IntroSlides with focus on statistics

Basic steps in workflow

- 1. Define environment
- 2. Import
- 3. Transform
- 4. Explore (general/outlier/distribution) (go back to 3?)
- 5. Classify scale level / distribution (based on 3/4)
- 6. Describe
- 7. Test / Model (may include step 6)
- 8. Report

Define environment

- Activate packages to use: library() / pacman::p_load()
- ggplot theme: theme_set() / theme_update()
- flextable settings: set_flextable_defaults()
- knitr::opts_chunk\$set()

```
[1] TRUE
```

```
1 set_flextable_defaults(
2    theme_fun = theme_zebra, font.size = 18, font.family = 'Roboto',
3    table.layout = 'autofit',
4    padding.bottom = .2, padding.top = .2, padding.left = 2, padding.right = 2)
5    knitr::opts_chunk$set(message = FALSE, warning = FALSE, comment = NA, echo = TRUE)
```

Import

- read_xlsx() / read_csv() / read_csv2()
- options related to separators, number formats, ranges etc.
- rename() / rename_with()

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Glimpse at data: Find the problems?

```
1 head(rawdata, n = 15) |> flextable()|>
       theme_zebra(even_body = 'aquamarine',odd_body = 'antiquewhite')
                                                          WEIGTH OF
                                                                       weight of
                                                                                  weight of
                                       ALMINIUM CAP +
                                       WEIGTH OF
                        WEIGHT OF
                                                                                             MOISTURE
                                                                         sample
CODE CODE OF
                                                                                    sample
                                                                         before after drying (Wts+d)
                            EMPTY
                                                     SAMPLE AFTER
                                                                                              CONTENT
OF CUP SAMPLE
                                   CUP + SAMPLE
                                                      DRYING (Wt-AL
                    ALUMINUM(wt)
                                                                                                    (%)
                                          (Wt + s)
                                                                                   (Wts+d)
                                                                          (Wts)
                                                               +s+d)
69
        D
                            4.1974
                                           9.3865
                                                              4.7000
                                                                         5.1891
                                                                                    4.6865
                                                                                              90.31431
                            4.1964
                                                                                               94.67008
        D
                                           9.2734
                                                                         5.0770
                                                                                    4.8064
                                                              4.4670
                            4.2108
4.2134
        D
                                           9.2653
                                                                         5.0545
                                                                                    4.5983
                                                                                               90.97438
                                                              4.6670
114
                                           9.3146
                                                                                    4.6801
                                                                                               91.74508
        D
                                                              4.6345
                                                                         5.1012
                                           9.3147
                                                              4.6171
                                                                         5.1291
                                                                                               91.58722
M1
        D
                            4.1856
                                                                                    4.6976
                                                                                    4.7543
a/17
        D
                            4.2090
                                           9.3204
                                                              4.5661
                                                                         5.1114
                                                                                               93.01366
        D
                                           9.2661
                                                                                    4.6883
                                                                         5.0767
                                                                                              92.34936
8
                            4.1894
                                                              4.5778
                                                                         5.0912
33
        D
                            4.1968
                                           9.2880
                                                              4.6057
                                                                                    4.6823
                                                                                               91.96849
                                           9.2872
        D
                                                                         5.1337
                                                                                    4.6522
Μ
                            4.1535
                                                              4.6350
                                                                                              90.62080
                                           9.2476
                                                                         4.9942
E/18/1
        D
                            4.2534
                                                              4.7403
                                                                                    4.5073
                                                                                               90.25069
24/A2
        D
                            4.2066
                                           8.3463
                                                              4.5849
                                                                         4.1397
                                                                                    3.7614
                                                                                               90.86166
13
                            4.1554
                                           9.2384
                                                              4.7402
                                                                         5.0830
                                                                                     4.4982
                                                                                               88.49498
                            4.1893
                                           9.2495
                                                                                    4.5114
                                                                                               89.15458
Хp
                                                              4.7381
                                                                         5.0602
2p/029
                            4.0654
                                           9.2173
                                                              4.6940
                                                                         5.1519
                                                                                    4.5233
                                                                                               87.79868
                                                                                    4.3908
                                                                                              85.43908
                                           9.2032
                                                              4.8124
                                                                         5.1391
                            4.0641
```

Rename

```
1 colnames (rawdata)
[1] "CODE OF CUP"
[2] "CODE OF SAMPLE"
[3] "WEIGHT OF EMPTY ALUMINUM(wt)"
[4] "WEIGTH OF ALMINIUM CUP + SAMPLE (Wt + s)"
[5] "WEIGTH OF ALUMINIUM CAP + SAMPLE AFTER DRYING (Wt-AL +s+d)"
[6] "weight of sample before drying (Wts)"
[7] "weight of sample after drying (Wts+d)"
[8] "MOISTURE CONTENT (%)"
 1 rawdata <- rawdata |>
     rename(Region=`CODE OF SAMPLE`) |>
     rename with (.fn = ~str replace all(.,
                                         c('AL.+UM'= 'Cup', 'C[UA]P' = 'Cup','\\(\\w+.*\\)
                                            'Cup Cup'='Cup',' '=' ')) |>
                    str to title() |> str trim())
   cn()
[1] "Code Of Cup"
                                          "Region"
                                          "Weigth Of Cup + Sample"
[3] "Weight Of Empty Cup"
[5] "Weigth Of Cup + Sample After Drying" "Weight Of Sample Before Drying"
                                         "Moisture Content (%)"
[7] "Weight Of Sample After Drying"
```

Transform

- Change or create columns with mutate() / mutate(across())
- e.g. for log-transformation, creation of factors, text recoding

```
1 rawdata <- rawdata |>
2 mutate(
3 `Weight Of Sample After Drying`=`Weight Of Cup + Sample After Drying`-
4 `Weight Of Empty Cup`,
5 `Dry Content (%)`=`Weight Of Sample After Drying`*100/
6 `Weight Of Sample Before Drying`,
7 `Moisture Content (%)`=100-`Dry Content (%)`)
```

	Code Of Cup	Region	Weight Of Empty Cup		+ Sample After Drying	Sample Before Drying	Sample After Drying	Moisture Content (%)	Content (%)
	69	D	4.1974	9.3865	4.7000	5.1891	0.5026	90.31431	9.685687
		D	4.1964	9.2734	4.4670	5.0770	0.2706	94.67008	5.329919
	Α	D	4.2108	9.2653	4.6670	5.0545	0.4562	90.97438	9.025621
	114	D	4.2134	9.3146	4.6345	5.1012	0.4211	91.74508	8.254920
	M1	D	4.1856	9.3147	4.6171	5.1291	0.4315	91.58722	8.412782

Explore / group variables

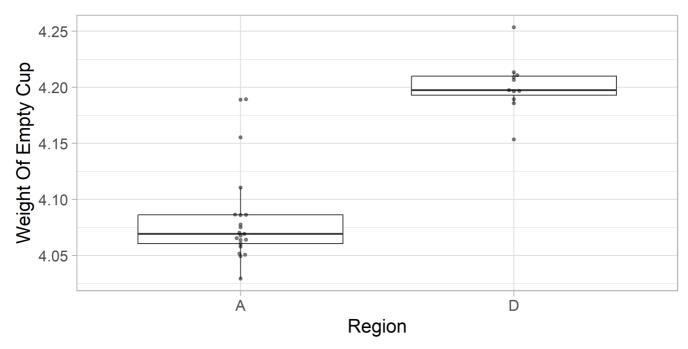
Explore (general/outlier/distribution)

- ggplot()+geom_boxplot() / geom_beeswarm() / geom_density()
- ks.test() / ksnormal() / shapiro.test()

Classify scale level / distribution

- gaussvars / ordvars / factvars, possibly more...
- Store variables accordingly, e.g. FindVars()

Explore: Outlier

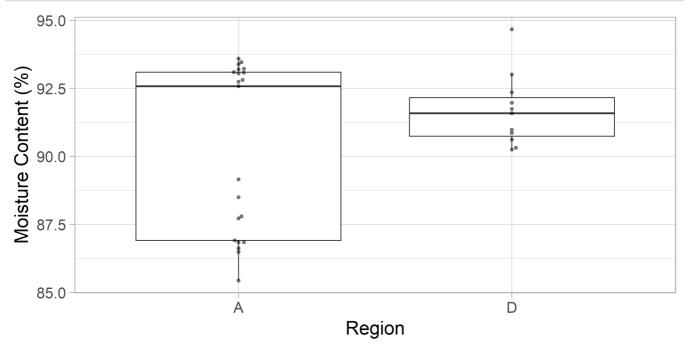


Handle outliers?

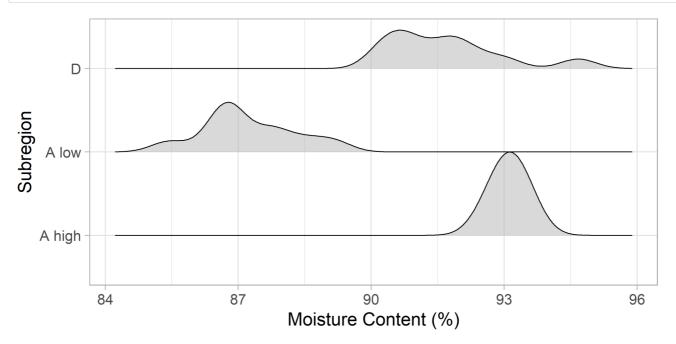
Removal is the worst strategy possible, correct errors, think about distributions, winsorize...



Explore: Unexpecteds



Transform Subregions?



Explore: Normal distribution 1

- Gaussian Normal distribution is required for many statistical procedures
- Common tests are graphical exploration, Shapiro-Wilk-test and Kolmogorov-Smirnov-test

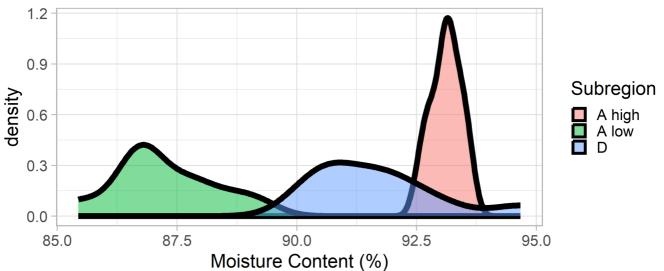
```
1 p_normal <-
2    shapiro.test(x = rawdata$`Moisture Content (%)`)
3 p_normal</pre>
```

```
Shapiro-Wilk normality test data: rawdata\$ Moisture Content (%) \$ W = 0.89133, p-value = 0.003752
```

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p (Shapiro) global = 0.004

A high: p=0.923; A low: p=0.756; D: p=0.207



Explore: Normal distribution 2

```
1 rawdata |>
     group by(`Subregion`) |>
     summarize(across(.cols = where(is.numeric), .fns = ksnormal)) |>
    pivot_longer(cols = -1, names_to = 'Variable', values_to = 'pKS') |>
    pivot_wider(names_from = `Subregion`, values_from = pKS)
\# A tibble: 7 \times 4
                                      `A high` `A low`
 Variable
                                                <dbl> <dbl>
 <chr>
                                         <dbl>
                                                0.444 0.697
1 Weight Of Empty Cup
                                        0.989
2 Weigth Of Cup + Sample
                                        0.785
                                                0.980 0.0187
3 Weigth Of Cup + Sample After Drying
                                        0.900
                                                0.710 0.969
4 Weight Of Sample Before Drying
                                        0.196
                                                0.999 0.0716
5 Weight Of Sample After Drying
                                        0.976
                                                0.555 1.00
                                        0.975 0.733 0.954
6 Moisture Content (%)
                                        0.975
7 Dry Content (%)
                                                0.733 0.954
```

Explore: Group variables by type/distribution

Scale level determines what statistics are appropriate Typical scale levels are

- nominal/categorical/factorial/qualitative: just different groups (species, eye color, genotype, treatment)
- ordered categories: few groups with inherent order (quality bad<medium<good, pain between 0 and 10)
- ordinal measures: many different values, natural order, no distribution assumption (satisfaction on a scale from 0 to 100)
- measures following a Normal distribution
- possibly measures from other known distributions (beta, log-normal, poisson...), often treated as ordinal

Make type decision obvious/reproducible

```
1 gaussvars <- FindVars(varnames = c('Weight Of Sample','Content'),</pre>
                           allnames = cn(rawdata))
 3 gaussvars
$index
[1] 6 7 8 9
[1] "Weight Of Sample Before Drying" "Weight Of Sample After Drying"
[3] "Moisture Content (%)"
                                      "Dry Content (%)"
$bticked
[1] "`Weight Of Sample Before Drying`" "`Weight Of Sample After Drying`"
                                       "`Dry Content (%)`"
[3] "`Moisture Content (%)`"
Scount
[1] 4
 1 ordvars <- FindVars(c('Cup'), exclude = 'Code')</pre>
 2 ordvars$names
[1] "Weight Of Empty Cup"
                                           "Weigth Of Cup + Sample"
[3] "Weigth Of Cup + Sample After Drying"
 1 factvars <- FindVars('egion', casesensitive = FALSE)</pre>
 2 factvars$bticked
[1] "`Region`"
               "`Subregion`"
```

Model

Describe

- mean() / sd() / meansd()
- median() / quantile() / median_quart()
- table() / prop.table() / cat_desc_stats()

Test

- t.test() / lm()+[Aa]nova() / compare2numvars()
- wilcox.test()
- fisher.test() / glm(family=binomial)

Model: Describe

Sample size n: per variable, if there are NAs

Mean: central tendency, the expected typical value

 $\frac{\sum x}{n}$

Variance: measure for variability/heterogeneity of data

Standard deviation SD: the *typical* weighted deviation from the mean

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Standard error of the mean SEM: how reliable is the mean *estimate*, what would be the expected SD of means from repeated experiments?

Median: Split between lower/upper 50% of data

Quartiles: Split at 25%/50%/75% of data (more

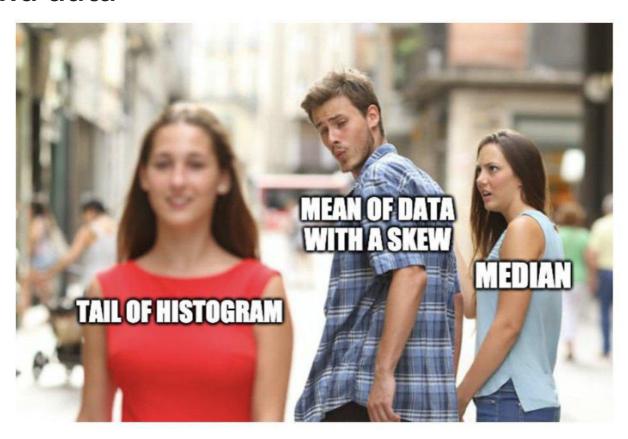
general: Quantiles, e.g. Percentiles), used in

boxplot

various computational approaches

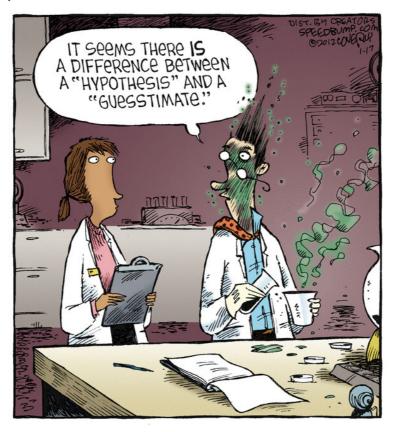
```
1 desc_gauss <- rawdata |>
    summarize(across(.cols = gaussvars$names,
                        .fns = meansd))
 4 desc gauss
# A tibble: 1 × 4
  `Weight Of Sample Before Drying` Weight Of Sample Aft...¹ `Moisture Content (%)`
15.1 \pm 0.2
                                     0.47 \pm 0.14
                                                             91 \pm 3
# i abbreviated name: 1`Weight Of Sample After Drying`
# i 1 more variable: `Dry Content (%) ` <chr>
 1 desc ord <- rawdata |>
    summarize(across(ordvars$names,.fns=~median quart(.x,roundDig = 3))) |>
    pivot_longer(everything(),
 4
                   names to = 'Measure', values to = 'Median[1Q/3Q]')
 5 desc_ord
\# A tibble: 3 \times 2
 Measure
                                        `Median[1Q/3Q]`
 <chr>
                                       <chr>
1 Weight Of Empty Cup 4.09 (4.06/4.19)
2 Weight Of Cup + Sample 9.25 (9.22/9.29)
3 Weigth Of Cup + Sample After Drying 4.61 (4.45/4.72)
```

Descriptive Stats should match distribution and data



Model: Test

Tests require hypotheses



Null hypothesis?

- Working hypothesis: This is what you expect!
 E.g. treatment is lowering blood pressure more than placebo,
 transgenic animals become obese, bio reactor A is more efficient than
 B, concentration of substance is correlated with speed of reaction ...
- Null hypothesis: This is what you test!
 No difference / relation, BP under therapy = BP under placebo

4 possibilities:

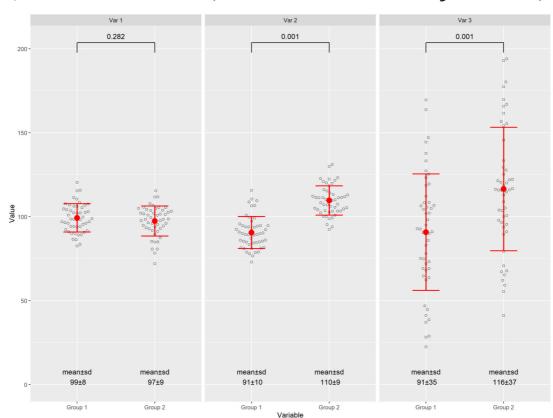
- Null hypothesis correct, test false positive (case A): alpha-error
- Null hypothesis correct, test correct negative (case B)
- Null hypothesis false, test false negative (case C): beta-error
- Null hypothesis false, test correct positive (case D)

Significance: NOT probability of case A, but probability of your data given the NULL hypothesis, calculated from your data, conventionally <0.05

Power: Probability of case D, *estimated* based on assumptions about effects and sample size, *calculation* would require knowledge of true difference, conventionally set at 0.80

Test functions

t-test / Wilcoxon-test (aka Mann-Whitney U-test)



t-test

- Assumptions: Continuous data with Normal distribution
- 1 or 2 (independent or dependent) samples with/without equal variances
- how big is the mean difference relative to uncertainty?
 t = (mean₁ mean₂)/SEM
- t follows a t-distribution, allows estimation of probability of t under the NULL hypothesis

Wilcoxon-test

- nonparametric, no distribution is assumed
- based on rank-transformed data
- insensitive to extreme values

Test examples: single variables

```
Welch Two Sample t-test

data: Moisture Content (%) by Region

t = -1.7274, df = 29.239, p-value = 0.09465

alternative hypothesis: true difference in means between group A and group D is not equal

to 0

95 percent confidence interval:

-2.9624835  0.2490486

sample estimates:

mean in group A mean in group D

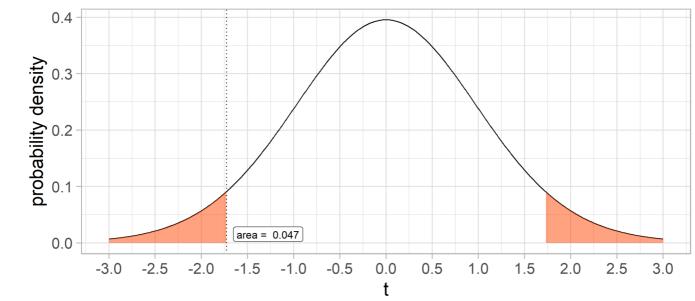
90.31199  91.66870

1 #Wilcoxon-Test
2 wilcox.test(`Moisture Content (%)`~`Region`,
3  data = rawdata)
```

```
Wilcoxon rank sum exact test data: Moisture Content (%) by Region W = 107, \; p\text{-value} = 0.7547 alternative hypothesis: true location shift is not equal to 0
```

From t to p

from t-test: t = -1.7, p = 0.095



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Model: Test 2 / multiple variables

desc_all Region A **p** 0.12830 **Region D** Weight Of Sample Before Drying 5.1 ± 0.2 5.2 ± 0.1 5.0 ± 0.3 Weight Of Sample After Drying $0.47 \pm 0.14 \ 0.50 \pm 0.16 \ 0.42 \pm 0.07 \ 0.04937$ 90 ± 3 9.7 ± 3.1 0.09465 Moisture Content (%) 91 ± 3 92 ± 1 9.2 ± 2.7 8.3 ± 1.3 Dry Content (%) 0.09465

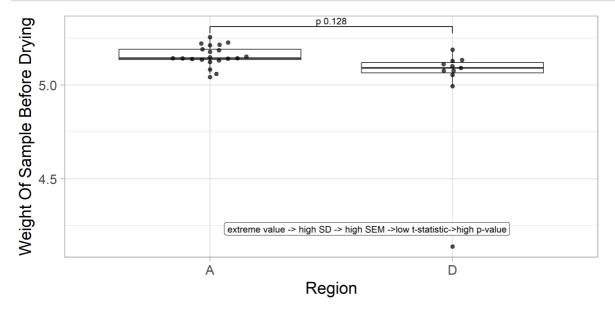
```
        Variable
        desc_all
        Region A
        Region D
        p

        Weight Of Empty Cup
        4.09 (4.06/4.19) 4.07 (4.06/4.09) 4.20 (4.19/4.21) 0.001

        Weigth Of Cup + Sample
        9.25 (9.22/9.29) 9.24 (9.22/9.25) 9.29 (9.27/9.31) 0.003

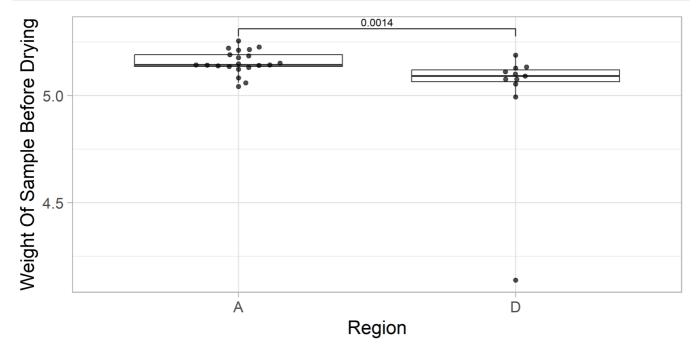
        Weigth Of Cup + Sample After Drying 4.61 (4.45/4.72) 4.46 (4.42/4.74) 4.62 (4.58/4.66) 0.611
```

Show results



Re-thinking test decision?

```
ggplot(rawdata, aes(x = `Region`,y = `Weight Of Sample Before Drying`))+
geom_boxplot(outlier.alpha = 0)+
geom_beeswarm(alpha=.7, size=2,cex = 2)+
geom_signif(comparisons = list(c(1,2)),test = wilcox.test)
```

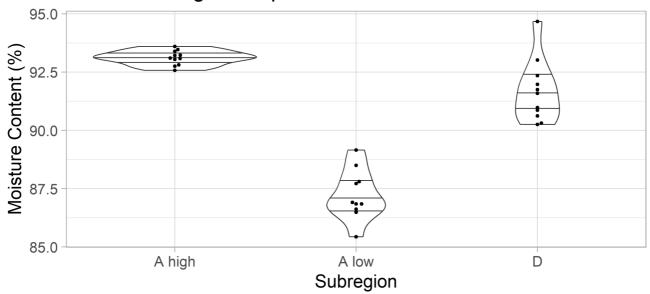


Model: linear models 1 / univariable

```
plottmp <- ggplot(rawdata,aes(Subregion, `Moisture Content (%) `))+
geom_violin(draw_quantiles = c(.25,.5,.75))+
geom_beeswarm()+

ggtitle('Are all Subregions equal?')
print(plottmp)</pre>
```

Are all Subregions equal?



ANOVA: build model

```
1 rawdata |> group_by(Subregion) |>
2  summarize(MeanMoisture=mean(`Moisture Content (%)`) |> roundR(4)) |>
3  pivot_wider(names_from = Subregion, values_from = MeanMoisture) |>
4  rename_with(~paste('Mean moisture %\n',.x)) |> flextable()|>
5  theme_zebra(even_body = 'aquamarine',odd_body = 'antiquewhite')
```

Mean moisture % Mean moisture % Mean moisture % A high A low D 93.11 87.23 91.67

```
1 lm1<- lm(`Moisture Content (%)`~Subregion, data=rawdata)
2 lm1</pre>
Call:
```

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ANOVA: get p-values

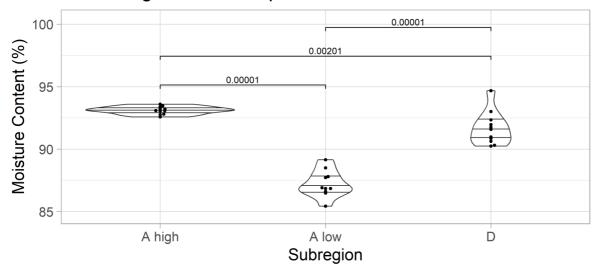
```
1 anova(lm1) |> broom::tidy() |> flextable()|>
2 theme_zebra(even_body = 'aquamarine',odd_body = 'antiquewhite')
```

term df sumsq meansq statistic p.value Subregion 2 194.34239 97.1711969 97.76477 0.0000000000001292127 Residuals 29 28.82393 0.9939285

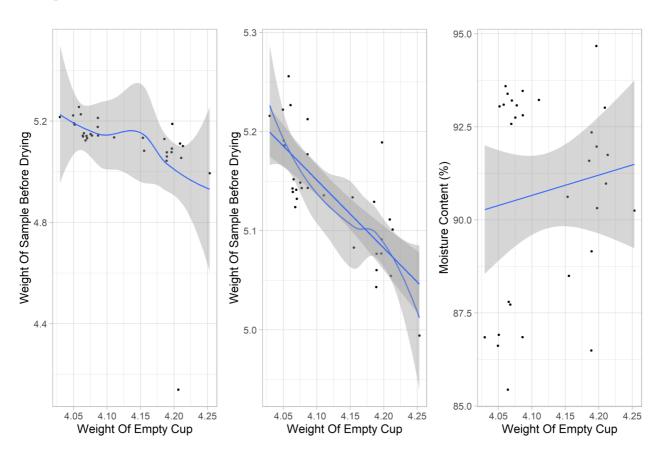
```
A low "0.00001" " NA"
D "0.00201" "0.00001"
```

Visualize ANOVA

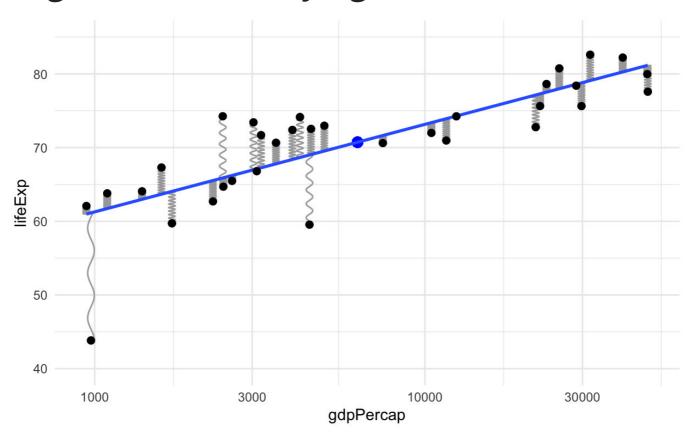
All Subregions are unequal



Regression: Scatterplot

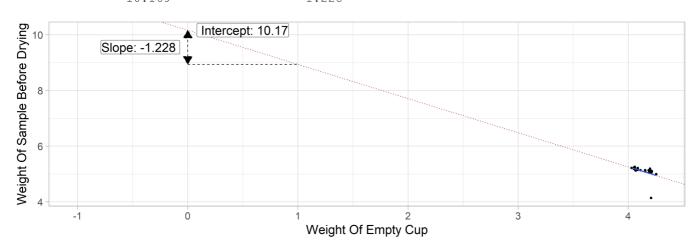


Regression: Underlying mechanics



Regression: Statistics

Coefficients:
(Intercept) `Weight Of Empty Cup`
10.169 -1.228



Regression: Significance

Report

- RMarkdown and quarto are powerful tools to create reports and presentations
- Export figures: ggsave() / png() / pdf()
- Export tables: write_xlsx()
- Package flextable provides nice features for table formatting

Flextable example

```
1 test_ord |> select(-desc_all) |> rename_with(~str_remove(.,'Code Of ')) |>
2    flextable() |>
3    theme_zebra(even_body = 'aquamarine',odd_body = 'antiquewhite')|>
4    italic(~p<=0.05,j = 1) |> bg(~p<=0.05,j = 4,bg = 'yellow') |>
5    set_caption('Treatment effects, measures following a normal distribution') |>
6    add_footer_lines('Significance level is set at 0.05') |>
7    fontsize(size = 12,part = 'footer')
```

```
        Variable
        Region A
        Region D
        p

        Weight Of Empty Cup
        4.07 (4.06/4.09) 4.20 (4.19/4.21) 0.001
        0.001

        Weigth Of Cup + Sample
        9.24 (9.22/9.25) 9.29 (9.27/9.31) 0.003

        Weigth Of Cup + Sample After Drying 4.46 (4.42/4.74) 4.62 (4.58/4.66) 0.611

        Significance level is set at 0.05
```

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Useful tools along the way

- Pick columns / rows: select() / pull() / filter() / slice()
- Change format of tibble wide <--> long (e.g. for repeated measures): pivot_longer()/pivot_wider()
- Regular expressions: str_replace() / str_detect() / str_...
- Merge text elements: paste() / str_glue()
- Apply functions: purrr::map_xxx