R Workshop #1

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Subsetting Data

Dealing with Missing Data

Managing Your Workflow with `R Projects`

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Subsetting Data

Subsetting Data: Matrix Analogy

- `data.frame` is a type of `matrix`: each cell is indexed by its `[row #, column #]`

```
m<-matrix(c("a","b","c","d","e","f"),nrow=2)
m
```

```
##      [,1] [,2] [,3]
## [1,] "a"  "c"  "e"
## [2,] "b"  "d"  "f"
```

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##      [,1] [,2] [,3]
## [1,] "a"  "c"  "e"
## [2,] "b"  "d"  "f"
```

- Subset a **specific row**:

```
m[2,]
```

```
## [1] "b" "d" "f"
```

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##      [,1] [,2] [,3]
## [1,] "a"  "c"  "e"
## [2,] "b"  "d"  "f"
```

- Subset a **specific row**:

```
m[2,]
```

```
## [1] "b" "d" "f"
```

- Subset a **specific column**:

```
m[,3]
```

```
## [1] "e" "f"
```

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Subsetting Data: Matrix Analogy

- `data.frame` is a type of `matrix`: each cell is indexed by its `[row #, column #]`

```
m<-matrix(c("a","b","c","d","e","f"),nrow=2)
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##      [,1] [,2] [,3]
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## [2,] "b"  "d"  "f"
```

- Subset a **specific row**:

```
m[2,]
```

```
## [1] "b" "d" "f"
```

- Subset a **specific column**:

```
m[,3]
```

```
## [1] "e" "f"
```

- Subset a **specific element**:

```
m[2,3]
```

```
## [1] "f"
```

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Also see the **dplyr** package.

Data Frames

```
df <- data.frame(x = 1:3, y = c('a', 'b', 'c'))
```

A special case of a list where all elements are the same length.

x	y
1	a

List subsetting

`df$x` 

`df[[2]]` 

- We can do the same thing for `data.frame`s:

```
df<-data.frame(Nums=c(1,2,3,4,5),
               Lets=c("a","b","c","d","e"))
df
```

```
##   Nums Lets
## 1    1    a
## 2    2    b
## 3    3    c
## 4    4    d
## 5    5    e
```

Subsetting Data: Data Frames

- Can also subset a `data.frame` by position:

```
df
```

```
##   Nums Lets
## 1    1    a
## 2    2    b
## 3    3    c
## 4    4    d
## 5    5    e
```

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- Can also subset a `data.frame` by position:

```
df
```

```
##   Nums Lets
## 1    1    a
## 2    2    b
## 3    3    c
## 4    4    d
## 5    5    e
```

- Subset a **specific row (observation)**:

```
df[2,]
```

```
##   Nums Lets
## 2    2    b
```

- Can also subset a `data.frame` by position:

```
df
```

```
##   Nums Lets
## 1    1    a
## 2    2    b
## 3    3    c
## 4    4    d
## 5    5    e
```

- Subset a **specific row (observation)**:

```
df[2,]
```

```
##   Nums Lets
## 2    2    b
```

- Subset a **specific column (variable)**:

```
df[,2]
```

```
## [1] "a" "b" "c" "d" "e"
```

- Can also subset a `data.frame` by position:

```
df
```

```
##   Nums Lets
## 1    1    a
## 2    2    b
## 3    3    c
## 4    4    d
## 5    5    e
```

- Subset a **specific row (observation)**:

```
df[2,]
```

```
##   Nums Lets
## 2    2    b
```

- Subset a **specific column (variable)**:

```
df[,2]
```

```
## [1] "a" "b" "c" "d" "e"
```

- Subset a **specific value**:

```
df[2,2]
```

```
## [1] "b"
```

- The nice thing about data frames is that instead of remembering the order of columns, we have the `names` of columns

```
df
```

```
##   Nums Lets
## 1    1    a
## 2    2    b
## 3    3    c
## 4    4    d
## 5    5    e
```

```
names(df)
```

```
## [1] "Nums" "Lets"
```

- The nice thing about data frames is that instead of remembering the order of columns, we have the `names` of columns

```
df
```

```
##   Nums Lets
## 1    1    a
## 2    2    b
## 3    3    c
## 4    4    d
## 5    5    e
```

```
names(df)
```

```
## [1] "Nums" "Lets"
```

Subsetting Conditionally

- We often want to subset a `data.frame` based on a condition

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- Can use original brackets `[]` to pick by **rows** (observations) for which `Num>2`

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Subsetting Conditionally

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  - e.g. look only at **observations for which Nums are larger than 2**
- Can use original brackets `[]` to pick by **rows** (observations) for which `Num>2`

- If we want **all** columns (variables)

```
df[df$Nums>2,]
```

```
##   Nums Lets
## 3    3    c
## 4    4    d
## 5    5    e
```

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Subsetting Conditionally

- We often want to subset a `data.frame` based on a condition
  - e.g. look only at **observations for which Nums are larger than 2**
- Can use original brackets `[]` to pick by **rows** (observations) for which `Num>2`

- If we want **all** columns (variables)

```
df[df$Nums>2,]
```

```
##   Nums Lets
## 3    3    c
## 4    4    d
## 5    5    e
```

- If we only want column 1 (”Nums”)

```
df[df$Nums>2,1]
```

```
## [1] 3 4 5
```

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Subsetting Conditionally

- We often want to subset a `data.frame` based on a condition
  - e.g. look only at **observations for which Nums are larger than 2**
- Can use original brackets `[]` to pick by **rows** (observations) for which `Num>2`

- If we want **all** columns (variables)

```
df[df$Nums>2,]
```

```
##   Nums Lets
## 3    3    c
## 4    4    d
## 5    5    e
```

- If we only want column 1 (`"Nums"`)

```
df[df$Nums>2,1]
```

```
## [1] 3 4 5
```

- If we only want column 2 (`"Lets"`)

```
df[df$Nums>2,2]
```

```
## [1] "c" "d" "e"
```

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Subsetting Conditionally: `subset()`

- One faster way that gets us away from `[]` is `subset(df, condition)`

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  - Keeps only values of `df` for which condition is `TRUE`

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Subsetting Conditionally: `subset()`

- One faster way that gets us away from `[]` is `subset(df, condition)`
  - Keeps only values of `df` for which condition is `TRUE`

```
subset(df, Nums>2)
```

```
##   Nums Lets
## 3    3    c
## 4    4    d
## 5    5    e
```

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Subsetting Conditionally: `filter()` with `dplyr`

- `dplyr` makes this easier with `filter()`

```
df %>%
  filter(Nums>2)
```

```
##   Nums Lets
## 1    3    c
## 2    4    d
## 3    5    e
```

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Useful Conditionals

Condition	Description	Example(s)
`>`	Values greater than	`Num>2`
`>=`	Values greater than or equal to	`Num>=2`
`==`	Values equal to (put value in quotes if a character)	`Num==2`; `Let=="a"`
`!=`	Values are NOT equal to	`Num!=2`; `Let!="a"`
`cond.1 & cond.2`	"AND": BOTH conditions must be met	`Num>2 & Num<5`
`cond.1 | cond.2`	"OR": Either one condition must be met	`Num>2 | Num<5`
`%in% c()`	Values are in a set of values defined in `c()`	`Num %in% c(1,2,3)`
`!%in% c()`	Values are NOT in defined set	`Num !%in% c(1,2,3)`

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Dealing with Missing Data

Missing Data: NA

- If any observation is missing a value of a variable, it will show up as NA

```
x<-c(1,2,NA,4,5)
y<-c("a",NA,"c","d","e")
df<-data.frame(x,y)

df
```

```
##    x    y
## 1  1    a
## 2  2 <NA>
## 3 NA    c
## 4  4    d
## 5  5    e
```

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NAs Propagate...

- Missing data propagates and will ruin many functions you run on it

```
mean(df$x)
```

```
## [1] NA
```

```
sd(df$x)
```

```
## [1] NA
```

```
sum(df$x)
```

```
## [1] NA
```

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- Several strategies to combat NAs

```
# with base R

df1<-df[!is.na(df$x),] # drop all observations for which there is NA for x
df1
```

```
##   x    y
## 1 1    a
## 2 2 <NA>
## 4 4    d
## 5 5    e
```

NA Strategy

- Several strategies to combat NAs

1. If looking at one variable:

```
# with base R

df1<-df[!is.na(df$x),] # drop all observations for which there is NA for x
df1
```

```
##   x    y
## 1 1    a
## 2 2 <NA>
## 4 4    d
## 5 5    e
```

NA Strategy

- Several strategies to combat NAs

1. If looking at one variable:
    - Keep only observations for which there are no NAs

```
# with base R

df1<-df[!is.na(df$x),] # drop all observations for which there is NA for x
df1
```

```
##   x    y
## 1 1    a
## 2 2 <NA>
## 4 4    d
## 5 5    e
```

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NA Strategy

- Several strategies to combat NAs

1. If looking at one variable:
    - Keep only observations for which there are no NAs

```
# with base R

df1<-df[!is.na(df$x),] # drop all observations for which there is NA for x
df1
```

```
##   x    y
## 1 1    a
## 2 2 <NA>
## 4 4    d
## 5 5    e
```

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2. Drop *all* observations that have some missing value across *any* variable with `na.omit(df)`

```
df2<-na.omit(df) # drop any row that has any NA value for any variable
df2
```

```
##   x y
## 1 1 a
## 4 4 d
## 5 5 e
```

2. Drop *all* observations that have some missing value across *any* variable with `na.omit(df)`
    - Often too extreme, may end up throwing out a lot of useful data!

```
df2<-na.omit(df) # drop any row that has any NA value for any variable
df2
```

```
##   x y
## 1 1 a
## 4 4 d
## 5 5 e
```

NA Strategy III

3. Most functions have a NA option built in

```
mean(df$x, na.rm=TRUE)
```

```
## [1] 3
```

```
sd(df$x, na.rm=TRUE)
```

```
## [1] 1.825742
```

```
sum(df$x, na.rm=TRUE)
```

```
## [1] 12
```

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NA Strategy III

3. Most functions have a NA option built in
    - Add "`, na.rm=TRUE`" inside any function's `()` to simply *ignore* all observations with NAs

```
mean(df$x, na.rm=TRUE)
```

```
## [1] 3
```

```
sd(df$x, na.rm=TRUE)
```

```
## [1] 1.825742
```

```
sum(df$x, na.rm=TRUE)
```

```
## [1] 12
```

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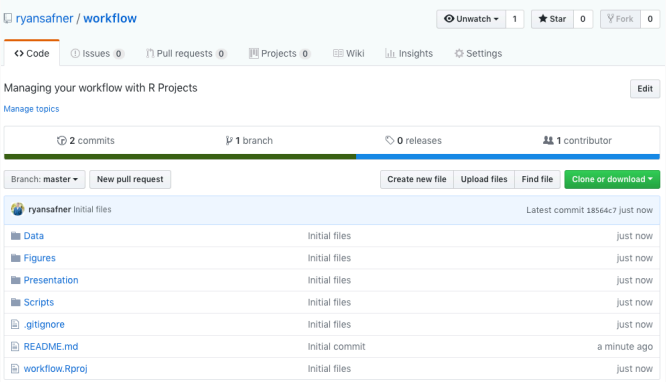
Managing Your Workflow with R Projects

Using RProj Projects



Go to github.com/ryansafner/workflow and follow the instructions!