### APPLICATION OF A FUNCTION ACROSS MULTIPLE VARIABLES

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# Let's start by creating a simple data frame

#### First, let's use sample() to create three numeric variables for our data frame:

### Then, we'll add a factor to create subgroups for analysis:

```
> dataX$grp = factor(rep(c("Grp1", "Grp2"),50))
```

#### Finally, we'll add some missingness to our data:

```
> mpat = sample(100,100,replace=TRUE)
> dataX$V2[which(mpat<25 & grp=="Grp1")] = NA
> dataX$V2[which(mpat<15 & grp=="Grp2")] = NA</pre>
```

Our sample data frame has four variables, three numeric and one factor with two levels. Moreover, one of the numeric variables has missing values, and there is more missingness corresponding to one of the two factor levels.

## A look at the sample data frame

```
> head(dataX)
 V1 V2 V3 grp
1 48 775 3 Grp1
2 88 314 7 Grp2
3 84 671 4 Grp1
4 49 706 8 Grp2
5 13 NA 2 Grp1
6 85 916 1 Grp2
> table(is.na(dataX$V2),dataX$grp)
       Grp1 Grp2
         38 46
 FALSE
      12
 TRUE
```

### Number of valid cases for all variables

```
It's easy to get the number of missing values for one variable:
sum(is.na(dataX$V2)) or even table(is.na(dataX$V2))
And thanks to colSums(), you don't need a loop for a whole set of variables!
X = matrix(dataX$V2,nrow=10)
colSums(is.na(X))
Now, let's write a function for that...
nvalid = function(x)  {
  invisible(colSums(!is.na(x)))
"invisible" suppresses automatic output of the evaluation of the colSums expression.
> print(nvalid(dataX))
    V2 V3 grp
100 84 100 100
```

# colSums() versus apply()

```
apply() can be used instead of colSums():
    apply(is.na(X),2,sum))
gives the same result as
    colSums(is.na(X))
but colSums() is both easier to read and more efficient.
There are other functions like colSums():
    rowSums(), colMeans(), and rowMeans()
Any that I don't know about?
In other situations, use apply().
And of course, there's Hadley Wickham's plyr package.
To paraphrase many wise speakers who've come before me:
```

### "DON'T USE LOOPS!"



## Missing values by subgroup

We can use by() to apply our function to slices of the sample data set:

```
by (dataX, dataX$grp, nvalid))
dataX$grp: Grp1
V1 V2 V3 grp
 50 38 50 50
dataX$grp: Grp2
V1 V2 V3 grp
 50 46 50 50
And the result can be cleaned up with sapply:
sapply(by(dataX, dataX$grp, nvalid)), identity)
        Grp1 Grp2
  V1 50 50
```

V2 38 46

V3 50 50

grp 50 50

### Turning it all into a function...

```
grpsum = function(mydata, grpvar) {
 byNs = by(mydata, grp, nvalid)
 Ns = sapply(byNs,identity)
 vars = names(mydata)
 result = as.data.frame(list(Var=vars, N=Ns))
 row.names(result) = NULL
 invisible (result)
> print(grpsum(dataX[,1:3],dataX$grp))
  Var N.Grp1 N.Grp2
 V1 50 50
2 V2 38 46
3 V3 50 50
```

### You can extend this to more complex functions

Here, we used by and sapply to apply our simple function nvalid to multiple variables and subgroups within our data frame.

You can use the same approach for simple built-in functions, such as mean or sd, or indeed more complex functions that you create.

| Var | Label                   | Avg.1 | Avg.2 | Avg.3 | N.1 | N.2 | N.3 | ANOVA.F | ANOVA.p | Flags       |
|-----|-------------------------|-------|-------|-------|-----|-----|-----|---------|---------|-------------|
| A1  | Multispecialty practice | 0.53  | 0.43  | 0.41  | 131 | 69  | 86  | 3.2     | 0.1     | 3,1         |
| Px  | Procedure Volume        | 30.44 | 86.94 | 1.44  | 131 | 69  | 86  | 7.3     | 0.0     | 2,1 3,1 3,2 |
| B1  | Academic Center         | 0.08  | 0.12  | 0.06  | 131 | 69  | 86  | 0.1     | 0.7     |             |
| B2  | Community Hospital      | 0.11  | 0.17  | 0.02  | 131 | 69  | 86  | 3.1     | 0.1     | 3,1 3,2     |
| В3  | Private Office          | 0.48  | 0.49  | 0.57  | 131 | 69  | 86  | 1.5     | 0.2     |             |
|     |                         |       |       |       |     |     |     |         |         |             |