

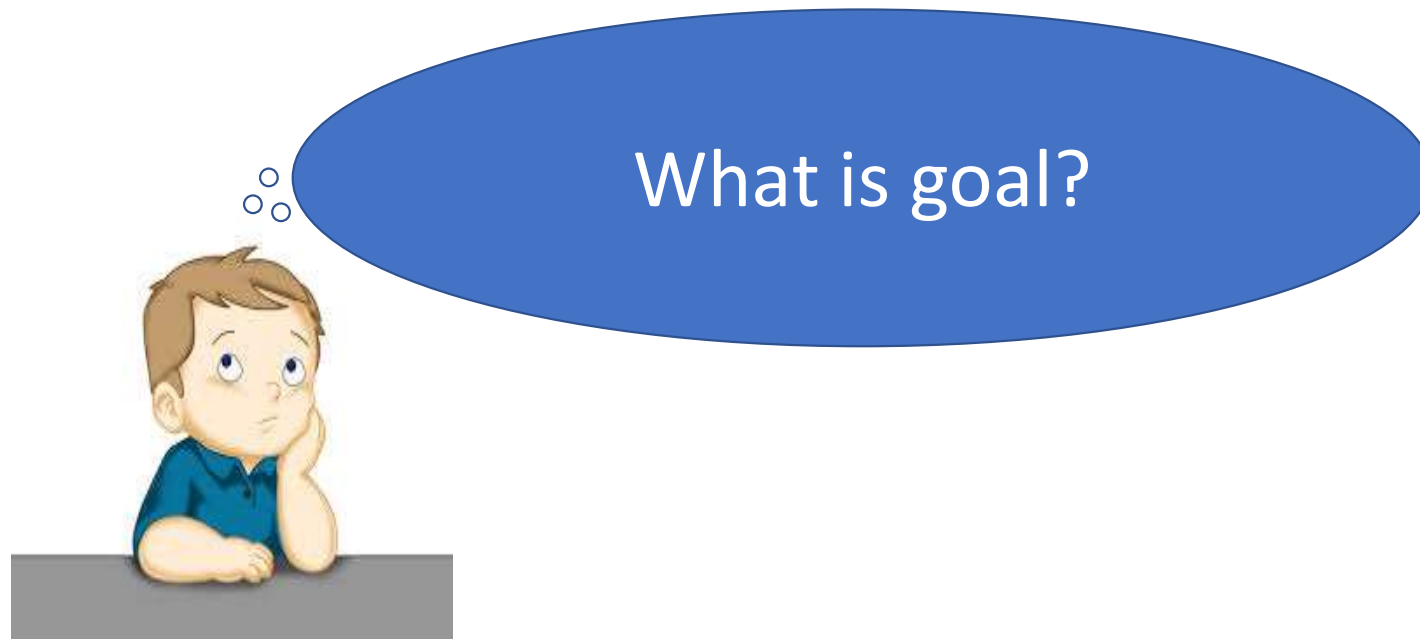


# Experimental design

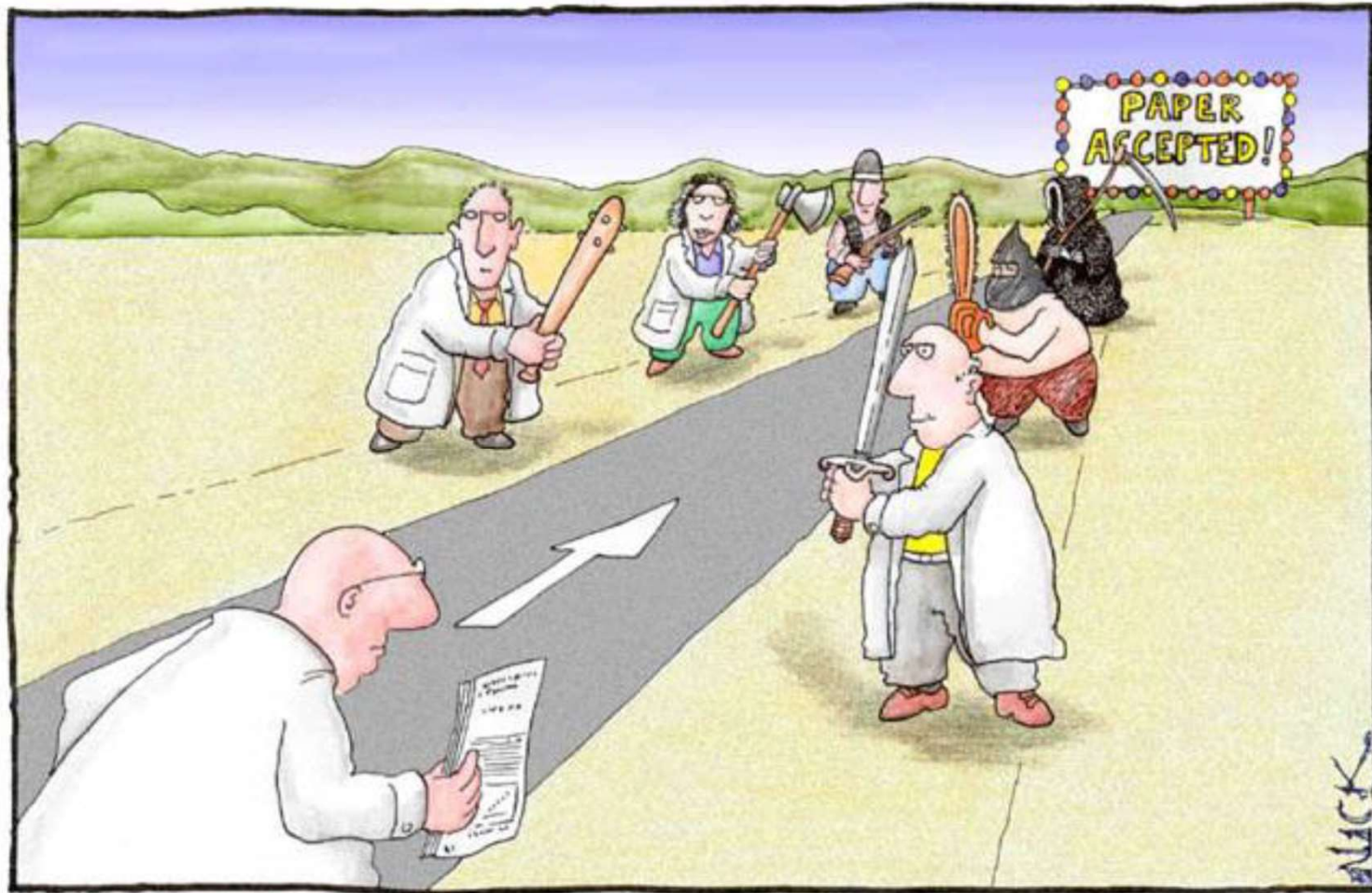
Nguyen Le Phuong Lien, PhD

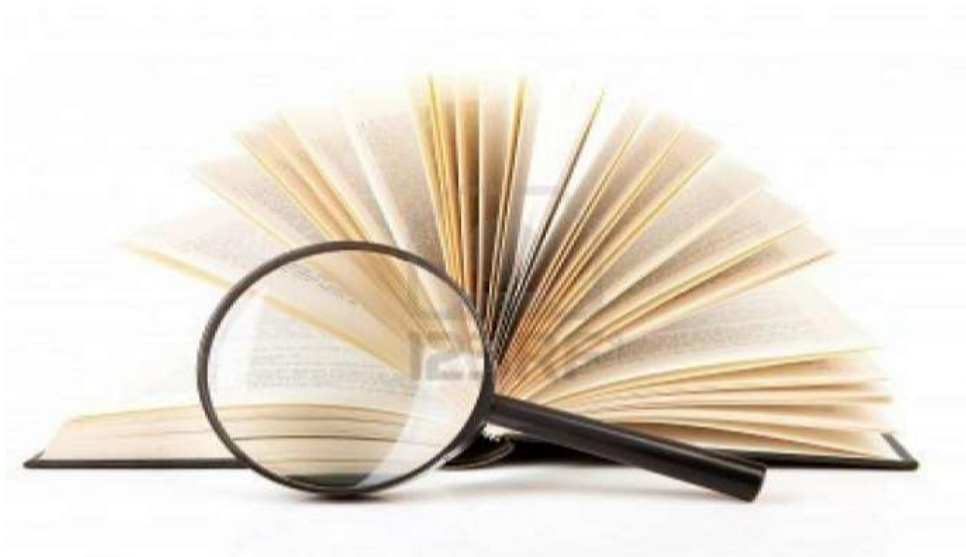
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Ho Chi Minh, December 29<sup>th</sup> 2020

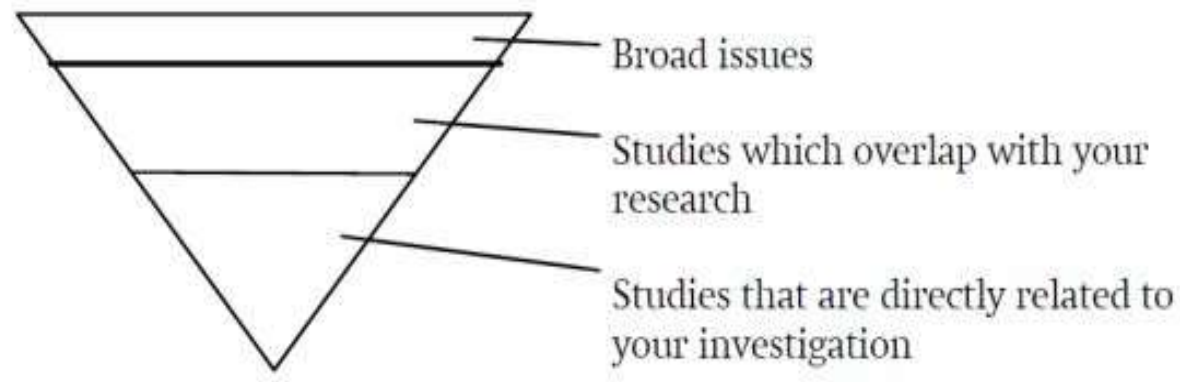


## Publishing is easy or not?





## Literature Review



## Methodology of planning experiments

1. What is the goal?  
Why we do the experiment?
2. How we can answer the question?  
What kind of analysis can calculate we need?
3. What we have to measure?  
How many measurements and what combinations  
have to be performed?

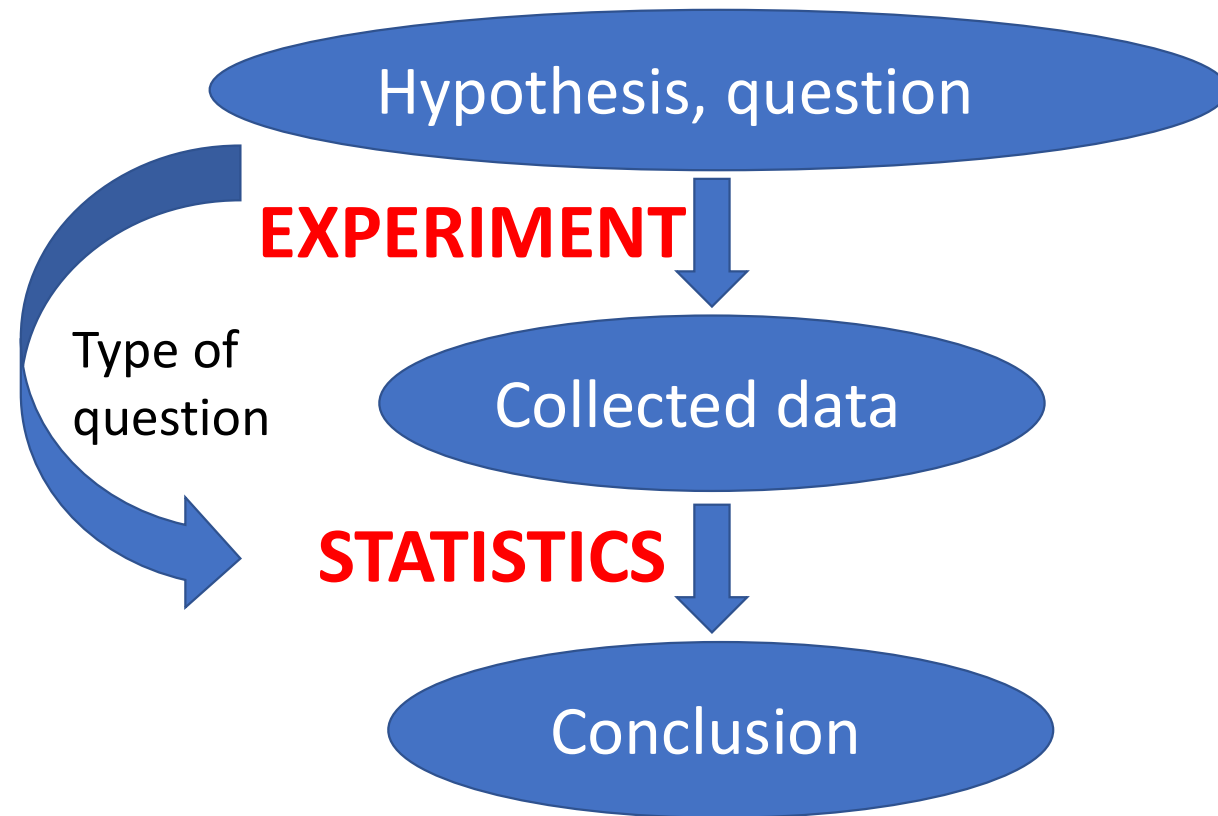
## Science vs Industry approach

There are different objectives for science and industry:

- Science: discover the statistical effect of particular factors on the feature of our interest.  
*For example: how storage temperature is affecting fruit firmness?*
- Industry: extract maximal unbiased (independent) information about factors affecting production from minimal observations.  
*For example: what we can change in recipe to increase consumer acceptance?*

- \* Hypothesis: a statement about the population
- \* Based on our collected data we conclude to the whole phenomenon (population)
- \* Whether our result (difference in samples) is greater than the difference caused **only by chance**.

# Experimental design





## What is the goal?

The objective of the experiment determines what kind of methods will be available for analysis, as well as how many samples we need.

- Is it univariate relationship?  $y = f(x)$   
Typically no. Sorry, life is more complex.
- Is it multivariate relationship?  $y = f(x_1, x_2, \dots)$   
This is more realistic. The problem is to select the meaningful factors to save money and time.

## What is the question?

The question determines the analysis type:

- Do you want to see **the effect of change**?  
*Question: how long can fruit keep quality if I decrease storage temperature from 10 °C to 5 °C?*  
**Need comparison methods.**
- Do you want to **predict value without measurement**?  
*Question: how can I predict soluble solids content of fruit without destroying it?*  
**Need regression analysis.**

## How we can answer?

### \* Comparison methods

- **Two groups:**

t-test (parametric test = requires normal distribution of data),  
Kolmogorov-Smirnov test (robust = for any distribution)

- **Many groups:**

Analysis of Variances (ANOVA).

Warning! **ANOVA** only can answer the question whether there is at least one group different from others! If you want to identify group differences, need to **run post-hoc test**: Tukey test, Duncan test, multiple range test, etc.

## How we can answer?

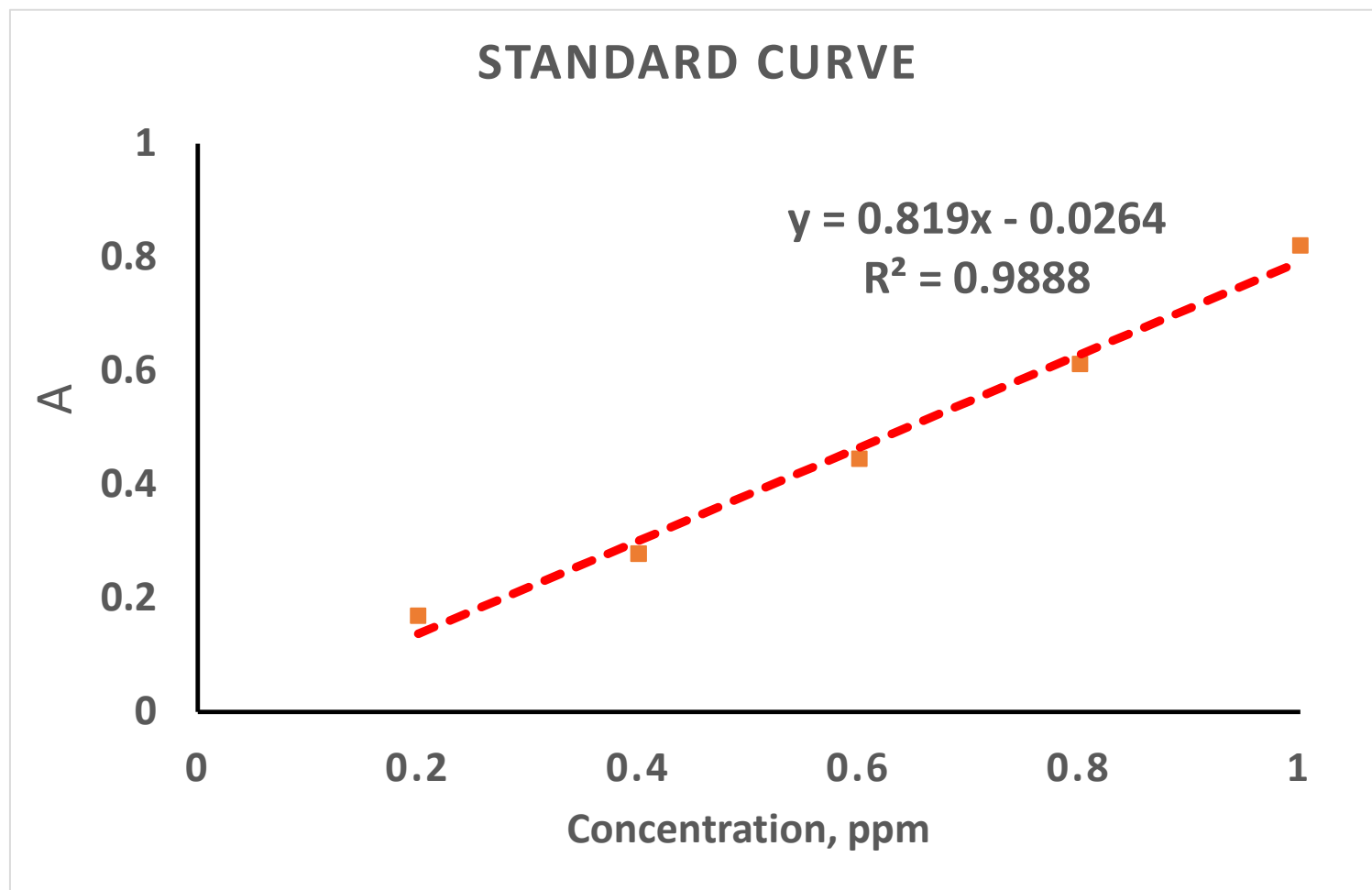
### \* Regression methods

- Is there **linear relationship** between them?  
Calculate Pearson correlation.
- Is there **non-linear relationship** between them?  
Calculate Kendall or Spearman (rank) correlation.
- How can I **predict feature or future value**?

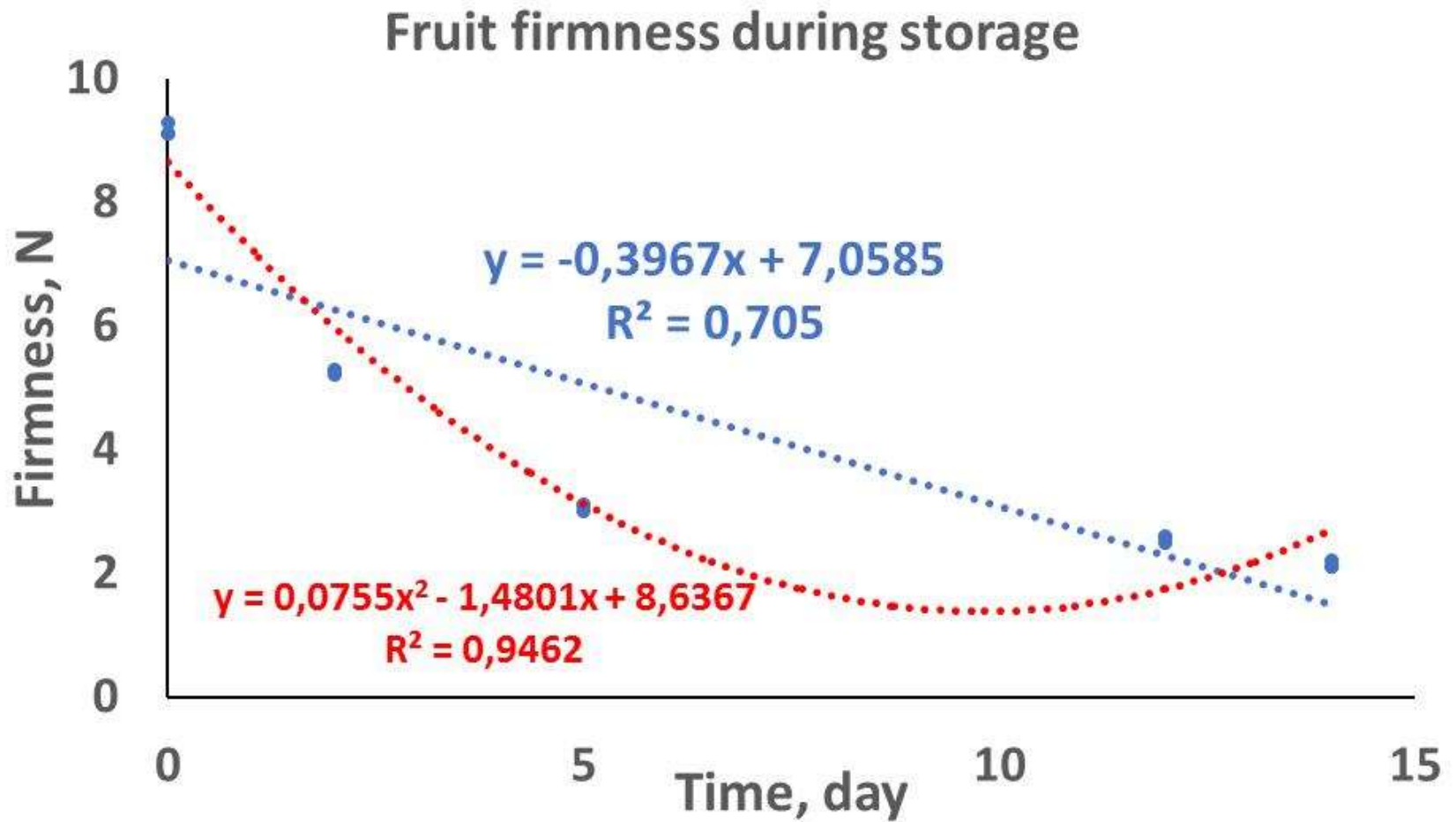
#### Fit curve on data:

- Linear model for trend or calibration.
- Model is usually given by chemistry (Arrhenius equation) or physics (Saturation model)

## Example



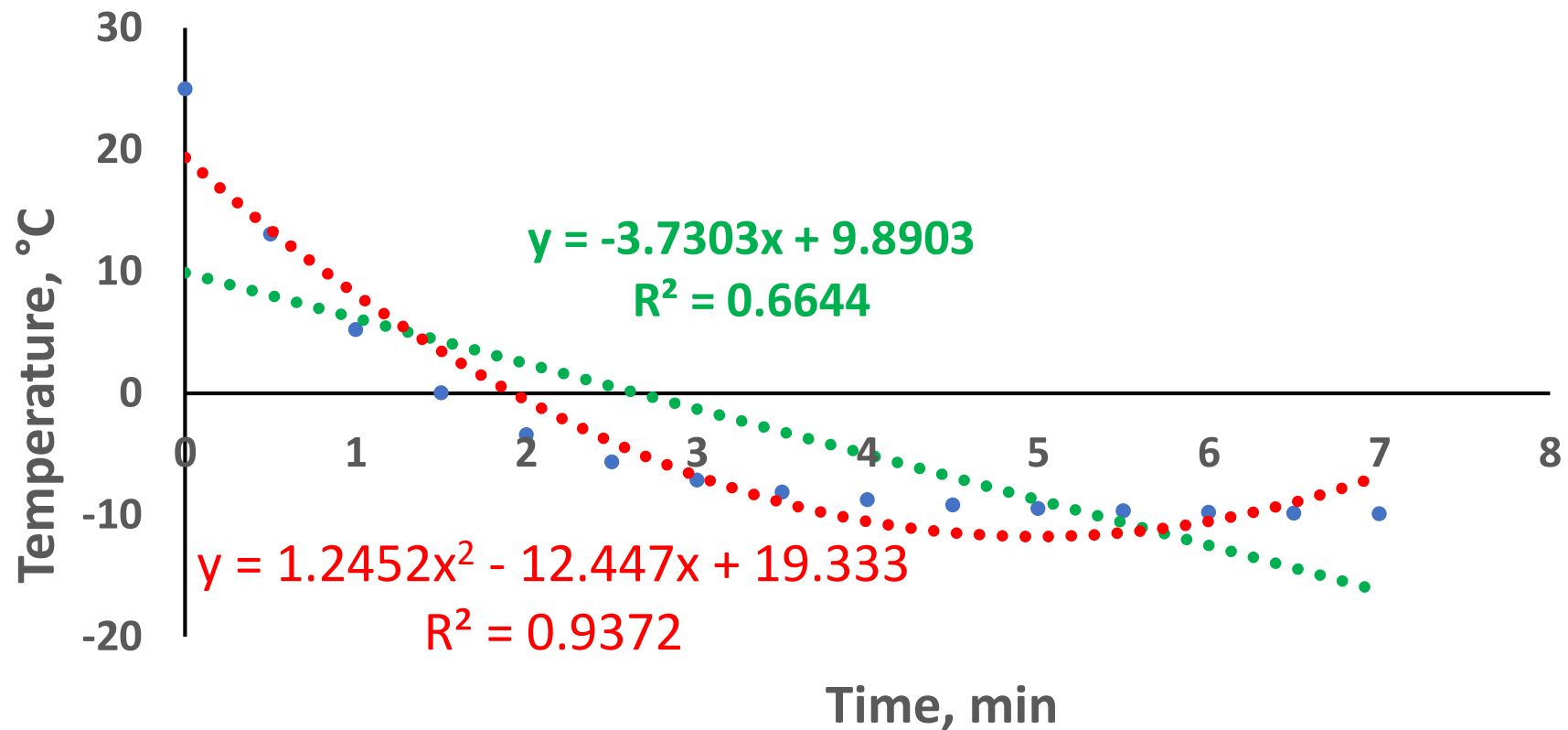
# Example



## Example

What is the equation that can describe the presented phenomenon?

Ice cream temperature in ice cream making machine



## How we can answer?

### \* Regression methods

- What is the **optimal combination**?

Response Surface Methodology (RSM).

$$y = a + bx_1 + cx_2 + dx_1x_2 + ex_1^2 + fx_2^2 + \text{error}$$

- How can I predict soluble solids content (SSC)?

Chemometry: use Near-Infrared (NIR) spectroscopy and **make model with Partial Least Squares (PLS) regression**.

$$\text{SSC} = f( LV_1(\text{spectra}), LV_2(\text{spectra}), \dots )$$



## What combinations are measured?

Measurement groups:

	Cooking time		
	5 min	10 min	15 min
Temperature			
80 °C	_____	_____	_____
100 °C	_____	_____	_____
120 °C	_____	_____	_____

This is called full factorial = all combinations are measured.

## What combinations measure?

Statistical analysis uses level codes to balance differences between range of values:

Cooking time

-1

0

+1

Temperature

-1

\_\_\_\_\_

\_\_\_\_\_

\_\_\_\_\_

0

\_\_\_\_\_

\_\_\_\_\_

\_\_\_\_\_

+1

\_\_\_\_\_

\_\_\_\_\_

\_\_\_\_\_

## How we can answer?

What if we want to **recognize quality class, infection, damage**, etc?

Need **classification methods**. They are supervised methods. We teach the computer and it will calculate which group is closest to the new measurement?

- Discriminant Analysis (DA) is based on the variance of data.
- Support Vector Machine (SVM) is based on the high density locations of the data cloud.

# Factorial design

## Terminology in experimental design

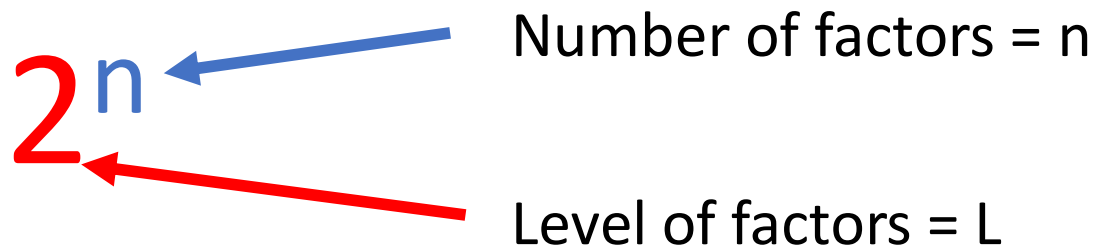
Experiment

Treatment or factor

Level

Experimental unit

## $2^n$ factorial design



Total number of sample groups:

$$N_G = L^n$$

## $2^n$ factorial design

- What are  $2^n$  factorial design?

- design to determine effect of  $n$  factors, each with 2 levels

- Why consider  $2^n$  factorial design?

- easy to analyze

- factors that have significant impact → full factorial design

- factors have little impact → not of interest

- How to select 2 levels

- if factor effect is expected to be unidirectional → select min, max

## $2^n$ factorial design

Number of factors = 1

Level of factors = 2

1 factor groups:

Low  
value

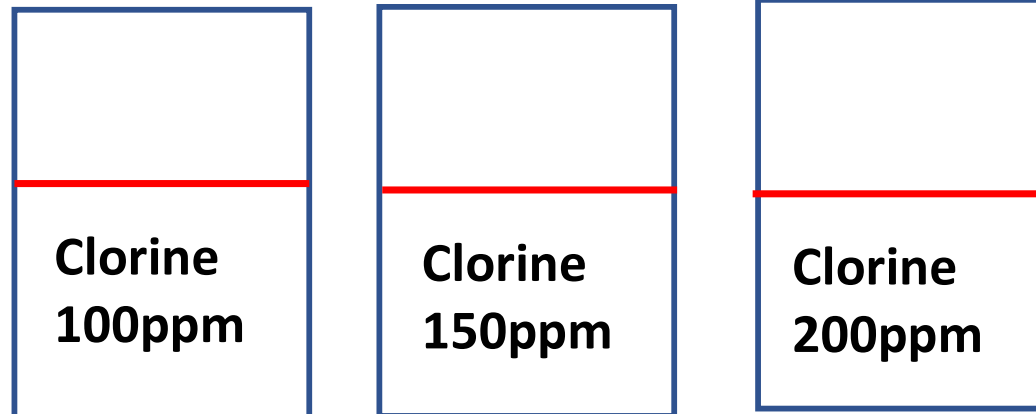
Group A

High  
value

Group B

## Example

Evaluating the effect of chlorine in washing fruit at different concentration: 100, 150, 200ppm.



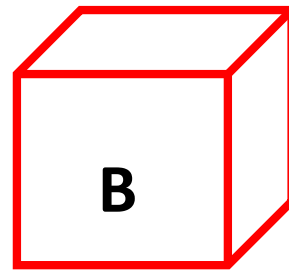
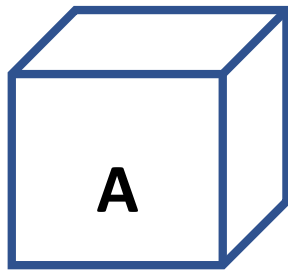


## Experimental design

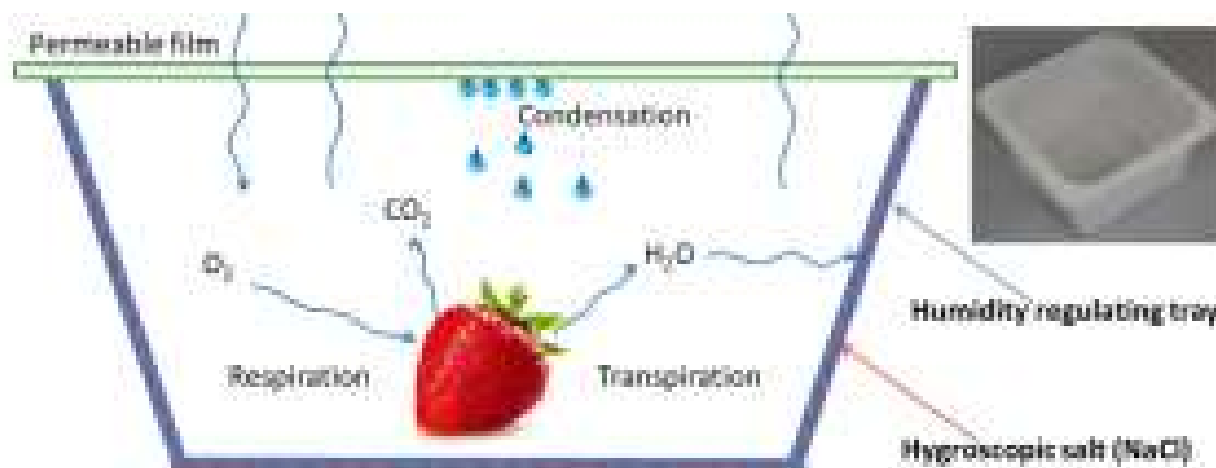
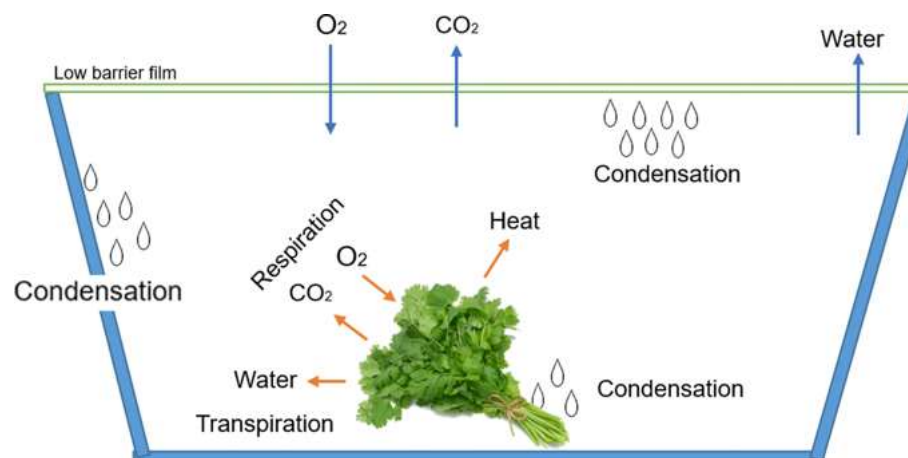


### Example

- There are 2 types of packaging: A and B.
- Evaluating the effect of packaging on vegetable during storage for 10 days. How many groups?



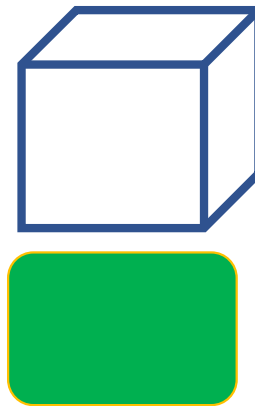
## Example



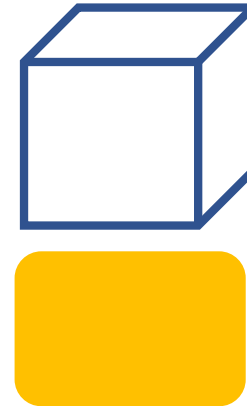
Moisture absorber pad

## Example

-There are 2 types of moisture absorber: 30% fructose and 90% fructose.  
Evaluating the effect of moisture absorber on packaging vegetable during storage for 10 days. How many groups?



**30% Fructose**



**90% Fructose**

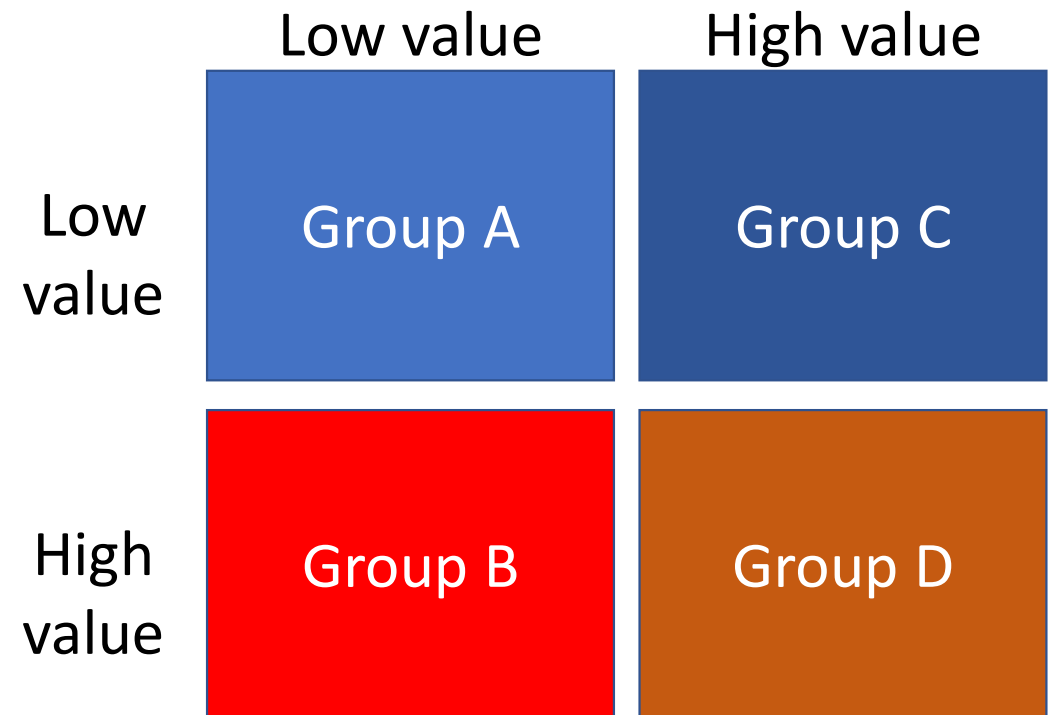
## 2<sup>n</sup> factorial design

Number of factors = 2

Level of factors = 2

1 factor groups:

2 factor groups:





## Example

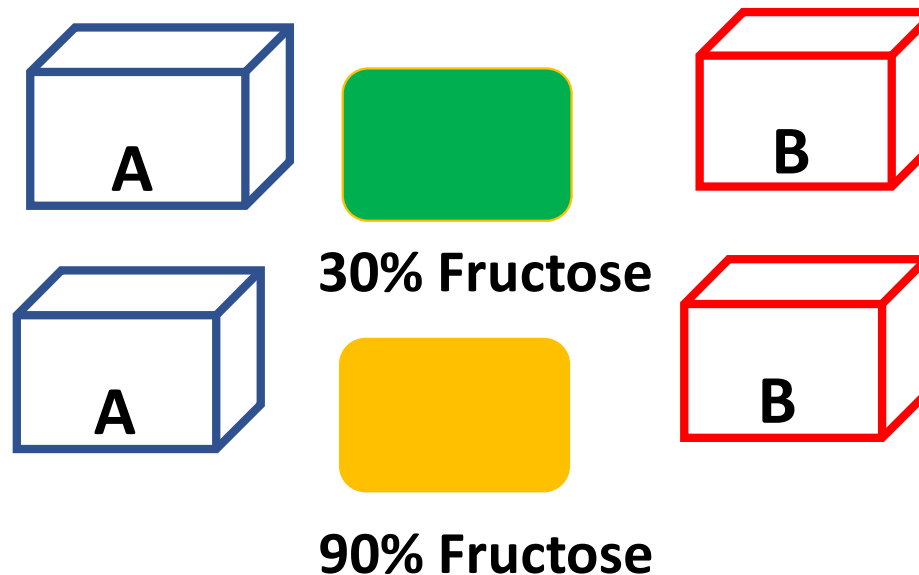
Measurement groups:

	Cooking time		
	5 min	10 min	15 min
Temperature			
80 °C	_____	_____	_____
100 °C	_____	_____	_____
120 °C	_____	_____	_____

Full factorial = all combinations are measured.

## Experiment

-There are 2 types of packaging: A and B and 2 types of moisture absorber. Evaluating the effect of packaging and absorber on vegetable during storage. How many groups?



## Experimental design:

- $2^n$  factorial design: two levels, n factor full design
- $3^n$  factorial design: three levels, n factor full design
- Central composite design (RSM)
- Latin square design
- Taguchi robust design
- Mixture design
- Fractional factorial design

## What combinations are measured?

Number of measurements (N):

$$N = L^n \times R$$

n = number of factors (2 = cooking time, temperature)

L = number of levels adjusted (3 = low, medium, high)

R = number of replicates (3)

$$N = 3^2 \times 3 = 27$$



## Example

Measurement groups:

	Cooking time		
	5 min	10 min	15 min
Temperature			
80 °C	_____	_____	_____
100 °C	_____	_____	_____
120 °C	_____	_____	_____

This is called full factorial design = all combinations are measured.

## Example

Temperature (°C)	pH	Time (min)
50	4.0	10
60	5.0	20
70	6.0	30

Full factorial design = ? combinations

$$Y = a_0 + a_1X_1 + a_2X_2 + a_3X_3 + a_4X_1^2 + a_5X_2^2 + a_6X_3^2 + a_7X_1X_2 + a_8X_1X_3 + a_9X_2X_3$$

$$Y = 45.82 - 1.39X_1 + 1.19X_2 - 2.65X_1^2 - 2.1X_2^2 - 1.13X_3^2 + 0.65X_1X_2 + 0.65X_1X_3 + 1.22X_2X_3$$

## Factorial design

### What is block?

Block is a homogeneous unit of samples:

- Crops harvested at the same time
- Feeding made by the same company
- Animals living in the same barn
- Solutions prepared by the same technician

Block is usually added as additional factor to avoid systematic error. It makes statistic more robust and stronger. This is called **block design**.

## What is mixture design?

In mixture design, the summary of all factors are always the same.

Example:

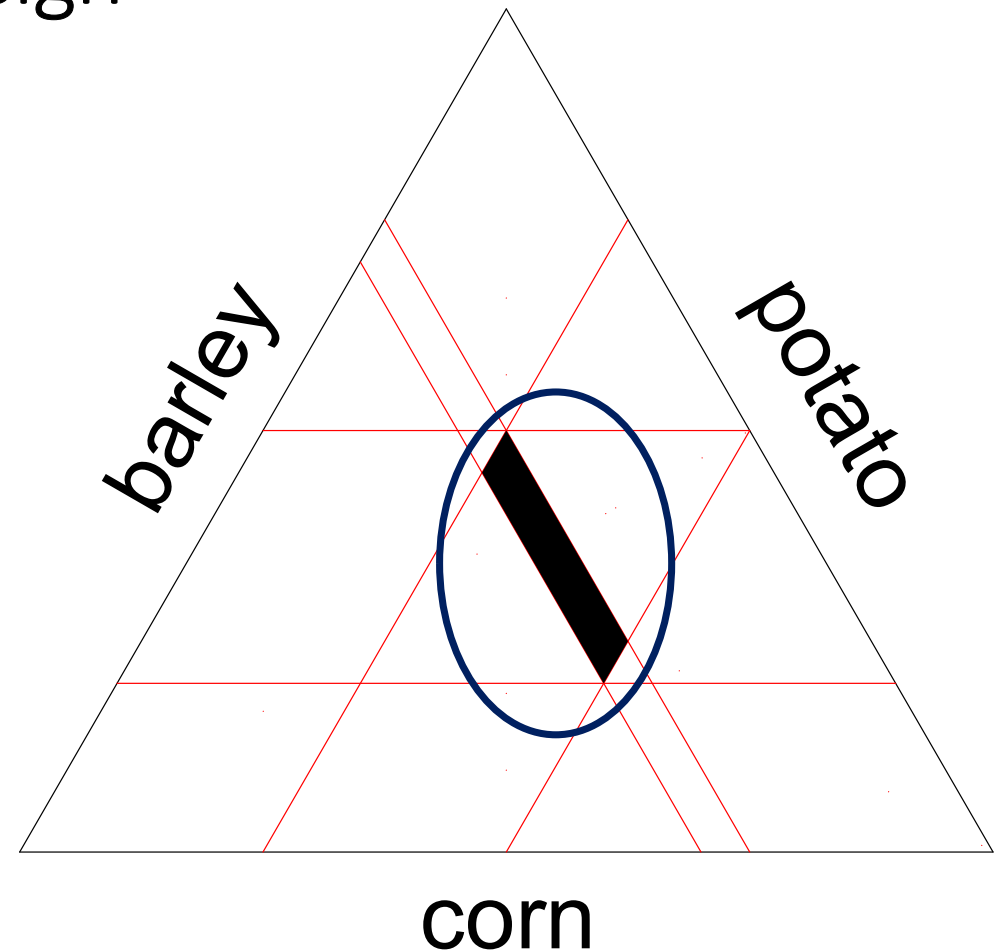
- Feeding #1: 20% corn, 50% barley, 30% potato
- Feeding #2: 50% corn, 25% barley, 25% potato

All feedings have the summary of 100%

## Mixture design

### Plotting mixture design

- Corn: 20 – 50%
- Barley: 25 – 50%
- Potato: 25 – 30%



## Question

1. Which type of experimental design is suitable for this problem?
2. Please propose an experiment!

Problem: Our yoghurt product received 79% satisfaction on consumer survey. We want to improve by changing ingredients. The current contents of one portion:

- 110 g natural yoghurt
- 8 g cherry fruit pulp
- 7 g extruded cereal

One portion totally: 125 g

The cup size is given, filling must be 125 g.

## Fractional factorial design

A full factorial design may require many experiments

- How can we get enough info with less: **fractional factorial design**

	Cooking time		
	5 min	10 min	15 min
Temperature			
80 °C	___ <b>x</b> ___	_____	___ <b>x</b> ___
100 °C	_____	___ <b>x</b> ___	_____
120 °C	___ <b>x</b> ___	_____	___ <b>x</b> ___

Full factorial design = all combinations are measured.

## Example

Temperature (°C)	pH	Time (min)
50	4.0	10
60	5.0	20
70	6.0	30

$$Y = a_0 + a_1X_1 + a_2X_2 + a_3X_3 + a_4X_1^2 + a_5X_2^2 + a_6X_3^2 + a_7X_1X_2 + a_8X_1X_3 + a_9X_2X_3$$

$$Y = 45.82 - 1.39X_1 + 1.19X_2 - 2.65X_1^2 - 2.1X_2^2 - 1.13X_3^2 + 0.65X_1X_2 + 0.65X_1X_3 + 1.22X_2X_3$$



## Why experimental design?

- Obtain maximum information from fewest experiments
  - minimize time spent gathering data
- Quantify effects from different factors using analysis
- Determine if a factor's effect is significant
  - differences might be random variations caused by
    - measurement errors
    - parameters not controlled

## Basic principles

1. Thinking about the question in advance
2. Comparison/control
3. Replication
4. Randomization
5. Factorial experiments

## Example

### Question

Does chemical treatment/packaging affect the quality of the food?

Does the temperature/pH affect the ...?

Does the solvent/time/... affect the...?

### Experiment:

1. Treat the fruit with chemical
2. Wait 7 days
3. Measure the quality

## Experimental design

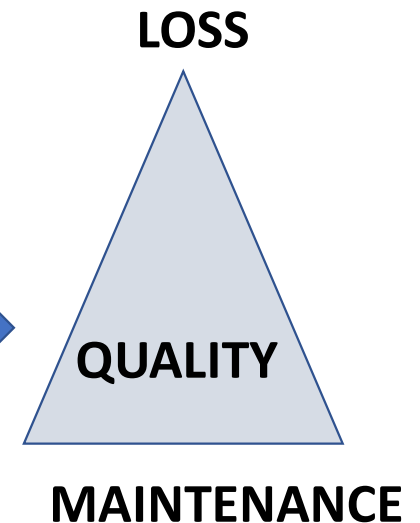
Example: How the chemical effect the quality of the fruit during storage?

**WITHOUT CHEMICAL  
TREATMENT**



**CHEMICAL  
TREATMENT**

**STORAGE**



## Experiment

Initial



**CONTROL**



**WITHOUT TREATMENT**



**TREATMENT**

## Comparison/control

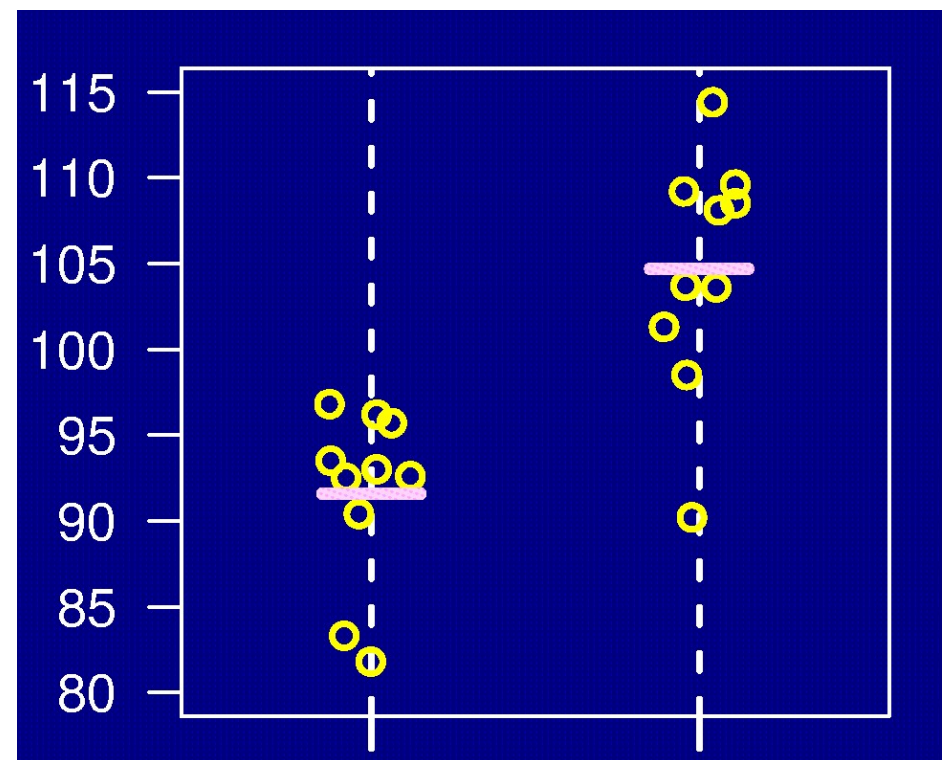
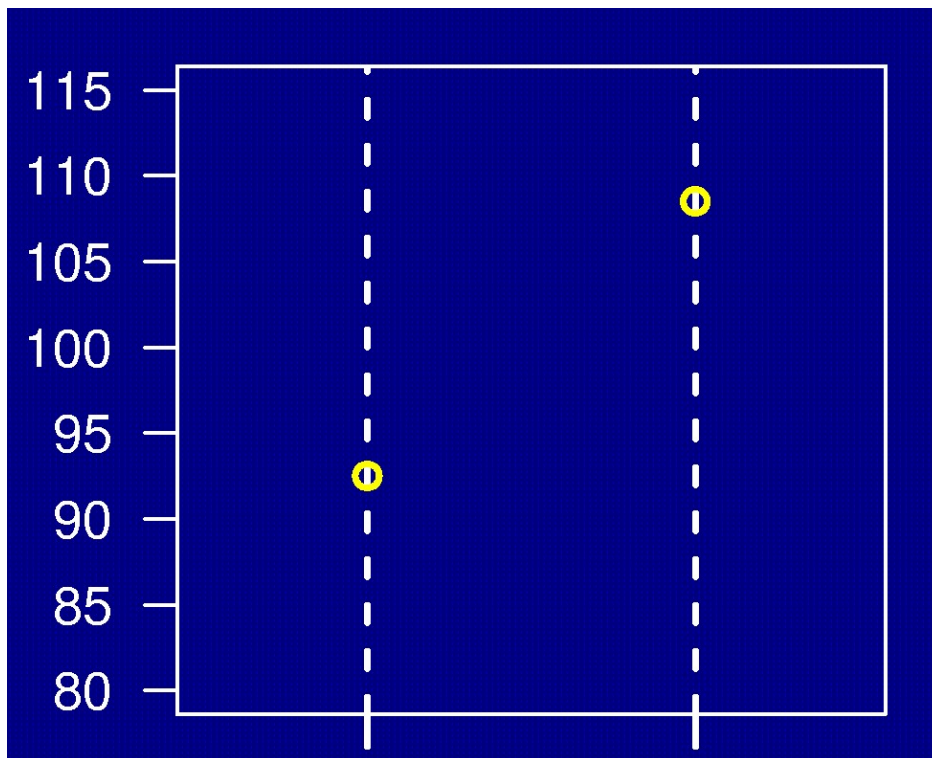
In the experimental design, group is compared to **concurrent** control (rather than to **initial** control).

Example: chitosan treatment: 1%, 1.5%, 2%

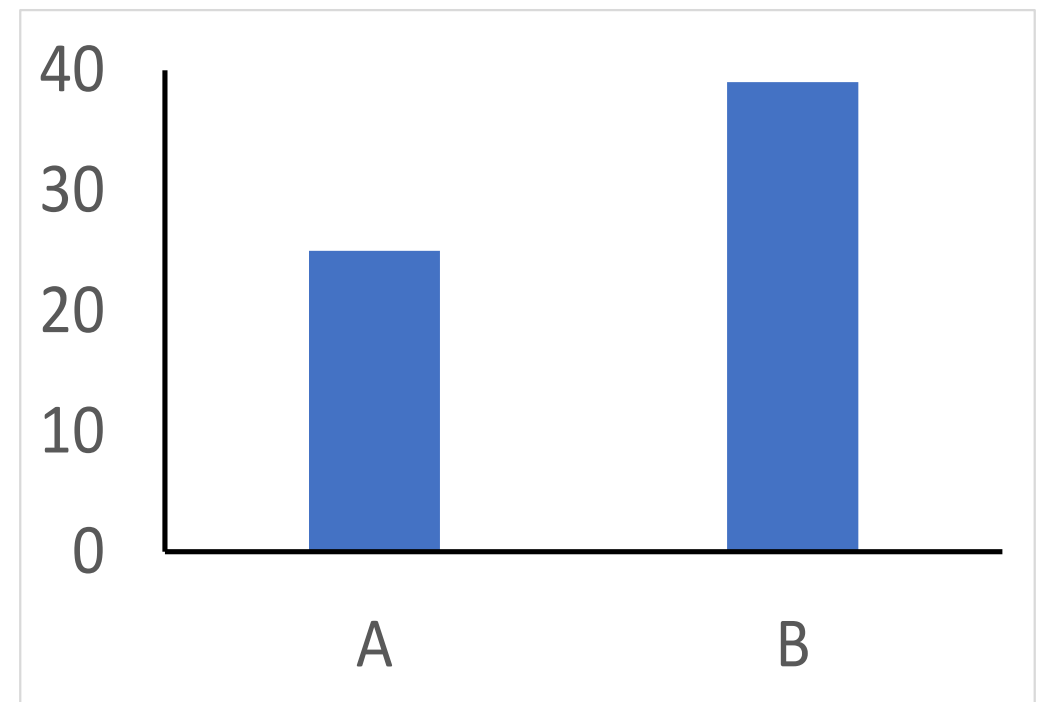
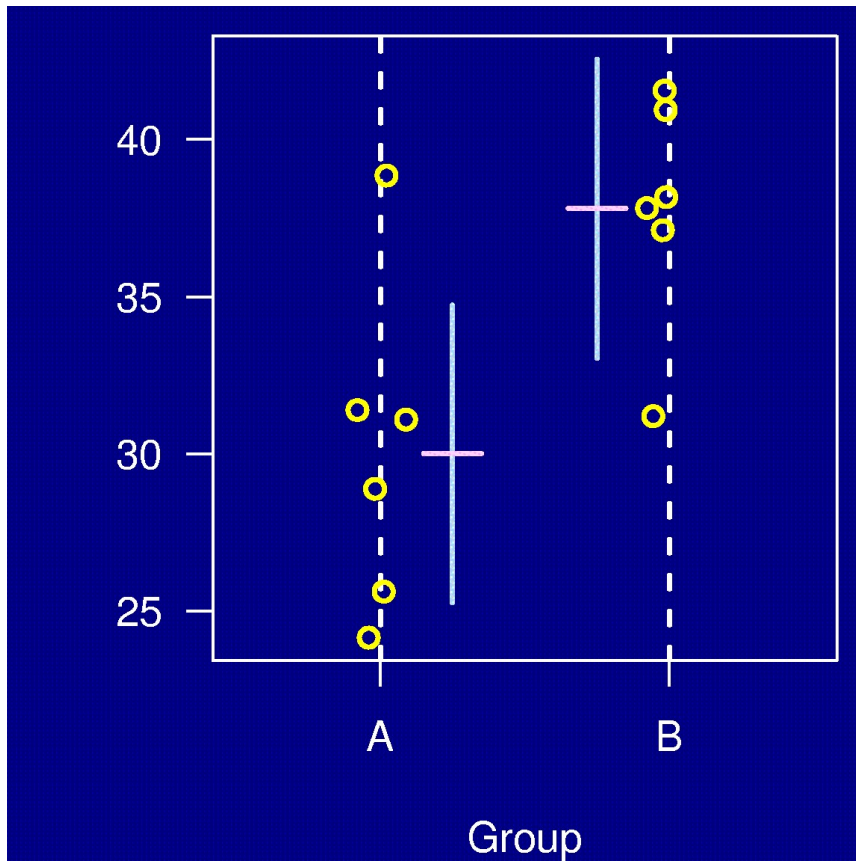
**– don't make decision about factor's effect without comparison!**

## Replication

**Example: Firmness (N) of apple (control & treated sample)**



## Why replicate?





## Replication

Group	Mean	(SEM)	Group	Mean	(SEM)
A	10.2	(0.6)	A	10.2965	(0.63)
B	12.4	(0.8)	B	12.49	(0.7913)
C	14.7	(0.6)	C	14.787	(0.6108)

## Why replicate?

- Reduce the effect of uncontrolled variation

## Randomization

Experimental subjects (samples) should be assigned to treatment groups **at random**.

Using

- A computer, or
- Coins, dice or cards.

**Why randomize?**

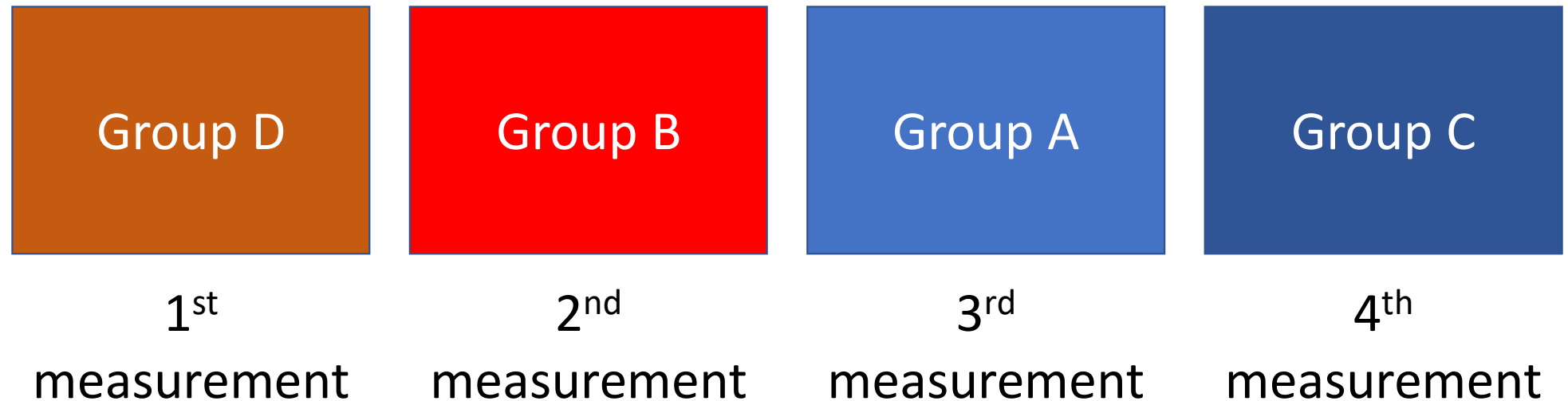
## Why randomize?

- Are the subjects you are studying really representative of the population you want to study?
- Ideally, your study material is a **random sample** from the population of interest.

## Why randomize?

Randomized design for measurements:  
in order to minimize systematic error

Randomized order / sequence:



## A Good measurement design or not?

**Day 1   Day 2   Day 3   Day 4**

G1	G1	G1	G1
G2	G2	G2	G2
G3	G3	G3	G3
G4	G4	G4	G4

G= group

## Factorial experiments

We are interested in the effect of both salt water and a high-fat diet on blood pressure.

**Good exp. design:** look at all 4 treatments in one experiment.

Plain water  
Salt water

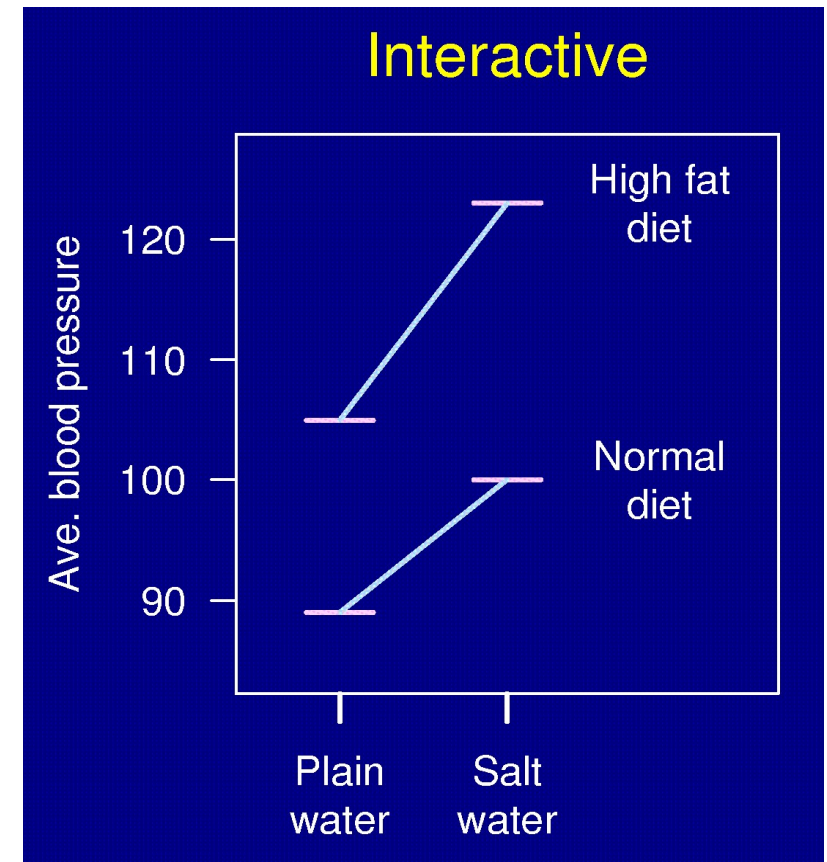
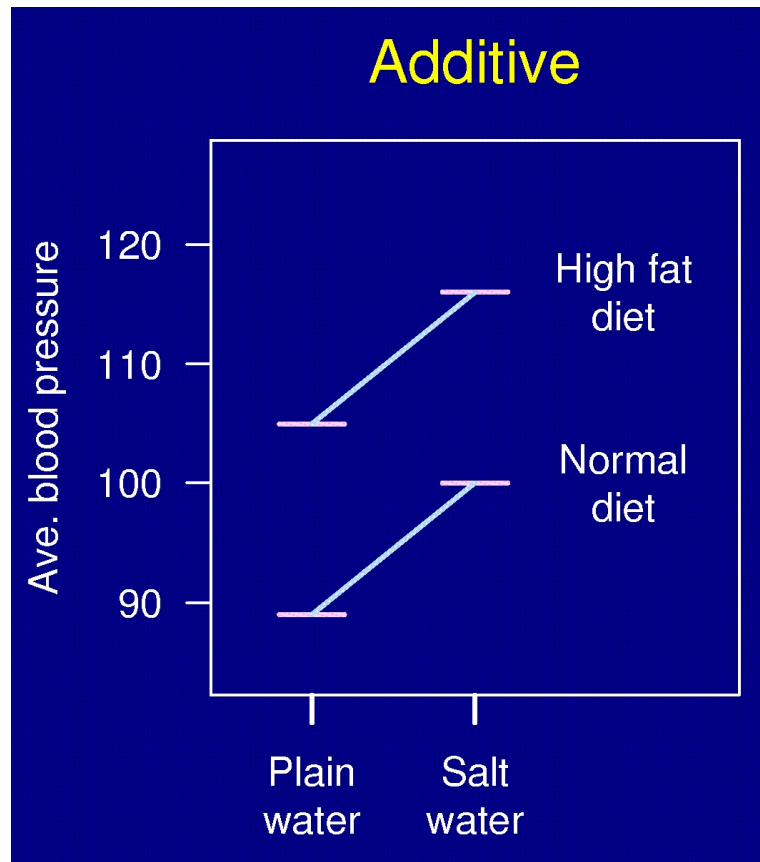
Normal diet  
High-fat diet

### Why?

- We can learn more.
- More efficient than doing all single-factor experiments.

## Factorial experiments

Interaction: if level of A changes effect of level change of B





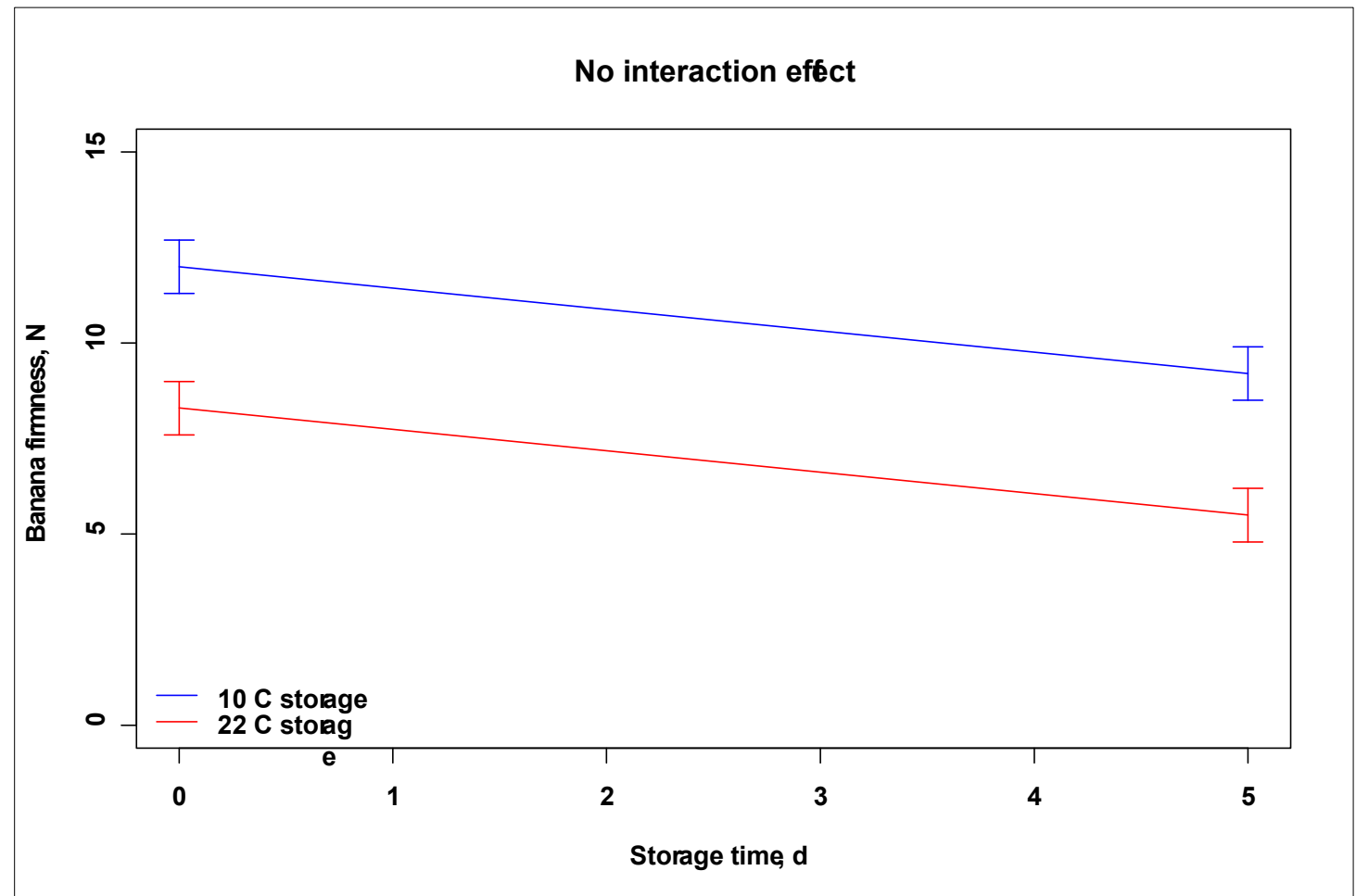
## Banana firmness

5 days storage

2 groups:

- 10 °C
- 22 °C

Cold stored pieces  
were harder.



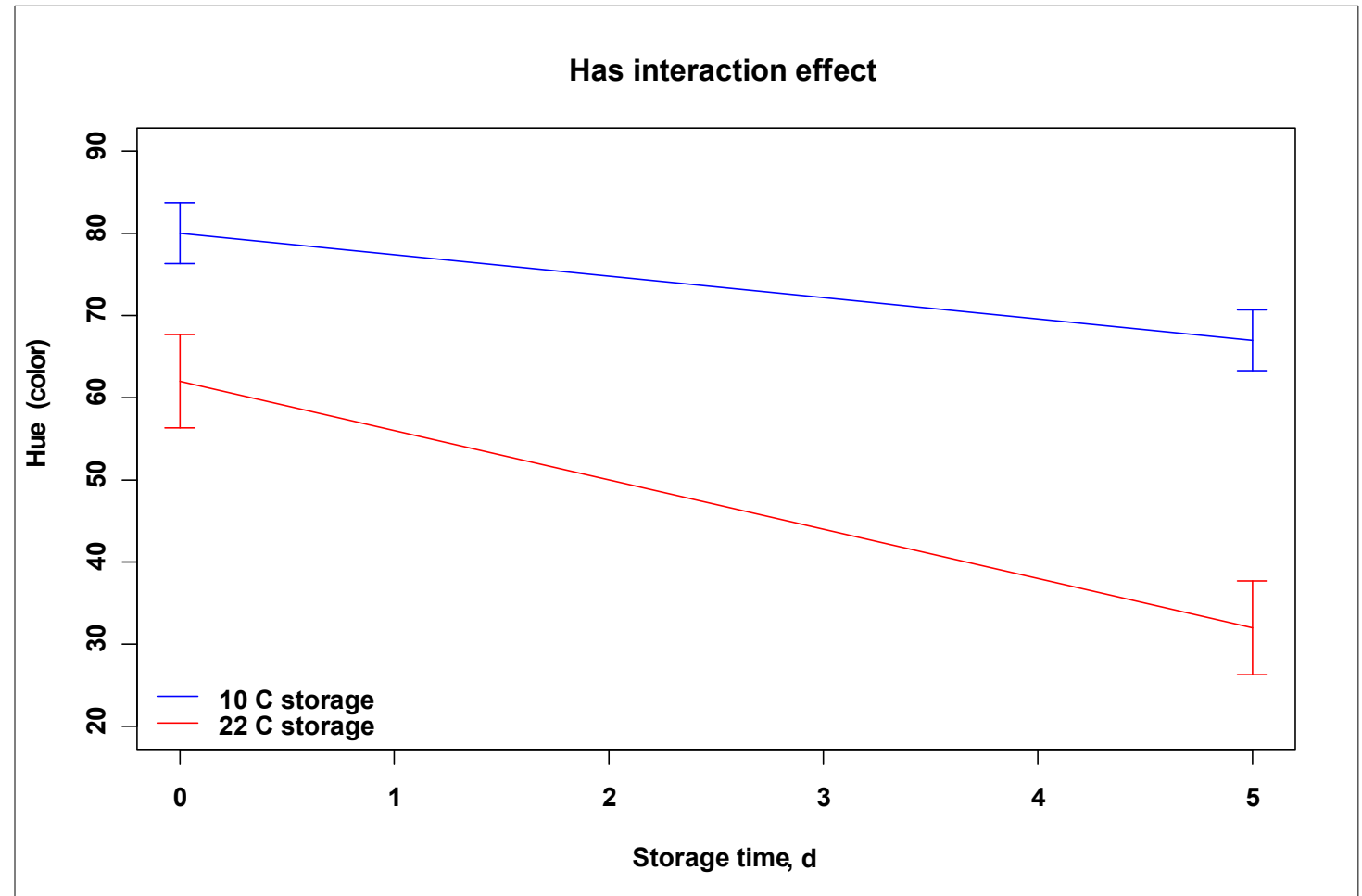
## Banana color

5 days storage

2 groups:

- 10 °C
- 22 °C

Cold stored pieces were more green, while warm pieces became brown.



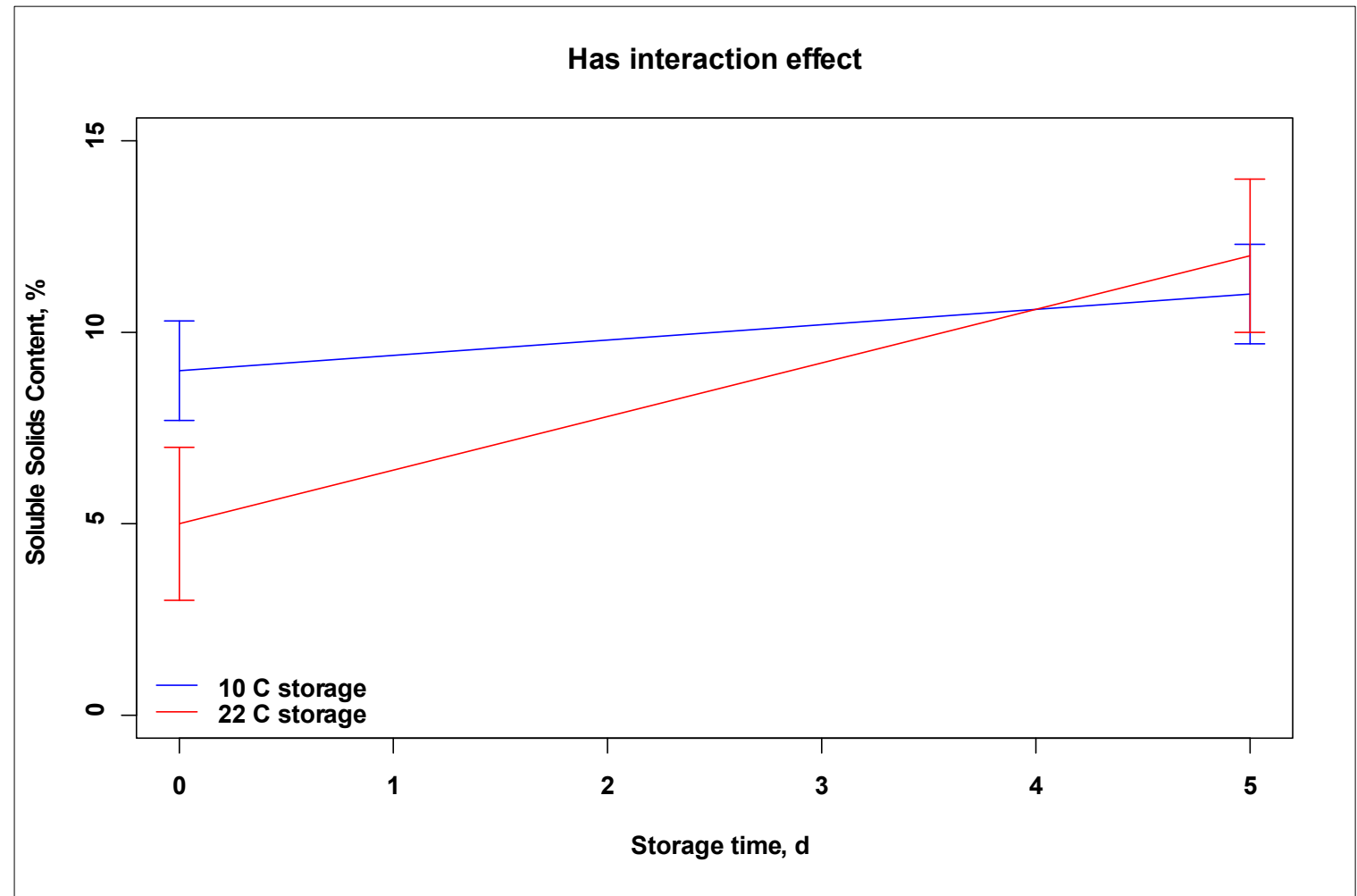
## Banana SSC

5 days storage

2 groups:

- 10 °C
- 22 °C

All pieces had similar SSC after 5 days, but warm pieces changed more.



## Summary

### Characteristics of good experiments:

- **Unbiased**
  - Randomization
  - Blinding
- **High precision**
  - Uniform material
  - Replication
  - Blocking
- **Simple**
  - Helps against mistakes



# Experimental design

## Common problems

## Common mistakes

Vary one factor at a time to see how performance changes

### Disadvantage

— if factors interact, may give wrong conclusions

### Cooking time

5 min

10 min

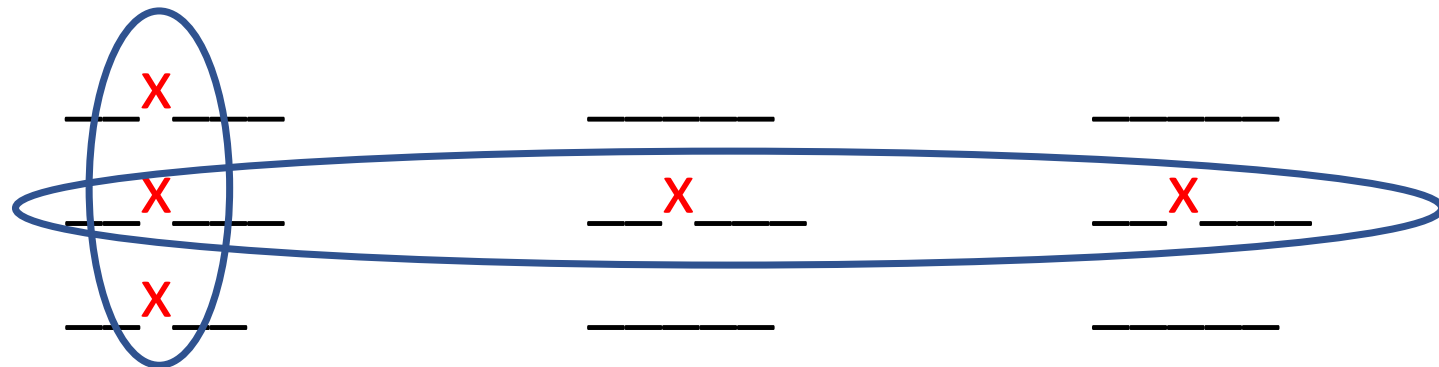
15 min

### Temperature

80 °C

100 °C

120 °C



This is called full factorial design = all combinations are measured.

## Common mistakes

Example

**Fix 2 factors, change 1 factor at a time?!**

Temperature (°C)	pH	Time (min)
50	4.0	10
60	5.0	20
70	6.0	30

Full factorial design =  $3^3$  combinations

## Common mistakes

- Use simple one-factor-at-a-time designs
  - leads to many experiments, provides too little information per experiment
- Ignore interactions between factors
  - cannot estimate interactions with one-factor-at-a-time experiments
- Conduct too many experiments
  - The number of experiments needed depends upon # factors, # factor levels



Be of simplicity, don't attempt to do too much  
write out the objectives, listed in order of priority



## Common mistakes

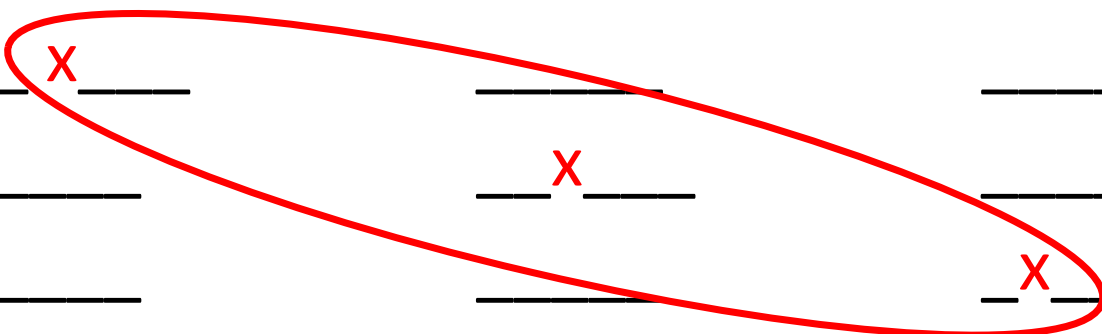
Fail to isolate effects of different factors

-if varying several factors are being varied simultaneously



Design experiments so that effects of factors can be separated

	Cooking time		
	5 min	10 min	15 min
Temperature			
80 °C	X		
100 °C		X	
120 °C			X



Explores every possible combination at all levels of factors

- Advantages

- detailed data

- can find effect of every factor, secondary factors, and interactions

- Disadvantages

- too many experiments, especially with repetitions

↓ **Solutions**

- reduce number of levels per factor (2 is very popular)

- reduce number of factors:

- initially only examine a few levels of each factor

- delete unimportant factors, then try more factors per level

- use fractional factorial designs

## What combinations are measured?

Measurement groups:

Cooking time

5 min

10 min

15 min

Temperature

80 °C

\_\_\_\_\_

\_\_\_\_\_

\_\_\_\_\_

100 °C

\_\_\_\_\_

\_\_\_\_\_

\_\_\_\_\_

120 °C

\_\_\_\_\_

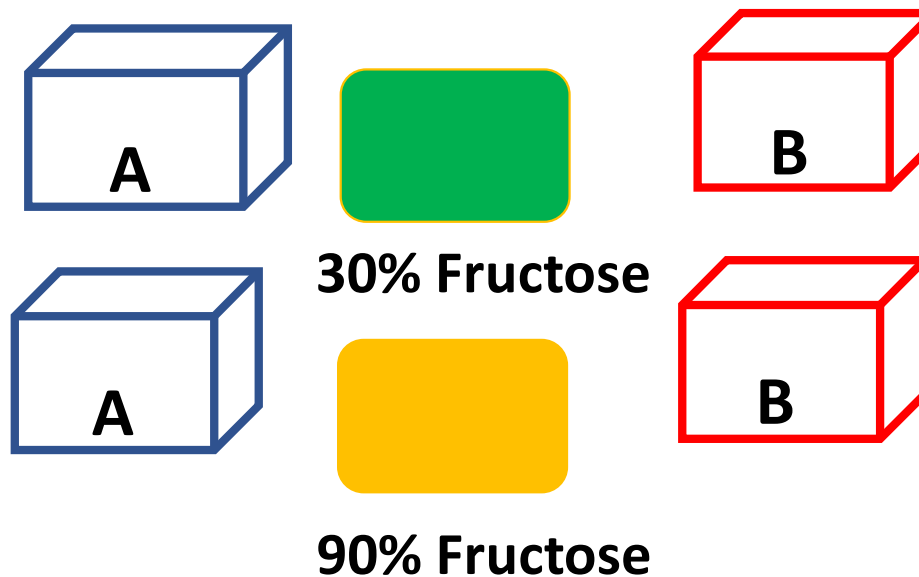
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Full factorial design = all combinations are measured.

## HOMework

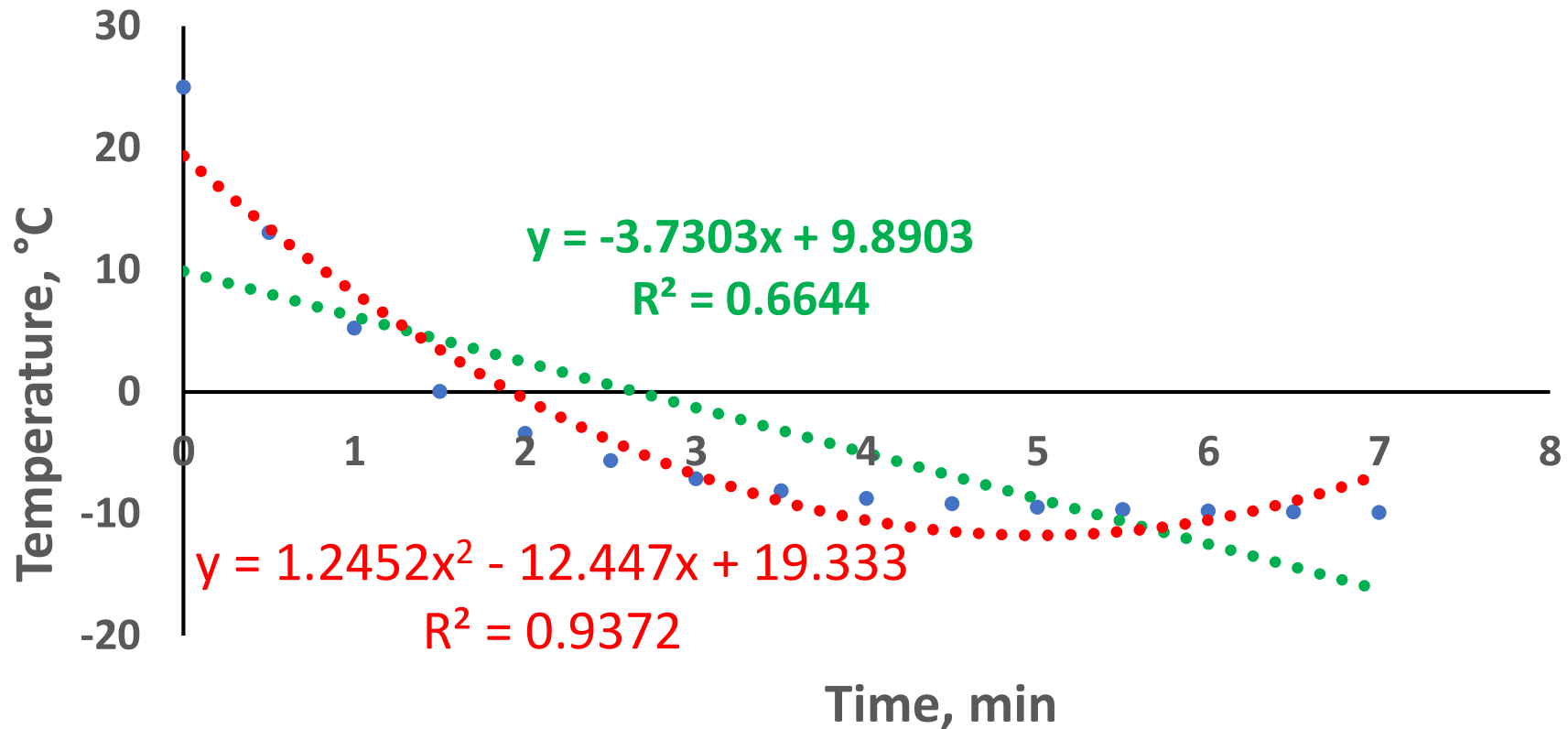
-There are 2 types of packaging: A and B and 2 types of moisture absorber. Evaluating the effect of packaging and absorber on vegetable during storage. How many groups?



## HOMEWORK

What is the equation that can describe the presented phenomenon?

Ice cream temperature in ice cream making machine



## Sources

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**THANK YOU FOR  
LISTENING!**