings time: "clock" uses daylight savings time as defined in the local timezone, "LST" uses "local standard time" all year round.
- `NAAQS`: Version of NAAQS levels to use. See Note.

**Value**

*A `mts_monitor` object containing daily counts of hours at or above a threshold value. (A list with `meta` and `data` dataframes.)*

**Note**

When `dayBoundary = "clock"`, the returned `monitor$data$datetime` time axis will be defined in the local timezone (not "UTC") with days defined by midnight as it appears on a clock in that timezone. The transition from DST to standard time will result in a 23 hour day and standard to DST in a 25 hour day.
When `dayBoundary = "LST"`, the returned `monitor$datetime` time axis will be defined in "UTC" with times as they appear in standard time in the local timezone. These days will be one hour off from clock time during DST but every day will consist of 24 hours.

On February 7, 2024, EPA strengthened the National Ambient Air Quality Standards for Particulate Matter (PM NAAQS) to protect millions of Americans from harmful and costly health impacts, such as heart attacks and premature death. Particle or soot pollution is one of the most dangerous forms of air pollution, and an extensive body of science links it to a range of serious and sometimes deadly illnesses. EPA is setting the level of the primary (health-based) annual PM2.5 standard at 9.0 micrograms per cubic meter to provide increased public health protection, consistent with the available health science. See PM NAAQS update.

### Examples

```r
library(AirMonitor)

# Hours at MODERATE or above
Carmel_Valley %>%
  monitor_dailyThreshold("Moderate") %>%
  monitor_getData()

# Hours at MODERATE or above with the 2024 updated NAAQS
Carmel_Valley %>%
  monitor_dailyThreshold("Moderate", NAAQS = "PM2.5_2024") %>%
  monitor_getData()

# Hours at UNHEALTHY or above
Carmel_Valley %>%
  monitor_dailyThreshold("Unhealthy") %>%
  monitor_getData()
```

---

```r
~

**monitor_distinct**

Retain only distinct data records in `monitor$data`

**Description**

Two successive steps are used to guarantee that the `datetime` axis contains no repeated values:

1. remove any duplicate records
2. guarantee that rows are in `datetime` order

**Usage**

`monitor_distinct(monitor)`

**Arguments**

- **monitor**: `mts_monitor` object
\textit{monitor\_dropEmpty}

\textbf{Value}

A \textit{mts\_monitor} object with no duplicated data records. (A list with \textit{meta} and \textit{data} dataframes.)

\textbf{Note}

This function is primarily for package-internal use.

\begin{verbatim}
monitor_dropEmpty
\end{verbatim}

\textit{monitor\_dropEmpty} \hspace{1cm} \textit{Drop device deployments with all missing data}

\textbf{Description}

The incoming \textit{mts\_monitor} object is subset to retain only time series with valid data.

\textbf{Usage}

\begin{verbatim}
mirror\_dropEmpty(monitor)
\end{verbatim}

\textbf{Arguments}

\begin{verbatim}
monitor\hspace{1cm} \textit{mts\_monitor} object. (A list with \textit{meta} and \textit{data} dataframes.)
\end{verbatim}

\textbf{Value}

A subset of the incoming \textit{mts\_monitor}. (A list with \textit{meta} and \textit{data} dataframes.)

\begin{verbatim}
monitor\_dygraph
\end{verbatim}

\textit{monitor\_dygraph} \hspace{1cm} \textit{Create Interactive Time Series Plot}

\textbf{Description}

This function creates interactive graphs that will be displayed in RStudio’s 'Viewer' tab.

\textbf{Usage}

\begin{verbatim}
monitor\_dygraph(
    monitor,
    title = "title",
    ylab = "PM2.5 Concentration",
    rollPeriod = 1,
    showLegend = TRUE
)
\end{verbatim}
monitor_filterByDistance

Arguments

- **monitor**: `mts_monitor` object.
- **title**: Title text.
- **ylab**: Title for the y axis
- **rollPeriod**: Rolling mean to be applied to the data.
- **showLegend**: Logical to toggle display of the legend.

Value

Initiates the interactive dygraph plot in RStudio’s ‘Viewer’ tab.

Examples

```r
## Not run:
library(AirMonitor)

# Multiple monitors
Camp_Fire %>%
  monitor_filter(countyName == "Alameda") %>%
  monitor_dygraph()
## End(Not run)
```

Description

Filters the `monitor` argument to include only those time series located within a certain radius of a target location. If no time series fall within the specified `radius`, an empty `mts_monitor` object will be returned.

When `count` is used, a `mts_monitor` object is created containing up to `count` time series, ordered by increasing distance from the target location. Note that the number of monitors returned may be less than the specified `count` value if fewer than `count` time series are found within the target area.

Usage

```r
monitor_filterByDistance(
  monitor,
  longitude = NULL,
  latitude = NULL,
  radius = 50,
  count = NULL,
  addToMeta = FALSE
)
```
monitor_filterDate

Arguments

monitor mts_monitor object.
longitude Target longitude.
latitude Target.
radius Distance (m) of radius defining a target area.
count Number of time series to return.
addToMeta Logical specifying whether to add distanceFromTarget as a field in monitor$meta.

Value

A mts_monitor object with monitors near a location.

Note

The returned mts_monitor will have an extra distance. (A list with meta and data dataframes.)

Examples

library(AirMonitor)

# Walla Walla
longitude <- -118.330278
latitude <- 46.065

Walla_Walla_monitors <-
NW_Megafires %>%
  monitor_filterByDistance(
    longitude = -118.330,
    latitude = 46.065,
    radius = 50000, # 50 km
    addToMeta = TRUE
  )

Walla_Walla_monitors %>%
  monitor_getMeta() %>%
  dplyr::select(c("locationName", "distanceFromTarget"))

monitor_filterDate Date filtering for mts_monitor objects

Description

Subsets a mts_monitor object by date. This function always filters to day-boundaries. For sub-day filtering, use monitor_filterDatetime().

Dates can be anything that is understood by MazamaCoreUtils::parseDatetime() including either of the following recommended formats:
monitor_filterDate

- "YYYYmmd"  
- "YYYY-mm-dd"

If either startdate or enddate is not provided, the start/end of the mts_monitor time axis will be used.

Timezone determination precedence assumes that if you are passing in POSIXct values then you know what you are doing.

1. get timezone from startdate if it is POSIXct
2. use passed in timezone
3. get timezone from mts_monitor

Usage

```r
monitor_filterDate(
  monitor = NULL,
  startdate = NULL,
  enddate = NULL,
  timezone = NULL,
  unit = "sec",
  ceilingStart = FALSE,
  ceilingEnd = FALSE
)
```

Arguments

- **monitor**  
  *mts_monitor* object.
- **startdate**  
  Desired start datetime (ISO 8601).
- **enddate**  
  Desired end datetime (ISO 8601).
- **timezone**  
  Olson timezone used to interpret dates.
- **unit**  
  Units used to determine time at end-of-day.
- **ceilingStart**  
  Logical instruction to apply `ceiling_date` to the startdate rather than `floor_date`
- **ceilingEnd**  
  Logical instruction to apply `ceiling_date` to the enddate rather than `floor_date`

Value

A subset of the given *mts_monitor* object. (A list with meta and data dataframes.)

Note

The returned data will run from the beginning of startdate until the **beginning** of enddate – *i.e.* no values associated with enddate will be returned. The exception being when enddate is less than 24 hours after startdate. In that case, a single day is returned.

See Also

- `monitor_filterDatetime`
- `monitor_filterMeta`
**monitor_filterDatetime**

*Datetime filtering for mts_monitor objects*

**Description**

Subsets a `mts_monitor` object by datetime. This function allows for sub-day filtering as opposed to `monitor_filterDate()` which always filters to day-boundaries.

Datetimes can be anything that is understood by MazamaCoreUtils::parseDatetime(). For non-POSIXct values, the recommended format is "YYYY-mm-dd HH:MM:SS".

If either `startdate` or `enddate` is not provided, the start/end of the `mts_monitor` time axis will be used.

Timezone determination precedence assumes that if you are passing in POSIXct values then you know what you are doing.

1. get timezone from `startdate` if it is POSIXct
2. use passed in timezone
3. get timezone from `mts_monitor`
Usage

monitor_filterDatetime(
  monitor = NULL,
  startdate = NULL,
  enddate = NULL,
  timezone = NULL,
  unit = "sec",
  ceilingStart = FALSE,
  ceilingEnd = FALSE
)

Arguments

monitor mts_monitor object.
startdate Desired start datetime (ISO 8601).
enddate Desired end datetime (ISO 8601).
timezone Olson timezone used to interpret startdate and enddate.
unit Units used to determine time at end-of-day.
ceilingStart Logical specifying application of ceiling_date to the startdate rather than floor_date
ceilingEnd Logical specifying application of ceiling_date to the enddate rather than floor_date

Value

A subset of the given mts_monitor object. (A list with meta and data dataframes.)

See Also

monitor_filterDate
monitor_filterMeta

Examples

library(AirMonitor)

Camp_Fire %>%
  monitor_timeRange()

# Reduced time range returned in "UTC"
Camp_Fire %>%
  monitor_filterDatetime(
    "2018-11-15 02:00:00",
    "2018-11-22 06:00:00",
    timezone = "America/Los_Angeles"
  ) %>%
  monitor_timeRange()
monitor_filterMeta

# Reduced time range returned in "America/Los_Angeles"
Camp_Fire %>%
  monitor_filterDatetime(
    "2018111502",
    "2018112206",
    timezone = "America/Los_Angeles"
  ) %>%
  monitor_timeRange(
    timezone = "America/Los_Angeles"
  )

monitor_filterMeta  General purpose metadata filtering for mts_monitor objects

Description

A generalized metadata filter for mts_monitor objects to choose cases where conditions are true. Multiple conditions are combined with & or separated by a comma. Only rows where the condition evaluates to TRUE are kept. Rows of monitor$meta where the condition evaluates to NA are dropped. Associated columns of monitor$data are also dropped for internal consistency in the returned mts_monitor object.

monitor_filter() is an alias for monitor_filterMeta().

Usage

monitor_filterMeta(monitor,...)

monitor_filter(monitor,...)

Arguments

monitor  mts_monitor object.
...

Logical predicates defined in terms of the variables in monitor$meta.

Value

A subset of the incoming mts_monitor. (A list with meta and data dataframes.)

Note

Filtering is done on variables in monitor$meta.

See Also

monitor_filterDate
monitor_filterDatetime
Examples

```r
library(AirMonitor)

# Filter based on countyName field
Camp_Fire %>%
    monitor_filter(countyName == "Alameda") %>%
    monitor_timeseriesPlot(main = "All Alameda County Monitors")

# Filter combining two fields
Camp_Fire %>%
    monitor_filter(latitude > 39.5, longitude > -121.5) %>%
    monitor_pull("locationName")

# Filter using string matching
Camp_Fire %>%
    monitor_filter(stringr::str_detect(locationName, "^San")) %>%
    monitor_pull("locationName")
```

---

**monitor_fromPWFSLSmoke**

Convert a ws_monitor object from the PWFSLSmoke package

---

**Description**

A PWFSLSmoke package ws_monitor object is enhanced and modified so that it becomes a valid mts_monitor object. This is a lossless operation and can be reversed with monitor_toPWFSLSmoke().

**Usage**

```r
monitor_fromPWFSLSmoke(ws_monitor = NULL)
```

**Arguments**

- `ws_monitor` : ws_monitor object. (A list with meta and data dataframes.)

**Value**

A mts_monitor object.
monitor_getCurrentStatus

*Get current status of monitors*

**Description**

This function augments `monitor$meta` with summary information derived from `monitor$data` reflecting recent measurements.

**Usage**

```r
monitor_getCurrentStatus(
  monitor,
  enddate = NULL,
  minHours = 18,
  dayBoundary = c("clock", "LST")
)
```

**Arguments**

- `monitor` *mts_monitor* object.
- `enddate` Time relative to which current status is calculated. By default, it is the latest time in `monitor$data$datetime`. This time can be given as a POSIXct time, or a string/numeric value in ymd format (e.g. 20190301). This time converted to UTC.
- `minHours` Minimum number of valid hourly records required to calculate `yesterday_PM2.5_avg`. Days with fewer valid records will be assigned NA.
- `dayBoundary` Treatment of daylight savings time: "clock" uses daylight savings time as defined in the local timezone, "LST" uses "local standard time" all year round. (See `monitor_dailyStatistic()` for more details.)

**Value**

The `monitor$meta` table augmented with current status information for each time series.

"Last" and "Previous"

The goal of this function is to provide useful information about what happened recently with each time series in the provided *mts_monitor* object. Devices don’t always consistently report data, however, and it is not always useful to have NA’s reported when there is recent valid data at earlier times. To address this, `monitor_getCurrentStatus()` uses `last` and `previous` valid times. These are the time when a monitor most recently reported data, and the most recent time of valid data before that, respectively. By reporting on these times, this function ensures that valid data is returned and provides information on how outdated this information is. This information can be used in maps to show AQI colored dots when data is only a few hours old but gray dots when data is older than some threshold.
Calculating latency

According to https://docs.airnowapi.org/docs/HourlyDataFactSheet.pdf a datum assigned to 2pm represents the average of data between 2pm and 3pm. So, if we check at 3:15pm and see that we have a value for 2pm but not 3pm then the data are completely up-to-date with zero latency.

monitor_getCurrentStatus() defines latency as the difference between a time index and the next most recent time index associated with a valid value. If there is no more recent time index, then the difference is measured to the given enddate parameter. Because mts_monitor objects are defined on an hourly axis, these differences have units of hours.

For example, if the recorded values for a monitor are [16.2, 15.8, 16.4, NA, 14.0, 12.5, NA, NA, 13.3, NA], then the last valid value is 13.3 with an index is 9, and the previous valid value is 12.4 with an index of 6. The last latency is then 1 (hour before the end), and the previous latency is 3 (hours before the last valid value).

Summary data

The table created by monitor_getCurrentStatus() includes per-time series summary information calculated from monitor$data. The additional data fields added to monitor$meta are listed below:

- currentStatus_processingTime: Time at which this function was run
- currentStatus_enddate: Time relative to which "currency" is calculated
- last_validIndex: Row index of the last valid measurement in monitor$data
- previous_validIndex: Row index of the previous valid measurement in monitor$data
- last_validTime: UTC time associated with last_validIndex
- previous_validTime: UTC time associated with previous_validIndex
- last_latency: Hours between last_validTime and endtime
- previous_latency: Hours between previous_validTime and last_validTime
- last_validLocalTimestamp: Local time representation of last_validTime
- previous_validLocalTimestamp: Local time representation of previous_validTime
- last_PM2.5: Last valid PM2.5 measurement
- previous_PM2.5: Previous valid PM2.5 measurement
- last_nowcast: Last valid PM2.5 NowCast value
- previous_nowcast: Previous valid PM2.5 NowCast value
- yesterday_PM2.5_avg: Daily average PM2.5 for the day prior to enddate

Examples

```r
# Fail gracefully if any resources are not available
try{

library(AirMonitor)

monitor <- airnow_loadLatest()
# TODO: Needed before rebuilding of v2 database with fullAQSID
```
monitor$meta$fullAQSID <- paste0("840", monitor$meta$AQSID)

currentStatus <-
  monitor %>%
  monitor_filter(stateCode == "WA") %>%
  monitor_getCurrentStatus()

}, silent = FALSE)

---

**monitor_getDataFrame**  
*Extract dataframes from mts_monitor objects*

**Description**

These functions are convenient wrappers for extracting the dataframes that comprise a `mts_monitor` object. These functions are designed to be useful when manipulating data in a pipeline using `%>%`.

Below is a table showing equivalent operations for each function.

<table>
<thead>
<tr>
<th>Function</th>
<th>Equivalent Operation</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>monitor_getData()</code></td>
<td><code>monitor$data</code></td>
</tr>
<tr>
<td><code>monitor_getMeta()</code></td>
<td><code>monitor$meta</code></td>
</tr>
</tbody>
</table>

**Usage**

- `monitor_getData(monitor)`
- `monitor_getMeta(monitor)`

**Arguments**

- `monitor`  
  *mts_monitor* object to extract dataframe from.

**Value**

A dataframe from the given `mts_monitor` object.

---

**monitor_getDistance**  
*Calculate distances from mts_monitor locations to a location of interest*

**Description**

This function returns the distances (meters) between `monitor` locations and a location of interest. These distances can be used to create a mask identifying monitors within a certain radius of the location of interest.
monitor_getDistance

Usage

monitor_getDistance(
  monitor = NULL,
  longitude = NULL,
  latitude = NULL,
  measure = c("geodesic", "haversine", "vincenty", "cheap")
)

Arguments

monitor  mts_monitor object.
longitude Longitude of the location of interest.
latitude Latitude of the location of interest.
measure One of "geodesic", "haversine" "vincenty", or "cheap".

Value

Named vector of distances (meters) with each distance identified by deviceDeploymentID.

Note

The measure "cheap" may be used to speed things up depending on the spatial scale being considered. Distances calculated with measure = "cheap" will vary by a few meters compared with those calculated using measure = "geodesic".

Examples

library(AirMonitor)

# Walla Walla
longitude <- -118.3302
latitude <- 46.065

distance <- monitor_getDistance(NW_Megafires, longitude, latitude)
closestIndex <- which(distance == min(distance))

# Distance in meters
distance[closestIndex]

# Monitor core metadata
str(NW_Megafires$meta[closestIndex, AirMonitor::coreMetadataNames])
### monitor_isEmpty

*Test for an empty mts_monitor object*

**Description**

This function returns true under the following conditions:

- no time series: \( \text{ncol(monitor$data)} == 1 \)
- no time series records: \( \text{nrow(monitor$data)} == 0 \)
- all timeseries values are NA

This makes for more readable code in functions that need to test for this.

**Usage**

```r
monitor_isEmpty(monitor)
```

**Arguments**

- `monitor` *mts_monitor object*

**Value**

Invisibly returns TRUE if no data exist in mts_monitor, FALSE otherwise.

### monitor_isValid

*Test mts_monitor object for correct structure*

**Description**

The mts_monitor is checked for the presence of core meta and data columns. Core meta columns include: (TODO: complete this list)

- deviceDeploymentID – unique identifier (see MazmaLocationUtils)
- deviceID – device identifier
- locationID – location identifier (see MazmaLocationUtils)
- locationName – English language name
- longitude – decimal degrees E
- latitude – decimal degrees N
- elevation – elevation of station in m
- countryCode – ISO 3166-1 alpha-2
- timezone – Olson time zone

Core data columns include:

- datetime – measurement time (UTC)
monitor_isValid(monitor = NULL, verbose = FALSE)

Arguments

monitor: mts_monitor object
verbose: Logical specifying whether to produce detailed warning messages.

Value

Invisibly returns TRUE if mts_monitor has the correct structure, FALSE otherwise.

monitor_leaflet

Leaflet interactive map of monitor locations

Description

This function creates interactive maps that will be displayed in RStudio’s ‘Viewer’ tab. The slice argument is used to collapse a mts_monitor timeseries into a single value. If slice is an integer, that row index will be selected from the monitor$data dataframe. If slice is a function (unquoted), that function will be applied to the timeseries with the argument na.rm=TRUE (e.g. max(..., na.rm=TRUE)).

If slice is a user defined function, it will be used with argument na.rm=TRUE to collapse the time dimension. Thus, user defined functions must accept na.rm as an argument.

Usage

monitor_leaflet(
  monitor,
  slice = "max",
  radius = 10,
  opacity = 0.7,
  maptype = "terrain",
  extraVars = NULL,
  jitter = 5e-04,
  NAAQS = c("PM2.5", "PM2.5_2024"),
  ...
)

Arguments

monitor: mts_monitor object.
slice: Either a formatted time string, a time index, the name of a (potentially user defined) function used to collapse the time axis.
radius: radius of monitor circles
The `maptype` argument is mapped onto leaflet "ProviderTile" names. Current map types include:

"roadmap" – "OpenStreetMap"
"satellite" – "Esri.WorldImagery"
"terrain" – "Esri.WorldTopoMap"
"toner" – "Stamen.Toner"

If a character string not listed above is provided, it will be used as the underlying map tile if available. See https://leaflet-extras.github.io/leaflet-providers/ for a list of "provider tiles" to use as the background map.

Value

Invisibly returns a leaflet map of class "leaflet".

Note

On February 7, 2024, EPA strengthened the National Ambient Air Quality Standards for Particulate Matter (PM NAAQS) to protect millions of Americans from harmful and costly health impacts, such as heart attacks and premature death. Particle or soot pollution is one of the most dangerous forms of air pollution, and an extensive body of science links it to a range of serious and sometimes deadly illnesses. EPA is setting the level of the primary (health-based) annual PM2.5 standard at 9.0 micrograms per cubic meter to provide increased public health protection, consistent with the available health science. See PM NAAQS update.

Examples

```r
## Not run:
library(AirMonitor)
# Fail gracefully if any resources are not available
try{

# Maximum AQI category at each site
monitor_loadLatest() %>%
  monitor_filter(stateCode %in% CONUS) %>%
  monitor_leaflet()
```
# Mean AQI category at each site
monitor_loadLatest() %>%
  monitor_filter(stateCode %in% CONUS) %>%
  monitor_leaflet(
    slice = "mean"
  )

# Mean AQI category at each site using the updated NAAQS
monitor_loadLatest() %>%
  monitor_filter(stateCode %in% CONUS) %>%
  monitor_leaflet(
    slice = "mean",
    NAAQS = "PM2.5_2024"
  )
}, silent = FALSE)

## End(Not run)

---

## monitor_load

Load monitoring data from all sources

### Description

Loads monitoring data for a given time range. Data from AirNow, AIRSIS and WRCC are combined into a single `mts_monitor` object.

Archival datasets are combined with 'daily' and 'latest' datasets as needed to satisfy the requested date range.

### Usage

```r
monitor_load(
  startdate = NULL,
  enddate = NULL,
  timezone = NULL,
  archiveBaseUrl = paste0("https://airfire-data-exports.s3.us-west-2.amazonaws.com/",
                          "monitoring/v2"),
  archiveBaseDir = NULL,
  QC_negativeValues = c("zero", "na", "ignore"),
  epaPreference = c("airnow", "epa_aqs")
)
```

### Arguments

- **startdate**: Desired start datetime (ISO 8601).
- **enddate**: Desired end datetime (ISO 8601).
- **timezone**: Olson timezone used to interpret dates.
monitor_loadAnnual

archiveBaseUrl Base URL for monitoring v2 data files.
archiveBaseDir Local base directory for monitoring v2 data files.
QC_negativeValues Type of QC to apply to negative values. Files are available from both 'epa' and 'airnow'.
epaPreference Preferred data source for EPA data when annual data files are available from both 'epa_aqs' and 'airnow'.

Value

A mts_monitor object with PM2.5 monitoring data. (A list with meta and data dataframes.)

See Also

monitor_loadAnnual
monitor_loadDaily
monitor_loadLatest

Examples

## Not run:
library(AirMonitor)

# Fail gracefully if any resources are not available
try({
  wa <-
  monitor_load(20210601, 20211001) %>%
  monitor_filter(stateCode == "WA")

  monitor_timeseriesPlot(wa)
}, silent = FALSE)

## End(Not run)

monitor_loadAnnual Load annual monitoring data from all sources

Description

Combine annual data from AirNow, AIRSIS and WRCC:
If archiveDataDir is defined, data will be loaded from this local archive. Otherwise, data will be loaded from the monitoring data repository maintained by the USFS AirFire team.
Current year files loaded by this function are updated once per week.
For the most recent data in the last 10 days, use monitor_loadLatest().
For daily updates covering the most recent 45 days, use monitor_loadDaily().
For data extended more than 45 days into the past, use monitor_load().
Usage

```r
monitor_loadAnnual(
  year = NULL,
  archiveBaseUrl = paste0("https://airfire-data-exports.s3.us-west-2.amazonaws.com/", "monitoring/v2"),
  archiveBaseDir = NULL,
  QC_negativeValues = c("zero", "na", "ignore"),
  epaPreference = c("airnow", "epa_aqs")
)
```

Arguments

- **year**: Year [YYYY].
- **archiveBaseUrl**: Base URL for monitoring v2 data files.
- **archiveBaseDir**: Local base directory for monitoring v2 data files.
- **QC_negativeValues**: Type of QC to apply to negative values.
- **epaPreference**: Preferred data source for EPA data when annual data files are available from both 'epa_aqs' and 'airnow'.

Value

A `mts_monitor` object with PM2.5 monitoring data. (A list with `meta` and `data` dataframes.)

Note

This function guarantees that only a single time series will be associated with each `locationID` using the following logic:

1. AirNow data takes precedence over data from AIRSIS or WRCC
2. more recent data takes precedence over older data

This relevant mostly for "temporary" monitors which may be replaced after they are initially deployed. If you want access to all device deployments associated with a specific `locationID`, you can use the provider specific functions: `airnow_loadAnnual`, `airsis_loadAnnual` and `wrcc_loadAnnual`

See Also

- `monitor_loadDaily`
- `monitor_loadLatest`

Examples

```r
## Not run:
library(AirMonitor)
# Fail gracefully if any resources are not available
try({
```
monitor_loadDaily

monitor_loadAnnual() %>%
  monitor_filter(stateCode %in% CONUS) %>%
  monitor_leaflet()

}, silent = FALSE)

## End(Not run)

monitor_loadDaily

Load daily monitoring data from all sources

Description

Combine daily data from AirNow, AIRSIS and WRCC:

If archiveDataDir is defined, data will be loaded from this local archive. Otherwise, data will be loaded from the monitoring data repository maintained by the USFS AirFire team.

The files loaded by this function are updated once per day and contain data for the previous 45 days.

For the most recent data in the last 10 days, use monitor_loadLatest().

For data extended more than 45 days into the past, use monitor_load().

Usage

monitor_loadDaily(
  archiveBaseUrl = paste0("https://airfire-data-exports.s3.us-west-2.amazonaws.com/", 
    "monitoring/v2"),
  archiveBaseDir = NULL,
  QC_negativeValues = c("zero", "na", "ignore")
)

Arguments

archiveBaseUrl  Base URL for monitoring v2 data files.
archiveBaseDir  Local base directory for monitoring v2 data files.
QC_negativeValues  Type of QC to apply to negative values.

Value

A mts_monitor object with PM2.5 monitoring data. (A list with meta and data dataframes.)

Note

This function guarantees that only a single time series will be associated with each locationID using the following logic:

1. AirNow data takes precedence over data from AIRSIS or WRCC
2. more recent data takes precedence over older data

This relevant mostly for “temporary” monitors which may be replaced after they are initially deployed. If you want access to all device deployments associated with a specific locationID, you can use the provider specific functions: airnow_loadDaily, airsis_loadDaily and wrcc_loadDaily

See Also

monitor_loadAnnual
monitor_loadLatest

Examples

```r
## Not run:
library(AirMonitor)
# Fail gracefully if any resources are not available
try({
  monitor_loadDaily() %>%
    monitor_filter(stateCode %in% CONUS) %>%
    monitor_leaflet()
}, silent = FALSE)
## End(Not run)
```

---

### monitor_loadLatest

Load most recent monitoring data from all sources

#### Description

Combine recent data from AirNow, AIRSIS and WRCC:

- If archiveDataDir is defined, data will be loaded from this local archive. Otherwise, data will be loaded from the monitoring data repository maintained by the USFS AirFire team.
- The files loaded by this function are updated multiple times an hour and contain data for the previous 10 days.
- For daily updates covering the most recent 45 days, use monitor_loadDaily().
- For data extended more than 45 days into the past, use monitor_load().

#### Usage

```r
monitor_loadLatest(
  archiveBaseUrl = paste0("https://airfire-data-exports.s3.us-west-2.amazonaws.com/",
                         "monitoring/v2"),
  archiveBaseDir = NULL,
  QC_negativeValues = c("zero", "na", "ignore")
)
```
monitor_loadLatest

Arguments

archiveBaseUrl  Base URL for monitoring v2 data files.
archiveBaseDir  Local base directory for monitoring v2 data files.
QC_negativeValues  Type of QC to apply to negative values.

Value

A mts_monitor object with PM2.5 monitoring data. (A list with meta and data dataframes.)

Note

This function guarantees that only a single time series will be associated with each locationID using the following logic:

1. AirNow data takes precedence over data from AIRSIS or WRCC
2. more recent data takes precedence over older data

This relevant mostly for "temporary" monitors which may be replaced after they are initially deployed. If you want access to all device deployments associated with a specific locationID, you can use the provider specific functions: airnow_loadLatest, airsis_loadLatest and wrcc_loadLatest

See Also

monitor_loadAnnual
monitor_loadDaily

Examples

## Not run:
library(AirMonitor)
# Fail gracefully if any resources are not available
try({
  monitor_loadLatest() %>%
    monitor_filter(stateCode %in% CONUS) %>%
    monitor_leaflet()
}, silent = FALSE)

## End(Not run)
**monitor_mutate**

Apply a function to mts_monitor time series

Description

This function works similarly to dplyr::mutate() and applies FUN to each time series found in `monitor$data`. FUN must be a function that accepts a numeric vector as its first argument and returns a vector of the same length.

Usage

```r
monitor_mutate(monitor = NULL, FUN = NULL, ...)
```

Arguments

- `monitor`: mts_monitor object.
- `FUN`: Function used to modify time series.
- `...`: Additional arguments to be passed to FUN.

Value

A modified mts_monitor object. (A list with meta and data dataframes.)

Examples

```r
library(AirMonitor)

Carmel_Valley %>%
  monitor_filterDatetime(2016080207, 2016080212) %>%
  monitor_toCSV(includeMeta = FALSE) %>%
  cat()

Carmel_Valley %>%
  monitor_filterDatetime(2016080207, 2016080212) %>%
  monitor_mutate(function(x) { return(x / 2) }) %>%
  monitor_toCSV(includeMeta = FALSE) %>%
  cat()
```
Monitor NowCast

A NowCast algorithm is applied to the data in the monitor object. The version argument specifies the minimum weight factor and number of hours to be used in the calculation.

Available versions include:

1. pm: hours = 12, weight = 0.5
2. pmAsian: hours = 3, weight = 0.1
3. ozone: hours = 8, weight = NA

The default, version = "pm", is appropriate for typical usage.

Usage

```r
monitor_nowcast(
  monitor,
  version = c("pm", "pmAsian", "ozone"),
  includeShortTerm = FALSE
)
```

Arguments

- **monitor**: mts_monitor object.
- **version**: Name of the type of nowcast algorithm to be used.
- **includeShortTerm**: Logical specifying whether to calculate preliminary NowCast values starting with the 2nd hour.

Details

This function calculates each hour’s NowCast value based on the value for the given hour and the previous N-1 hours, where N is the number of hours appropriate for the version argument. For example, if version = "pm", the NowCast value for Hour 12 is based on the data from hours 1-12.

The function returns values when at least two of the previous three hours have data. NA’s are returned for hours where this condition is not met.

By default, the function will not return a valid value until the Nth hour. If includeShortTerm = TRUE, the function will return a valid value after only the 2nd hour (provided, of course, that both hours are valid).

Calculated Nowcast values are truncated to the nearest .1 ug/m3 for ‘pm’ and nearest .001 ppm for ‘ozone’ regardless of the precision of the data in the incoming mts_monitor object.
Value

A modified mts_monitor object. (A list with meta and data dataframes.)

References

https://en.wikipedia.org/wiki/Nowcast_(Air_Quality_Index)

AQI Technical Assistance Document

---

monitor_pull

Extract a column of metadata or data

Description

This function acts similarly to pull working on monitor$meta or monitor$data. Data are returned as a simple array. Data are pulled from whichever dataframe contains var.

Usage

monitor_pull(monitor = NULL, var = NULL)

Arguments

monitor

mts_monitor object.

var

A variable name found in the meta or data dataframe of the incoming mts_monitor time series object.

Value

An array of values.

Examples

library(AirMonitor)

# Metadata
Camp_Fire %>%
  monitor_pull("deploymentType") %>%
  table()

# Data for a specific ID
Camp_Fire %>%
  monitor_dailyStatistic(mean) %>%
  monitor_pull("6bbab08e3786ef66_840060450006") %>%
  round(0)

# Associated dates
Camp_Fire %>%
  monitor_dailyStatistic(mean) %>>
**monitor_replaceValues**

monitor_pull("datetime")

---

**monitor_replaceValues**  Replace mts_monitor data with another value

---

**Description**

Use an R expression to identify values for replacement.

The R expression given in `filter` is used to identify elements in `monitor$data` that should be replaced. The `datetime` column will be retained unmodified. Typical usage would include

1. replacing negative values with 0
2. replacing unreasonably high values with `NA`

Expressions should use `data` for the left hand side of the comparison.

**Usage**

```r
monitor_replaceValues(monitor = NULL, filter = NULL, value = NULL)
```

**Arguments**

- `monitor`  
  *mts_monitor* object.
- `filter`  
  R expression used to identify values for replacement.
- `value`  
  Numeric replacement value.

**Value**

A modified `mts_monitor` object. (A list with meta and data dataframes.)

**Examples**

```r
library(AirMonitor)

wa <- monitor_filterMeta(NW_Megafires, stateCode == 'WA')
any(wa$data < 5, na.rm = TRUE)

wa_zero <- monitor_replaceValues(wa, data < 5, 5)
any(wa_zero$data < 5, na.rm = TRUE)
```
monitor_select

Subset and reorder time series within an mts_monitor object

Description

This function acts similarly to dplyr::select() working on `monitor$data`. The returned `mts_monitor` object will contain only those time series identified by `id` in the order specified.

This can be helpful when using faceted plot functions based on `ggplot` such as those found in the `AirMonitorPlots` package.

Usage

```r
monitor_select(monitor, id)
monitor_reorder(monitor, id)
```

Arguments

- `monitor` mts_monitor object.
- `id` Vector of deviceDeploymentIDs.

Value

A reordered (subset) of the incoming `mts_monitor` object. (A list with `meta` and `data` dataframes.)

See Also

`monitor_filterMeta`

monitor_selectWhere

Data-based subsetting of time series within an mts_monitor object.

Description

Subsetting of `monitor` acts similarly to tidyselect::where() working on `monitor$data`. The returned `mts_monitor` object will contain only those time series where `FUN` applied to the time series data returns TRUE.

Usage

```r
monitor_selectWhere(monitor, FUN)
```

Arguments

- `monitor` mts_monitor object.
- `FUN` A function applied to time series data that returns TRUE or FALSE.
**monitor_setTimeAxis**

**Value**

A subset of the incoming *mts_monitor* object. (A list with *meta* and *data* dataframes.)

**See Also**

monitor_select

**Examples**

```r
library(AirMonitor)

# Show all Camp_Fire locations
Camp_Fire$meta$locationName

# Use package US_AQI data for HAZARDOUS
name <- US_AQI$names_eng[6]
threshold <- US_AQI$breaks_PM2.5[6]

# Find HAZARDOUS locations
worst_sites <-
  Camp_Fire %>%
  monitor_selectWhere(
    function(x) { any(x >= threshold, na.rm = TRUE) }
  )

# Show the worst locations
worst_sites$meta$locationName
```

### **monitor_setTimeAxis**  

Extend/contract *mts_monitor* time series to new start and end times

**Description**

Extends or contracts the time range of an *mts_monitor* object by adding/removing time steps at the start and end and filling any new time steps with missing values. The resulting time axis is guaranteed to be a regular, hourly axis with no gaps using the same timezone as the incoming *mts_monitor* object. This is useful when you want to place separate *mts_monitor* objects on the same time axis for plotting.

If either *startdate* or *enddate* is missing, the start or end of the timeseries in *monitor* will be used.

**Usage**

```r
monitor_setTimeAxis(
  monitor = NULL,
  startdate = NULL,
  enddate = NULL,
  timezone = NULL
)
```
monitor_setTimeAxis

Arguments

- **monitor**: *mts_monitor* object.
- **startdate**: Desired start date (ISO 8601).
- **enddate**: Desired end date (ISO 8601).
- **timezone**: Olson timezone used to interpret *startdate* and *enddate*.

Value

The incoming *mts_monitor* time series object defined on a new time axis. (A list with meta and data dataframes.)

Note

If *startdate* or *enddate* is a POSIXct value, then timezone will be set to the timezone associated with *startdate* or *enddate*. In this common case, you don’t need to specify timezone explicitly.

If neither *startdate* nor *enddate* is a POSIXct value AND no timezone is supplied, the timezone will be inferred from the most common timezone found in *monitor*.

Examples

```r
library(AirMonitor)

# Default range
Carmel_Valley %>%
  monitor_timeRange()

# One-sided extend with user specified timezone
Carmel_Valley %>%
  monitor_setTimeAxis(enddate = 20160820, timezone = "UTC") %>%
  monitor_timeRange()

# Two-sided extend with user specified timezone
Carmel_Valley %>%
  monitor_setTimeAxis(20190720, 20190820, timezone = "UTC") %>%
  monitor_timeRange()

# Two-sided extend without timezone (uses monitor$meta$timezone)
Carmel_Valley %>%
  monitor_setTimeAxis(20190720, 20190820) %>%
  monitor_timeRange()
```
**monitor_slice_head**

Subset time series based on their position within an mts_monitor object

**Description**

An mts_monitor object is reduced so as to contain only the first or last n timeseries. These functions work similarly to dplyr::slice_head and dplyr::slice_tail but apply to both dataframes in the mts_monitor object.

This is primarily useful when the mts_monitor object has been ordered by a previous call to monitor_arrange or by some other means.

monitor_slice_head() selects the first and monitor_slice_tail() the last timeseries in the object.

**Usage**

```r
monitor_slice_head(monitor, n = 5)
monitor_slice_tail(monitor, n = 5)
```

**Arguments**

- `monitor`: mts_monitor object.
- `n`: Number of rows of monitor$meta to select.

**Value**

A subset of the incoming mts_monitor time series object. (A list with meta and data dataframes.)

**Examples**

```r
library(AirMonitor)

# Find lowest elevation sites
Camp_Fire %>%
  monitor_filter(!is.na(elevation)) %>%
  monitor_arrange(elevation) %>%
  monitor_slice_head(n = 5) %>%
  monitor_getMeta() %>%
  dplyr::select(elevation, locationName)

# Find highest elevation sites
Camp_Fire %>%
  monitor_filterMeta(!is.na(elevation)) %>%
  monitor_arrange(elevation) %>%
  monitor_slice_tail(n = 5) %>%
  monitor_getMeta() %>%
  dplyr::select(elevation, locationName)
```
monitor_timeInfo  

Get time related information for a monitor

Description

Calculate the local time for a monitor, as well as sunrise, sunset and solar noon times, and create several temporal masks.

The returned dataframe will have as many rows as the length of the incoming UTC time vector and will contain the following columns:

- `localStdTime_UTC` – UTC representation of local standard time
- `daylightSavings` – logical mask = TRUE if daylight savings is in effect
- `localTime` – local clock time
- `sunrise` – time of sunrise on each localTime day
- `sunset` – time of sunset on each localTime day
- `solarnoon` – time of solar noon on each localTime day
- `day` – logical mask = TRUE between sunrise and sunset
- `morning` – logical mask = TRUE between sunrise and solarnoon
- `afternoon` – logical mask = TRUE between solarnoon and sunset
- `night` – logical mask = opposite of day

Usage

monitor_timeInfo(monitor = NULL, id = NULL)

Arguments

monitor  
`mts_monitor` object.

id  
deviceDeploymentID used to select a single time series found in monitor. –
optional if monitor only has one time series.

Details

While the lubridate package makes it easy to work in local timezones, there is no easy way in R to work in "Local Standard Time" (LST) (i.e. never shifting to daylight savings) as is often required when working with air quality data. US EPA regulations mandate that daily averages be calculated based on LST.

The `localStdTime_UTC` is primarily for use internally and provides an important tool for creating LST daily averages and LST axis labeling.

Value

A dataframe with times and masks.
Examples

```r
library(AirMonitor)

carmel <-
  Carmel_Valley %>%
  monitor_filterDate(20160801, 20160810)

# Create timeInfo object for this monitor
ti <- monitor_timeInfo(carmel)

# Subset the data based on day/night masks
data_day <- carmel$data[ti$day,]
data_night <- carmel$data[ti$night,]

# Build two monitor objects
carmel_day <- list(meta = carmel$meta, data = data_day)
carmel_night <- list(meta = carmel$meta, data = data_night)

# Plot them
  carmel_day %>%
    monitor_timeseriesPlot(
      pch = 8,
      col = "goldenrod",
      shadedNight = TRUE
    )

  carmel_night %>%
    monitor_timeseriesPlot(
      add = TRUE,
      pch = 16,
      col = "darkblue"
    )
```

---

**monitor_timeRange**  
*Get the time range for a monitor*

**Description**

This function is a wrapper for `range(monitor$data$datetime)` and is convenient for use in data pipelines.

Dates will be returned in the timezone associated with `monitor$data$datetime` which is typically "UTC" unless timezone is specified.

**Usage**

```r
monitor_timeRange(monitor = NULL, timezone = NULL)
```
Arguments

monitor  
mts_monitor object.

timezone  
Olson timezone for the returned dates.

Value

A vector containing the minimum and maximum times of a mts_monitor object.

Examples

Carmel_Valley %>%
  monitor_timeRange(timezone = "America/Los_Angeles")

monitor_timeseriesPlot

Create timeseries plot

Description

Creates a time series plot of data from a mts_monitor object. By default, points are plotted as semi-transparent squares. All data values are plotted from all monitors found in the mts_monitor object.

Reasonable defaults are chosen for annotations and plot characteristics. Users can override any defaults by passing in parameters accepted by graphics::plot.default.

Usage

monitor_timeseriesPlot(
  monitor = NULL,
  id = NULL,
  shadedNight = FALSE,
  add = FALSE,
  addAQI = FALSE,
  palette = c("EPA", "subdued", "deuteranopia"),
  opacity = NULL,
  NAAQS = c("PM2.5", "PM2.5_2024"),
  ...
)

Arguments

monitor  
mts_monitor object.

id  
deviceDeploymentID used to limit plotting to a single time series found in monitor.

shadedNight  
Logical specifying whether to add nighttime shading.

add  
Logical specifying whether to add to the current plot.
monitor_timeseriesPlot

addAQI Logical specifying whether to add visual AQI decorations.
palette Named color palette to use when adding AQI decorations.
opacity Opacity to use for points. By default, an opacity is chosen based on the number of points so that trends are highlighted while outliers diminish in visual importance as the number of points increases.
NAAQS Version of NAAQS levels to use. See Note.
... Additional arguments to be passed to graphics::plot.default().

Value

No return value. This function is called to draw an air quality time series plot on the active graphics device.

Note

On February 7, 2024, EPA strengthened the National Ambient Air Quality Standards for Particulate Matter (PM NAAQS) to protect millions of Americans from harmful and costly health impacts, such as heart attacks and premature death. Particle or soot pollution is one of the most dangerous forms of air pollution, and an extensive body of science links it to a range of serious and sometimes deadly illnesses. EPA is setting the level of the primary (health-based) annual PM2.5 standard at 9.0 micrograms per cubic meter to provide increased public health protection, consistent with the available health science. See PM NAAQS update.

Examples

library(AirMonitor)

# Single monitor
Carmel_Valley %>%
  monitor_timeseriesPlot()

# Multiple monitors
Camp_Fire %>%
  monitor_filter(countyName == "Alameda") %>%
  monitor_timeseriesPlot(main = "All Alameda County Monitors")

# Standard extras
Carmel_Valley %>%
  monitor_timeseriesPlot(
    shadedNight = TRUE,
    addAQI = TRUE
  )
  addAQILegend()

# Standard extras using the updated PM NAAQS
Carmel_Valley %>%
  monitor_timeseriesPlot(
    shadedNight = TRUE,
    addAQI = TRUE,
    NAAQS = "PM2.5_2024"
  )
```r
# Fancy plot based on pm2.5 values
pm2.5 <- Carmel_Valley$data[,2]
Carmel_Valley %>%
  monitor_timeseriesPlot(
    shadedNight = TRUE,
    pch = 16,
    cex = pmax(pm2.5 / 100, 0.5),
    col = aqiColors(pm2.5),
    opacity = 0.8
  )
addAQILegend(pch = 16, cex = 0.6, bg = "white")
```

---

**monitor_toAQCTable**  
*Convert monitor data into an AQI category table*

**Description**

Creates a table of AQI category vs monitoring site with a count of the number of times each AQI category was experienced at each site. The count will be a count of hours or days depending on averaging period of the incoming monitor object.

When `siteIdentifier` is used, the identifiers must be in the same order as `monitor$meta`.

**Usage**

```r
monitor_toAQCTable(
  monitor,
  NAAQS = c("PM2.5", "PM2.5_2024"),
  siteIdentifier = "locationName"
)
```

**Arguments**

- `monitor`  
  *mts_monitor* object.
- `NAAQS`  
  Version of NAAQS levels to use. See Note.
- `siteIdentifier`  
  Metadata column used to identify sites or a character vector with site identifiers.

**Value**

Table of AQI category counts.
Note

On February 7, 2024, EPA strengthened the National Ambient Air Quality Standards for Particulate Matter (PM NAAQS) to protect millions of Americans from harmful and costly health impacts, such as heart attacks and premature death. Particle or soot pollution is one of the most dangerous forms of air pollution, and an extensive body of science links it to a range of serious and sometimes deadly illnesses. EPA is setting the level of the primary (health-based) annual PM2.5 standard at 9.0 micrograms per cubic meter to provide increased public health protection, consistent with the available health science. See PM NAAQS update.

Examples

library(AirMonitor)

# Lane County, Oregon AQSIDs all begin with "41039"
LaneCounty <-
  NW_Megafires %>%
  monitor_filter(stringr::str_detect(AQSID, '^41039')) %>%
  monitor_filterDate(20150801, 20150901)

# Count of hours each site spent in each AQ category in August
LaneCounty %>%
  monitor_toAQCTable()

# Count of days each site spent in each AQ
LaneCounty %>%
  monitor_dailyStatistic(mean) %>%
  monitor_toAQCTable()

# Count of days each site spent in each AQ (simplified names)
siteNames <- c(
  "Eugene 1", "Eugene 2", "Eugene 3",
  "Springfield", "Oakridge", "Cottage Grove"
)
LaneCounty %>%
  monitor_dailyStatistic(mean) %>%
  monitor_toAQCTable(siteIdentifier = siteNames)

# Count of days at each AQ level with the new, 2024 NAAQS
LaneCounty %>%
  monitor_dailyStatistic(mean) %>%
  monitor_toAQCTable(NAAQS = "PM2.5_2024")
monitor_toPWFSLSmoke

Description
Converts the contents of the monitor argument to CSV. By default, the output is a text string with "human readable" CSV that includes both meta and data. When saved as a file, this format is useful for point-and-click spreadsheet users who want to have everything on a single sheet.
To obtain a machine parseable CSV string for just the data, you can use includeMeta = FALSE. To obtain machine parseable metadata, use includeData = FALSE.

Usage
monitor_toCSV(monitor, includeMeta = TRUE, includeData = TRUE)

Arguments
- monitor: mts_monitor object.
- includeMeta: Logical specifying whether to include monitor$meta.
- includeData: Logical specifying whether to include monitor$data.

Value
CSV formatted text.

Examples
library(AirMonitor)

monitor <- Carmel_Valley %>%
  monitor_filterDate(20160802, 20160803)

monitor_toCSV(monitor) %>% cat()
monitor_toCSV(monitor, includeData = FALSE) %>% cat()
monitor_toCSV(monitor, includeMeta = FALSE) %>% cat()

monitor_toPWFSLSmoke

Convert a mts_monitor object to a ws_monitor object for the PWFSLSmoke package

Description
A mts_monitor object is modified so that it becomes a PWFSLSmoke package ws_monitor object.
While some information will be lost, this operation can be reversed with monitor_fromPWFSLSmoke().

Usage
monitor_toPWFSLSmoke(monitor = NULL)
monitor_trimDate

Arguments

monitor mts_monitor object

Value

A PWFSLSmoke ws_monitor object. (A list with meta and data dataframes.)

Note

In order to avoid duplicated monitorID values in the returned ws_monitor object, the full deviceDeploymentID will be used as the monitorID.

---

monitor_trimDate Trim a mts_monitor object to full days

Description

Trims the date range of a mts_monitor object to local time date boundaries which are within the range of data. This has the effect of removing partial-day data records at the start and end of the timeseries and is useful when calculating full-day statistics.

By default, multi-day periods of all-missing data at the beginning and end of the timeseries are removed before trimming to date boundaries. If trimEmptyDays = FALSE all records are retained except for partial days beyond the first and after the last date boundary.

Day boundaries are calculated using the specified timezone or, if NULL, from monitor$meta$timezone.

Usage

monitor_trimDate(monitor = NULL, timezone = NULL, trimEmptyDays = TRUE)

Arguments

monitor mts_monitor object.

timezone Olson timezone used to interpret dates.

trimEmptyDays Logical specifying whether to remove days with no data at the beginning and end of the time range.

Value

A subset of the given mts_monitor object. (A list with meta and data dataframes.)
Examples

library(AirMonitor)

# Non-day boundaries
monitor <-
  Camp_Fire %>%
  monitor_filterDatetime(
    "2018111502",
    "2018112206",
    timezone = "America/Los_Angeles"
  )

monitor %>%
  monitor_timeRange(timezone = "America/Los_Angeles")

# Trim to full days only
monitor %>%
  monitor_trimDate() %>%
  monitor_timeRange(timezone = "America/Los_Angeles")

NW_Megafires

NW_Megafires example dataset

Description

The NW_Megafires dataset provides a quickly loadable version of a mts_monitor object for practicing and code examples.

Usage

NW_Megafires

Format

A mts_monitor object with 1080 rows and 143 columns of data.

Details

In the summer of 2015, Washington state had several catastrophic wildfires that led to many days of heavy smoke in eastern Washington, Oregon and northern Idaho. The NW_Megafires dataset contains monitoring data for the Pacific Northwest from July 24 through September 06, 2015. This dataset was generated on 2022-10-28 by running:

library(AirMonitor)

NW_Megafires <-
  monitor_loadAnnual(2015, epaPreference = "epa_aqs")
 pollutantNames

monitor_filterMeta(stateCode
monitor_filterDate(20150724, 20150907, timezone = "America/Los_Angeles")
monitor_dropEmpty()

save(NW_Megafires, file = "data/NW_Megafires.rda")

| pollutantNames | Names of standard pollutants |

**Description**

Character string identifiers of recognized pollutant names.

**Usage**

pollutantNames

**Format**

A vector of character strings

**Details**

pollutantNames

**Examples**

print(coreMetadataNames, width = 80)

---

**QC_invalidateConsecutiveSuspectValues**

Invalidates values within a timeseries that appear "sticky". Some temporary monitoring data has stretches of consecutive values, sometimes well outside the range of reasonable. This QC function identifies these "sticky" stretches and returns the original timeseries data with "sticky" stretches replaced with NA.

**Usage**

QC_invalidateConsecutiveSuspectValues(
  x = NULL,
  suspectValues = c(0:10 * 1000, NA),
  consecutiveCount = 2
)
Arguments

x Timeseries data.
suspectValues Vector of numeric values considered suspect.
consecutiveCount How many suspectValues must appear in a row before they are invalidated.

Value

Returns x with some values potentially replaced with NA.

US_52

US state codes

Description

State codes for the 50 states +DC +PR (Puerto Rico).


Usage

US_52

Format

A vector with 52 elements

Details

US state codes

US_AQI

US EPA AQI Index levels, names, colors and action text

Description

Official, US EPA AQI levels, names, colors and action text are provided in a list for easy coloring and labeling.

Usage

US_AQI
Format

A list with named elements

Details

AQI breaks and associated names and colors

Breaks

Breakpoints are given in units reported for each parameter and include:

- breaks_AQI
- breaks_CO
- breaks_NO2
- breaks_OZONE_1hr
- breaks_OZONE_8hr
- breaks_PM2.5
- breaks_PM10

Colors

Several different color palettes are provided:

- colors_EPA – official EPA AQI colors
- colors_subdued – subdued colors fo use with leaflet maps
- colors_deuteranopia – color vision impaired colors

Names

Names of AQI categories are provided in several languages identified by the ISO 639-2 alpha-3 code:

- names_eng
- names_spa

Actions

Text for "actions to protect yourself" are provided for each category in several languages identified by the ISO 639-2 alpha-3 code:

- actions_eng
- actions_spa

Currently supported languages include English (eng) and Spanish (spa).


AQI colors are defined at [https://docs.airnowapi.org/aq101](https://docs.airnowapi.org/aq101)
wrcc_loadAnnual

Note

The low end of each break category is used as the breakpoint.

Examples

```r
print(US_AQI$breaks_AQI)
paste(US_AQI$colors_EPA)
paste(US_AQI$names_eng)
paste(US_AQI$names_spa)
```

Description

Loads pre-generated .rda files containing annual WRCC data. If archiveDataDir is defined, data will be loaded from this local archive. Otherwise, data will be loaded from the monitoring data repository maintained by the USFS AirFire team. Current year files loaded by this function are updated once per week. For the most recent data in the last 10 days, use wrcc_loadLatest(). For daily updates covering the most recent 45 days, use wrcc_loadDaily().

Usage

```r
wrcc_loadAnnual(
  year = NULL,
  archiveBaseUrl = paste0("https://airfire-data-exports.s3.us-west-2.amazonaws.com/",
    "monitoring/v2"),
  archiveBaseDir = NULL,
  QC_negativeValues = c("zero", "na", "ignore"),
  QC_removeSuspectData = TRUE
)
```

Arguments

- `year` Year [YYYY].
- `archiveBaseUrl` Base URL for monitoring v2 data files.
- `archiveBaseDir` Local base directory for monitoring v2 data files.
- `QC_negativeValues` Type of QC to apply to negative values.
- `QC_removeSuspectData` Removes monitors determined to be misbehaving.

Value

A `mts_monitor` object with WRCC data. (A list with meta and data dataframes.)
wrcc_loadDaily

Note

Some older WRCC timeseries contain only values of 0, 1000, 2000, 3000, ... ug/m3. Data from these deployments pass instrument-level QC checks but these timeseries generally do not represent valid data and should be removed. With QC_removeSuspectData = TRUE (the default), data is checked and periods reporting only values of 0:10 * 1000 ug/m3 are invalidated.

Only those personally familiar with the individual instrument deployments should work with the "suspect" data.

See Also

wrcc_loadDaily
wrcc_loadLatest

Examples

## Not run:
library(AirMonitor)

# Fail gracefully if any resources are not available
try({
# See https://en.wikipedia.org/wiki/Snake_River_Complex_Fire
# WRCC monitors during the Snake River Complex Fire
wrcc_loadAnnual(2021) \n  monitor_filter(stateCode \n    monitor_filterDate(20210707, 20210820, timezone = "America/Denver") \n    monitor_timeseriesPlot(
      ylim = c(0, 300),
      xpd = NA,
      addAQI = TRUE,
      main = "WRCC monitors during Snake River Complex Fire"
    )
  ), silent = FALSE)

## End(Not run)

wrcc_loadDaily  Load daily WRCC monitoring data

Description

Loads pre-generated .rda files containing daily WRCC data.

If archiveDataDir is defined, data will be loaded from this local archive. Otherwise, data will be loaded from the monitoring data repository maintained by the USFS AirFire team.

The files loaded by this function are updated once per day and contain data for the previous 45 days.
For the most recent data in the last 10 days, use `wrcc_loadLatest()`.
For data extended more than 45 days into the past, use `wrcc_loadAnnual()`.

**Usage**

```r
wrcc_loadDaily(
  archiveBaseUrl = paste0("https://airfire-data-exports.s3.us-west-2.amazonaws.com/", 
  "monitoring/v2"),
  archiveBaseDir = NULL,
  QC_negativeValues = c("zero", "na", "ignore"),
  QC_removeSuspectData = TRUE
)
```

**Arguments**

- `archiveBaseUrl`: Base URL for monitoring v2 data files.
- `archiveBaseDir`: Local base directory for monitoring v2 data files.
- `QC_negativeValues`: Type of QC to apply to negative values.
- `QC_removeSuspectData`: Removes monitors determined to be misbehaving.

**Value**

A `mts_monitor` object with WRCC data. (A list with `meta` and `data` dataframes.)

**Note**

Some older WRCC timeseries contain only values of 0, 1000, 2000, 3000, ... ug/m3. Data from these deployments pass instrument-level QC checks but these timeseries generally do not represent valid data and should be removed. With `QC_removeSuspectData = TRUE` (the default), data is checked and periods reporting only values of 0:10 * 1000 ug/m3 are invalidated.

Only those personally familiar with the individual instrument deployments should work with the "suspect" data.

**See Also**

- `wrcc_loadAnnual`
- `wrcc_loadDaily`

**Examples**

```r
## Not run:
library(AirMonitor)

# Fail gracefully if any resources are not available
try({
  wrcc_loadDaily()
})
```
Description

Loads pre-generated .rda files containing the most recent WRCC data.

If `archiveDataDir` is defined, data will be loaded from this local archive. Otherwise, data will be loaded from the monitoring data repository maintained by the USFS AirFire team.

The files loaded by this function are updated multiple times an hour and contain data for the previous 10 days.

For daily updates covering the most recent 45 days, use `wrcc_loadDaily()`.

For data extended more than 45 days into the past, use `wrcc_loadAnnual()`.

Usage

```r
wrcc_loadLatest(
  archiveBaseUrl = paste0("https://airfire-data-exports.s3.us-west-2.amazonaws.com/",
                         "monitoring/v2"),
  archiveBaseDir = NULL,
  QC_negativeValues = c("zero", "na", "ignore"),
  QC_removeSuspectData = TRUE
)
```

Arguments

- `archiveBaseUrl`: Base URL for monitoring v2 data files.
- `archiveBaseDir`: Local base directory for monitoring v2 data files.
- `QC_negativeValues`: Type of QC to apply to negative values.
- `QC_removeSuspectData`: Removes monitors determined to be misbehaving.

Value

A `mts_monitor` object with WRCC data. (A list with `meta` and `data` dataframes.)
Note

Some older WRCC timeseries contain only values of 0, 1000, 2000, 3000, ... ug/m3. Data from these deployments pass instrument-level QC checks but these timeseries generally do not represent valid data and should be removed. With `QC_removeSuspectData = TRUE` (the default), data is checked and periods reporting only values of 0:10 * 1000 ug/m3 are invalidated.

Only those personally familiar with the individual instrument deployments should work with the "suspect" data.

See Also

`wrcc_loadAnnual`

`wrcc_loadDaily`

Examples

```r
## Not run:
library(AirMonitor)

# Fail gracefully if any resources are not available
try({
    wrcc_loadLatest() \n    monitor_leaflet()
}, silent = FALSE)

## End(Not run)
```
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