Standardizing Data

Here are a few different ways we can scale the data, the first is using a loop, which are incredibly versatile BUT kinda tricky to wrap your head around. Here is more info on loops that is a good reference

Loop

First, lets read in the data:

```
dat<- read.csv(
  here("Data", "Yellowtail", "2024Env-annual-yellowtail_GLORYS_UNSTANDARDIZED.csv")
)%>%
  filter(year>=1998&year<=2021) # lets filter the data so we dont have any missing values
```

Now, lets create a dataframe. We will "fill" this dataframe using a loop

```
stand_dat <-dat #using the same shape as dat will also let us retain the column names
```

Now we can create the actual loop:

```
#i is what we want the loop to iterate over, in this case i is columns. If we want the
# loop to loop over every column, we can start it at 1 and go to the total number
# of columns. ncol() is a function that counts the number of columns in a dataframe
# so in this case we will loop from column 1 to the last column

for(i in 1:ncol(dat)){
 stand_dat[,i]<- (dat[,i]-mean(dat[,i]))/sd(dat[,i])
}
```

Our stand_dat dataframe now contains each column with a mean of 0 and standard deviation of 1. Lets check it...

```r
colMeans(stand_dat)
```

```
         year         DDpre         DDegg        DDlarv        DDpjuv
-9.251859e-18  8.367150e-16 -2.151057e-16  2.035409e-16  9.390636e-16
        DDben          Tcop         Tpart       MLDpart       MLDlarv
-1.292947e-15  1.327642e-15  8.835525e-16  3.238150e-16 -3.469447e-16
      MLDpjuv       CSTlarv       CSTpjuv       LSTlarv       LSTpjuv
 8.974303e-16  3.700743e-17  1.850372e-17  8.789266e-17 -2.220446e-16
    hci1_larv     hci1_pjuv     hci2_larv     hci2_pjuv       oni_pre
 0.000000e+00 -1.249001e-16  0.000000e+00 -1.249001e-16 -1.850372e-17
     oni_larv      oni_pjuv      pdo_larv      pdo_pjuv   lusi_annual
-1.850372e-17  2.312965e-17  7.401487e-17 -2.775558e-17 -3.654484e-16
     ZOOpjuvN       ZOObenN      ZOOpjuvS       ZOObenS             X
 0.000000e+00 -6.013708e-17  3.238150e-17 -6.245005e-17 -9.251859e-18
  CutiSTIpjuv  CutiTUMIpjuv          Year BeutiTUMIpjuv        Year.1
 2.312965e-17  2.914335e-16 -9.251859e-18 -7.401487e-17 -9.251859e-18
 BeutiSTIpjuv       CHLpjuv        PPpjuv     bakun_sti
-1.989150e-16  6.476301e-17  8.396062e-16 -1.850372e-17
```

```r
apply(stand_dat, 2, sd)
```

```
         year         DDpre         DDegg        DDlarv        DDpjuv
            1             1             1             1             1
        DDben          Tcop         Tpart       MLDpart       MLDlarv
            1             1             1             1             1
      MLDpjuv       CSTlarv       CSTpjuv       LSTlarv       LSTpjuv
            1             1             1             1             1
    hci1_larv     hci1_pjuv     hci2_larv     hci2_pjuv       oni_pre
            1             1             1             1             1
     oni_larv      oni_pjuv      pdo_larv      pdo_pjuv   lusi_annual
            1             1             1             1             1
     ZOOpjuvN       ZOObenN      ZOOpjuvS       ZOObenS             X
            1             1             1             1             1
  CutiSTIpjuv  CutiTUMIpjuv          Year BeutiTUMIpjuv        Year.1
            1             1             1             1             1
 BeutiSTIpjuv       CHLpjuv        PPpjuv     bakun_sti
            1             1             1             1
```

The means of each column is pretty close to 0! And the sd is as well!

Apply

Now you may have noticed that when calculating the sd of each function I used apply. Rather than writing a loop we could have just used this! Lets use it here to standardize the data...

```
stand_data_apply <- apply(dat, 2, scale)
head(stand_data_apply )
```

```
           year       DDpre      DDegg     DDlarv      DDpjuv      DDben
[1,] -1.6263456  2.79432823  2.0652425  2.2468318  1.97182592  0.3776698
[2,] -1.4849242  0.03494097  0.2375537 -0.7622902 -0.91217826 -0.4744541
[3,] -1.3435029 -0.31874437 -0.1290300  0.1215045 -0.03340699 -0.9624530
[4,] -1.2020815 -0.67892077 -0.9378641 -0.4239759 -0.73198157 -0.6951369
[5,] -1.0606602 -0.79438257 -0.4608233 -0.9887974 -1.35184416 -0.6292660
[6,] -0.9192388 -0.23015000  0.2453647  0.8054927  0.23617465 -0.2484564
           Tcop       Tpart     MLDpart     MLDlarv    MLDpjuv     CSTlarv
[1,]  2.6666699  2.19889421  3.82323623  3.39269147  0.1485115 -0.57884030
[2,]  0.4379791 -0.85057733  1.06985511  2.11898054  1.0200999  0.90561290
[3,] -0.5602335 -0.05980499 -0.51832950 -0.18220965 -1.0067481  0.15083688
[4,] -0.5305484 -0.42450685 -0.23769989 -0.53042904  0.2297167 -0.44020836
[5,] -0.6697892 -0.85819721 -0.38622627 -0.80924151  0.1426017  0.07036301
[6,] -1.9395112  0.97069017 -0.01705528 -0.04951702 -0.3553550  0.60915913
         CSTpjuv    LSTlarv    LSTpjuv   hci1_larv   hci1_pjuv   hci2_larv
[1,] -0.03802977  1.7264370  0.4889551 -1.56798699 -1.27115913 -1.56798699
[2,]  0.94350146  1.4760473 -1.2856393  1.25559625  1.19972209  1.25559625
[3,]  1.19176038  0.2180359 -0.7130612  0.09881201  0.06562815  0.09881201
[4,]  0.38474393 -0.6262714 -0.5019289  0.39967244  1.13446472  0.39967244
[5,]  0.53120956 -0.9255748 -1.3013491  1.07861953  1.10480228  1.07861953
[6,] -0.10811819 -0.6993861 -0.1147803 -1.38296587 -0.50883443 -1.38296587
       hci2_pjuv    oni_pre   oni_larv    oni_pjuv   pdo_larv   pdo_pjuv
[1,] -1.27115913  2.1429247  2.0090279 -0.05089855  1.9366648  0.3970298
[2,]  1.19972209 -1.2653885 -1.2340431 -1.98712104 -0.5796399 -1.4867611
[3,]  0.06562815 -1.3052852 -1.2961926 -1.14365931 -0.5104292 -0.5346554
[4,]  1.13446472 -0.5632076 -0.4374003 -0.18801302  0.2904379 -0.8282001
[5,]  1.10480228 -0.1083859  0.1727943  1.41165577 -0.5697527 -0.2309005
[6,] -0.50883443  0.9893417  0.6756398  0.12361077  1.5461185  0.9994344
     lusi_annual     ZOOpjuvN    ZOObenN    ZOOpjuvS    ZOObenS          X
[1,]   1.0055400 -1.952427025 -0.3585604  1.7126823  0.7954300 -1.6263456
[2,]   1.5213590 -0.169297841  0.7000897 -0.7176185 -1.6095081 -1.4849242
[3,]  -0.4682286 -0.004760062  1.2082420 -0.9834969 -0.9465854 -1.3435029
[4,]   0.3791884  0.235604522  1.3160679 -0.9022688 -1.0887306 -1.2020815
[5,]   1.2266053  0.556019342  1.3121875 -0.8130939 -0.5867960 -1.0606602
```

```
[6,]  -0.3945401 -0.067985973 -0.4133460  0.5036207 -0.3517562 -0.9192388
     CutiSTIpjuv CutiTUMIpjuv       Year BeutiTUMIpjuv     Year.1 BeutiSTIpjuv
[1,]  -0.1918014   -0.4939859 -1.6263456  -1.568685313 -1.6263456   -0.1790974
[2,]   0.3504340    0.1430873 -1.4849242   0.001992683 -1.4849242    0.4387064
[3,]  -0.7109629   -0.6343623 -1.3435029  -1.099932577 -1.3435029   -0.7706116
[4,]  -1.2993460    0.9778535 -1.2020815   0.712209704 -1.2020815   -1.4409943
[5,]  -0.9878491    1.5685664 -1.0606602   1.339731782 -1.0606602   -0.6785982
[6,]   0.9157433    0.3323588 -0.9192388  -1.220105642 -0.9192388    1.0827996
        CHLpjuv     PPpjuv  bakun_sti
[1,] -1.9364182 -2.0454212 -1.5268571
[2,] -1.4005172 -1.0273683 -1.2167479
[3,] -0.7564655 -0.6926709  1.1090711
[4,] -0.8155079 -0.6337877  0.3854830
[5,]  2.2314761  2.3045496 -0.3381052
[6,]  0.6224340  0.9496238  0.2304284
```

Let's check, does the data from apply look the same as the data from the loop? You can use apply to also run a function across rows.

```
diff<-stand_data_apply - stand_dat
head(diff)
```

```
  year DDpre        DDegg        DDlarv DDpjuv        DDben          Tcop Tpart
1    0     0 1.332268e-15 -1.332268e-15      0 2.609024e-15 -2.664535e-15     0
2    0     0 1.498801e-15 -6.661338e-16      0 2.498002e-15 -2.664535e-15     0
3    0     0 1.498801e-15 -8.465451e-16      0 2.331468e-15 -2.664535e-15     0
4    0     0 1.554312e-15 -7.216450e-16      0 2.442491e-15 -2.664535e-15     0
5    0     0 1.498801e-15 -6.661338e-16      0 2.442491e-15 -2.664535e-15     0
6    0     0 1.498801e-15 -9.992007e-16      0 2.525757e-15 -2.664535e-15     0
       MLDpart      MLDlarv      MLDpjuv       CSTlarv       CSTpjuv
1 1.776357e-15 2.220446e-15 7.133183e-15 -5.551115e-16 -1.318390e-16
2 1.110223e-15 2.220446e-15 7.105427e-15 -4.440892e-16 -1.110223e-16
3 7.771561e-16 2.775558e-15 7.105427e-15 -5.273559e-16 -2.220446e-16
4 8.049117e-16 2.886580e-15 7.105427e-15 -5.551115e-16 -1.110223e-16
5 7.216450e-16 2.886580e-15 7.105427e-15 -5.134781e-16 -1.110223e-16
6 8.326673e-16 2.754741e-15 7.160939e-15 -5.551115e-16 -1.387779e-16
        LSTlarv      LSTpjuv hci1_larv hci1_pjuv hci2_larv hci2_pjuv
1 -4.440892e-16 7.216450e-16         0         0         0         0
2 -4.440892e-16 4.440892e-16         0         0         0         0
3 -2.498002e-16 5.551115e-16         0         0         0         0
4 -2.220446e-16 5.551115e-16         0         0         0         0
5 -1.110223e-16 4.440892e-16         0         0         0         0
```

```
6 -2.220446e-16 6.245005e-16         0         0         0         0
        oni_pre     oni_larv      oni_pjuv      pdo_larv     pdo_pjuv
1  0.000000e+00 0.000000e+00  9.714451e-17  2.220446e-16 5.551115e-17
2  0.000000e+00 0.000000e+00  2.220446e-16 -3.330669e-16 2.220446e-16
3  0.000000e+00 0.000000e+00  0.000000e+00 -2.220446e-16 0.000000e+00
4  0.000000e+00 5.551115e-17  1.110223e-16 -5.551115e-17 0.000000e+00
5 -1.387779e-17 8.326673e-17 -2.220446e-16 -2.220446e-16 8.326673e-17
6  0.000000e+00 0.000000e+00  6.938894e-17  2.220446e-16 0.000000e+00
  lusi_annual     ZOOpjuvN ZOObenN ZOOpjuvS      ZOObenS X  CutiSTIpjuv
1           0 0.000000e+00       0        0 1.110223e-16 0 4.718448e-16
2           0 2.775558e-17       0        0 2.220446e-16 0 4.996004e-16
3           0 1.734723e-17       0        0 1.110223e-16 0 3.330669e-16
4           0 5.551115e-17       0        0 0.000000e+00 0 4.440892e-16
5           0 0.000000e+00       0        0 1.110223e-16 0 3.330669e-16
6           0 2.775558e-17       0        0 5.551115e-17 0 5.551115e-16
  CutiTUMIpjuv Year BeutiTUMIpjuv Year.1 BeutiSTIpjuv       CHLpjuv PPpjuv
1 6.106227e-16    0 -1.110223e-15      0 5.273559e-16 -1.332268e-15      0
2 6.106227e-16    0 -1.004839e-15      0 6.661338e-16 -1.110223e-15      0
3 6.661338e-16    0 -8.881784e-16      0 4.440892e-16 -1.110223e-15      0
4 5.551115e-16    0 -9.992007e-16      0 2.220446e-16 -1.221245e-15      0
5 6.661338e-16    0 -8.881784e-16      0 4.440892e-16 -1.332268e-15      0
6 5.551115e-16    0 -8.881784e-16      0 6.661338e-16 -1.221245e-15      0
      bakun_sti
1  0.000000e+00
2  0.000000e+00
3  0.000000e+00
4  5.551115e-17
5 -5.551115e-17
6  8.326673e-17
```

It sure does! The values are identical, so we know both work

Scale

Now a useful thing about scale is that you actually don't need to loop or apply because that functionality is built into it! We can scale columns using the function directly

```
scale_dat_scale <- scale(dat)
head(scale_dat_scale)
```

```
          year       DDpre      DDegg     DDlarv      DDpjuv      DDben
```

```
[1,] -1.6263456  2.79432823  2.0652425  2.2468318  1.97182592  0.3776698
[2,] -1.4849242  0.03494097  0.2375537 -0.7622902 -0.91217826 -0.4744541
[3,] -1.3435029 -0.31874437 -0.1290300  0.1215045 -0.03340699 -0.9624530
[4,] -1.2020815 -0.67892077 -0.9378641 -0.4239759 -0.73198157 -0.6951369
[5,] -1.0606602 -0.79438257 -0.4608233 -0.9887974 -1.35184416 -0.6292660
[6,] -0.9192388 -0.23015000  0.2453647  0.8054927  0.23617465 -0.2484564
          Tcop       Tpart     MLDpart     MLDlarv    MLDpjuv     CSTlarv
[1,]  2.6666699  2.19889421  3.82323623  3.39269147  0.1485115 -0.57884030
[2,]  0.4379791 -0.85057733  1.06985511  2.11898054  1.0200999  0.90561290
[3,] -0.5602335 -0.05980499 -0.51832950 -0.18220965 -1.0067481  0.15083688
[4,] -0.5305484 -0.42450685 -0.23769989 -0.53042904  0.2297167 -0.44020836
[5,] -0.6697892 -0.85819721 -0.38622627 -0.80924151  0.1426017  0.07036301
[6,] -1.9395112  0.97069017 -0.01705528 -0.04951702 -0.3553550  0.60915913
        CSTpjuv    LSTlarv    LSTpjuv   hci1_larv   hci1_pjuv   hci2_larv
[1,] -0.03802977  1.7264370  0.4889551 -1.56798699 -1.27115913 -1.56798699
[2,]  0.94350146  1.4760473 -1.2856393  1.25559625  1.19972209  1.25559625
[3,]  1.19176038  0.2180359 -0.7130612  0.09881201  0.06562815  0.09881201
[4,]  0.38474393 -0.6262714 -0.5019289  0.39967244  1.13446472  0.39967244
[5,]  0.53120956 -0.9255748 -1.3013491  1.07861953  1.10480228  1.07861953
[6,] -0.10811819 -0.6993861 -0.1147803 -1.38296587 -0.50883443 -1.38296587
      hci2_pjuv    oni_pre   oni_larv    oni_pjuv   pdo_larv   pdo_pjuv
[1,] -1.27115913  2.1429247  2.0090279 -0.05089855  1.9366648  0.3970298
[2,]  1.19972209 -1.2653885 -1.2340431 -1.98712104 -0.5796399 -1.4867611
[3,]  0.06562815 -1.3052852 -1.2961926 -1.14365931 -0.5104292 -0.5346554
[4,]  1.13446472 -0.5632076 -0.4374003 -0.18801302  0.2904379 -0.8282001
[5,]  1.10480228 -0.1083859  0.1727943  1.41165577 -0.5697527 -0.2309005
[6,] -0.50883443  0.9893417  0.6756398  0.12361077  1.5461185  0.9994344
     lusi_annual     ZOOpjuvN    ZOObenN   ZOOpjuvS    ZOObenS          X
[1,]   1.0055400 -1.952427025 -0.3585604  1.7126823  0.7954300 -1.6263456
[2,]   1.5213590 -0.169297841  0.7000897 -0.7176185 -1.6095081 -1.4849242
[3,]  -0.4682286 -0.004760062  1.2082420 -0.9834969 -0.9465854 -1.3435029
[4,]   0.3791884  0.235604522  1.3160679 -0.9022688 -1.0887306 -1.2020815
[5,]   1.2266053  0.556019342  1.3121875 -0.8130939 -0.5867960 -1.0606602
[6,]  -0.3945401 -0.067985973 -0.4133460  0.5036207 -0.3517562 -0.9192388
     CutiSTIpjuv CutiTUMIpjuv       Year BeutiTUMIpjuv     Year.1 BeutiSTIpjuv
[1,]  -0.1918014   -0.4939859 -1.6263456  -1.568685313 -1.6263456   -0.1790974
[2,]   0.3504340    0.1430873 -1.4849242   0.001992683 -1.4849242    0.4387064
[3,]  -0.7109629   -0.6343623 -1.3435029  -1.099932577 -1.3435029   -0.7706116
[4,]  -1.2993460    0.9778535 -1.2020815   0.712209704 -1.2020815   -1.4409943
[5,]  -0.9878491    1.5685664 -1.0606602   1.339731782 -1.0606602   -0.6785982
[6,]   0.9157433    0.3323588 -0.9192388  -1.220105642 -0.9192388    1.0827996
       CHLpjuv     PPpjuv  bakun_sti
[1,] -1.9364182 -2.0454212 -1.5268571
```

```
[2,] -1.4005172 -1.0273683 -1.2167479
[3,] -0.7564655 -0.6926709  1.1090711
[4,] -0.8155079 -0.6337877  0.3854830
[5,]  2.2314761  2.3045496 -0.3381052
[6,]  0.6224340  0.9496238  0.2304284
```

Is it right???

```
apply(scale_dat_scale, 2, mean)
```

```
         year         DDpre         DDegg        DDlarv        DDpjuv
-1.850372e-17  7.968163e-16  1.292947e-15 -6.059967e-16  9.529414e-16
        DDben          Tcop         Tpart       MLDpart       MLDlarv
 1.309138e-15 -1.327642e-15  8.789266e-16  1.119475e-15  2.377728e-15
     MLDpjuv       CSTlarv       CSTpjuv       LSTlarv       LSTpjuv
 8.049117e-15 -4.440892e-16 -1.110223e-16 -2.220446e-16  4.255855e-16
    hci1_larv     hci1_pjuv     hci2_larv     hci2_pjuv       oni_pre
 0.000000e+00 -5.088522e-17  0.000000e+00 -5.088522e-17 -3.700743e-17
     oni_larv      oni_pjuv      pdo_larv      pdo_pjuv   lusi_annual
 4.163336e-17  8.789266e-17 -7.401487e-17  1.110223e-16 -3.469447e-16
     ZOOpjuvN       ZOObenN      ZOOpjuvS       ZOObenS             X
 0.000000e+00 -3.700743e-17  6.476301e-17  1.225871e-16 -1.850372e-17
  CutiSTIpjuv  CutiTUMIpjuv          Year BeutiTUMIpjuv        Year.1
 5.458597e-16  8.974303e-16 -1.850372e-17 -1.110223e-15 -1.850372e-17
 BeutiSTIpjuv       CHLpjuv        PPpjuv     bakun_sti
 3.590878e-16 -1.110223e-15  8.141636e-16  5.551115e-17
```

```
apply(scale_dat_scale, 2, sd)
```

```
         year         DDpre         DDegg        DDlarv        DDpjuv
            1             1             1             1             1
        DDben          Tcop         Tpart       MLDpart       MLDlarv
            1             1             1             1             1
     MLDpjuv       CSTlarv       CSTpjuv       LSTlarv       LSTpjuv
            1             1             1             1             1
    hci1_larv     hci1_pjuv     hci2_larv     hci2_pjuv       oni_pre
            1             1             1             1             1
     oni_larv      oni_pjuv      pdo_larv      pdo_pjuv   lusi_annual
            1             1             1             1             1
     ZOOpjuvN       ZOObenN      ZOOpjuvS       ZOObenS             X
            1             1             1             1             1
```

```
 CutiSTIpjuv  CutiTUMIpjuv          Year BeutiTUMIpjuv        Year.1
           1             1             1             1             1
BeutiSTIpjuv       CHLpjuv        PPpjuv     bakun_sti
           1             1             1             1
```

There are a few things I did not execute flawlessly in this…you will notice I did not remove all variables. I also left year in the dataframe so it standardized that as well!! Which we don't want…I will leave it up to you to troubleshoot that :)

One thing to note is that if we want to use your long table, the easiest way to create standardized data in that format is looping over your variable name column. Alternatively, you could use the methods here and then merge or join the datasets together after you create a table of the standardized values.

Hope this helps! Happy standardizing!