

B4M36SAN
BE4M36SAN
Seminar/Tutorial

Alikhan Anuarbekov

Ali

anuarali@fel.cvut.cz

- For those who was absent last week, the course webpage:
<https://cw.fel.cvut.cz/wiki/courses/b4m36san>
- If you want to have all R libraries ready for labs, download **r_setup.zip**

Materials
 san_intro.zip ,  r_setup.zip ,  pres-1.pdf

Big team project (Final assignment section in courseware)

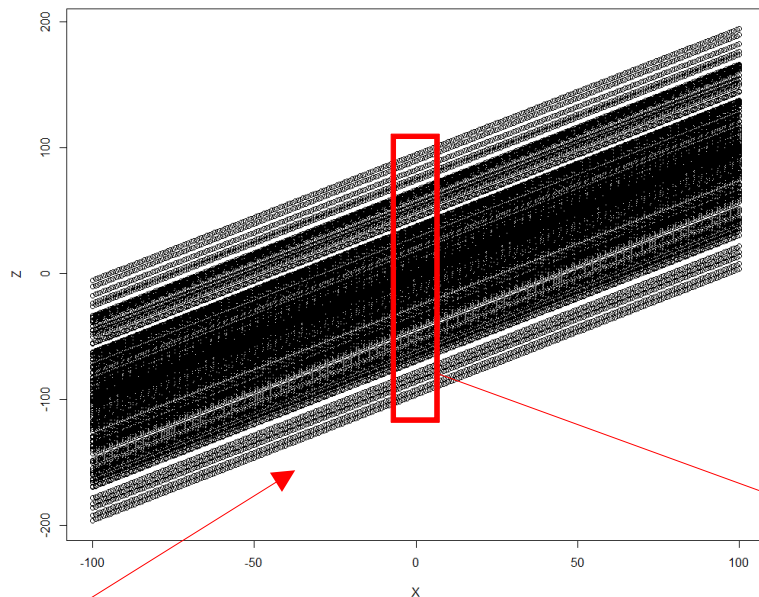
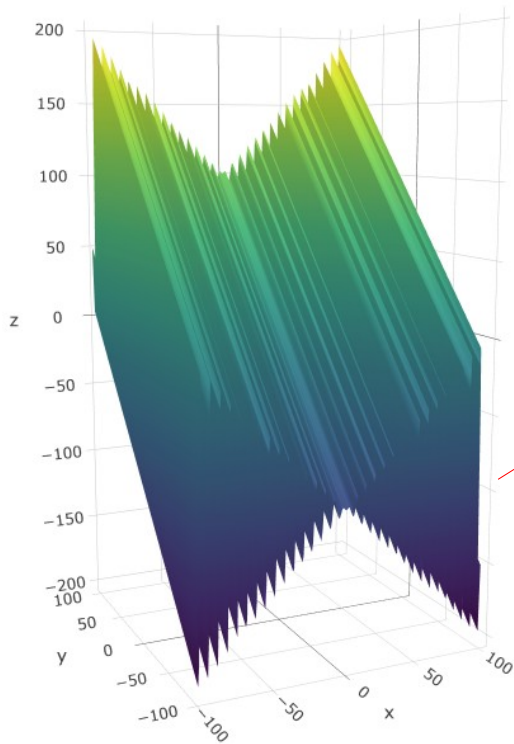
- Start looking for 4 people team **NOW**
- Try to find a question/dataset that is interesting for you
- Look for previous year best projects for inspiration
- Ask question before the deadline for team creation (3.11)
- If no team on 3.11 (midterm week), then I will forcefully assign a team for you

Tips and recommendations:

- Do not focus on single question, try to combine two or more questions
 - Example: Not only look for medical question, but try to add socio-economy
- Try to find more than one dataset
 - Example: UNICEF + World Bank
- Start working early, at least to formulate a question

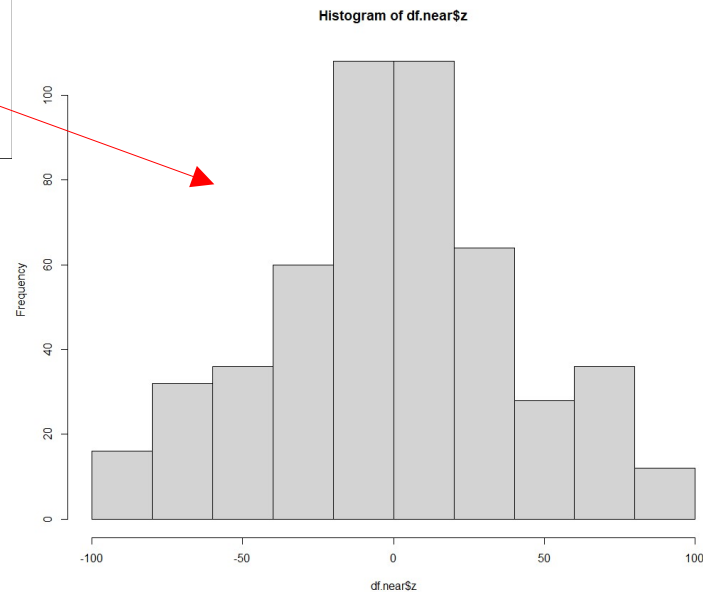
From deterministic formula to randomness (Last week)

$$Z = X + Y * \sin(Y)$$



Hide Y axis

Look at some X-Z range:



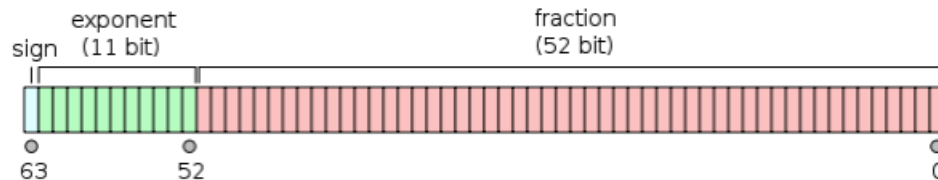
Sometimes deterministic formulas are infinitely complex

$$\sin x = x - \frac{x^3}{3!} + \frac{x^5}{5!} - \frac{x^7}{7!} + \dots \quad \text{For small } x, \sin x \approx x$$

$$\cos x = 1 - \frac{x^2}{2!} + \frac{x^4}{4!} - \frac{x^6}{6!} + \dots \quad \text{For small } x, \cos x \approx 1$$

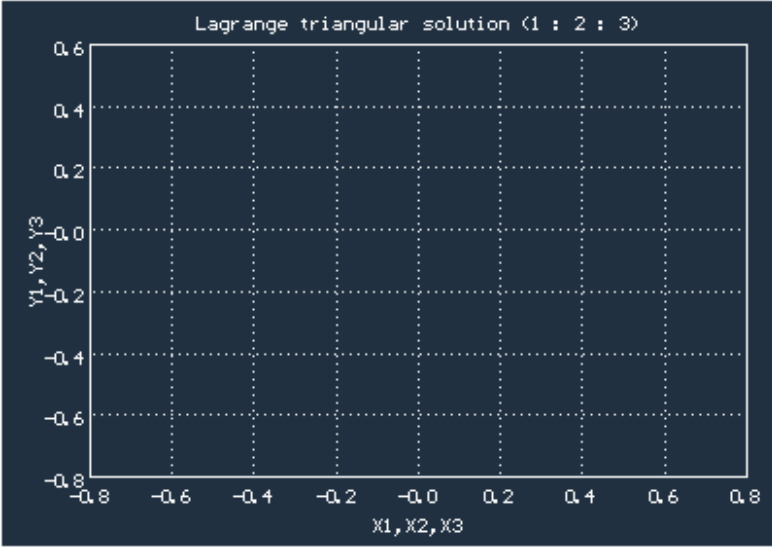
$$\tan x = x + \frac{x^3}{3} + \frac{2x^5}{15} + \frac{17x^7}{315} + \dots \quad \text{For small } x, \tan x \approx x$$

- You could compute every single digit
- But it has ∞ number of digits!
- If you cut it somewhere, you introduce an error!



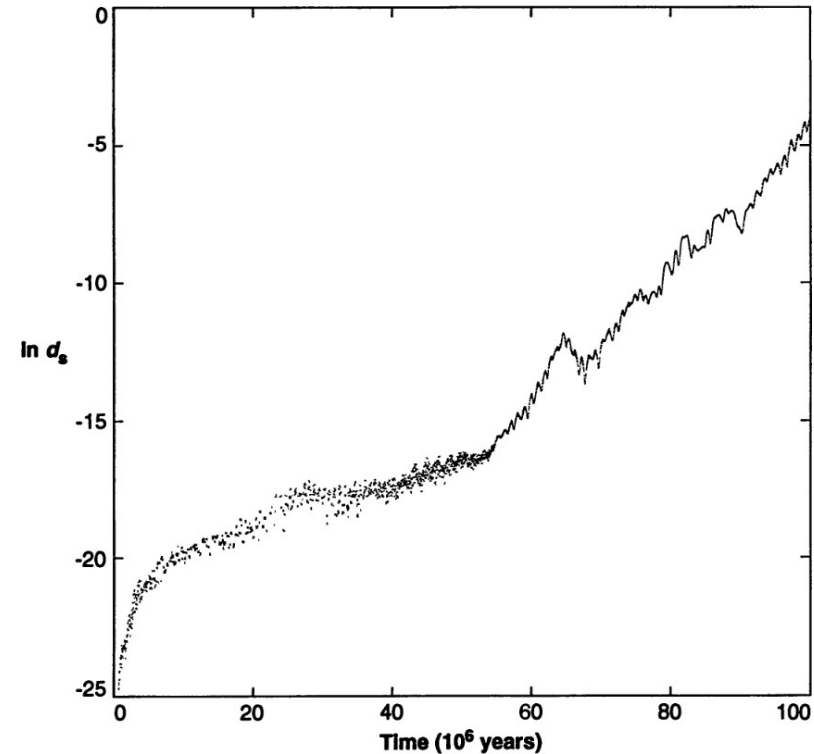
$1/3 =$ 0011 1111 1101 0101 0101 0101 0101 0101 0101 0101 0101 0101 0101 0101 0101 0101

$+ \Delta_{error}$



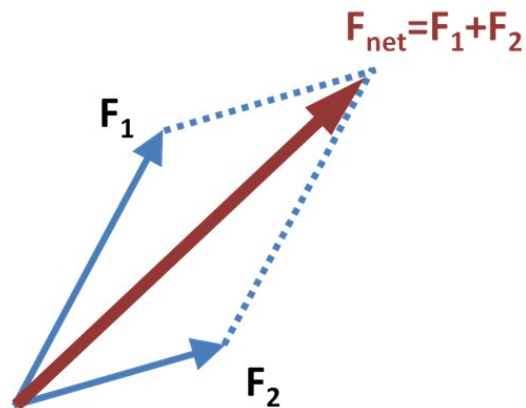
Practical case: Three body problem (Chaotic behavior)

$$\begin{cases} \ddot{\mathbf{r}}_1 = \sum_{j=1, j \neq 1}^3 m_j \frac{\mathbf{r}_j - \mathbf{r}_1}{|\mathbf{r}_j - \mathbf{r}_1|^3} \\ \ddot{\mathbf{r}}_2 = \sum_{j=1, j \neq 2}^3 m_j \frac{\mathbf{r}_j - \mathbf{r}_2}{|\mathbf{r}_j - \mathbf{r}_2|^3} \\ \ddot{\mathbf{r}}_3 = \sum_{j=1, j \neq 3}^3 m_j \frac{\mathbf{r}_j - \mathbf{r}_3}{|\mathbf{r}_j - \mathbf{r}_3|^3} \end{cases}$$



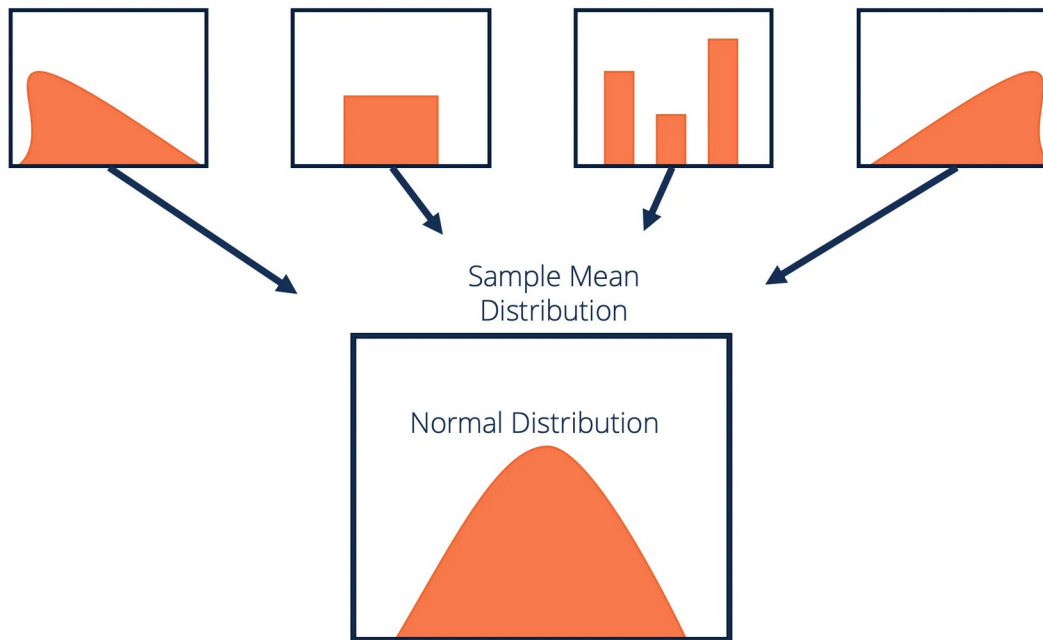
Why Gaussian normal distribution?

- Why we use a Gaussian normal? What if there is another distribution?
- Maybe, but typically when you have multiple factors/causes, they are typically summed



Example: Physics
sum of multiple forces

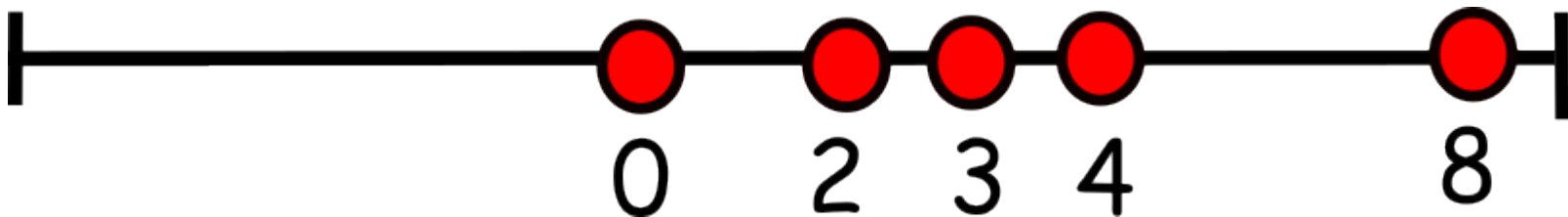
- Then, when we compute their mean (sum / number), it is mathematically Gaussian in a limit! (Central limit theorem)



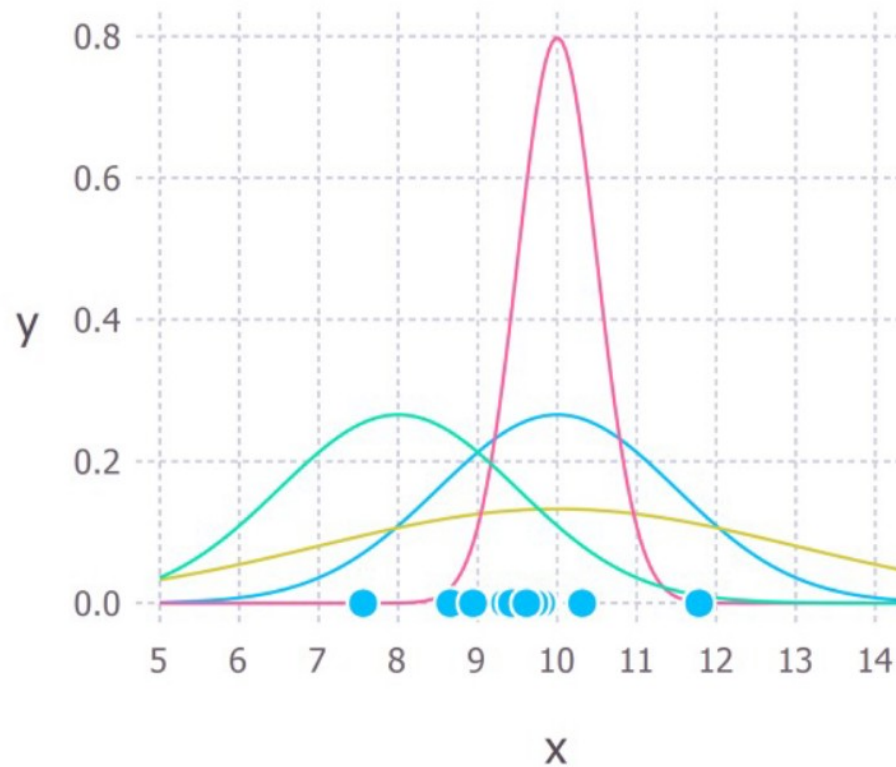
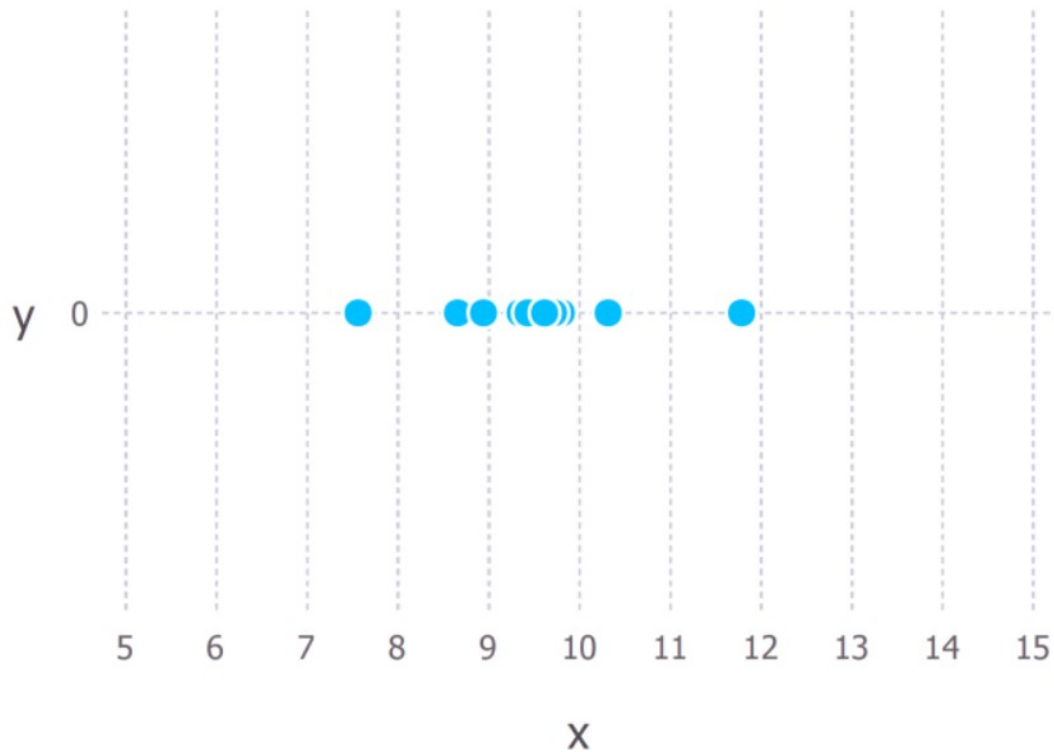
Consider the following problem:

- We get a sample of points
- We know that they come from the Gaussian Normal distribution
- But how do we estimate the parameters?

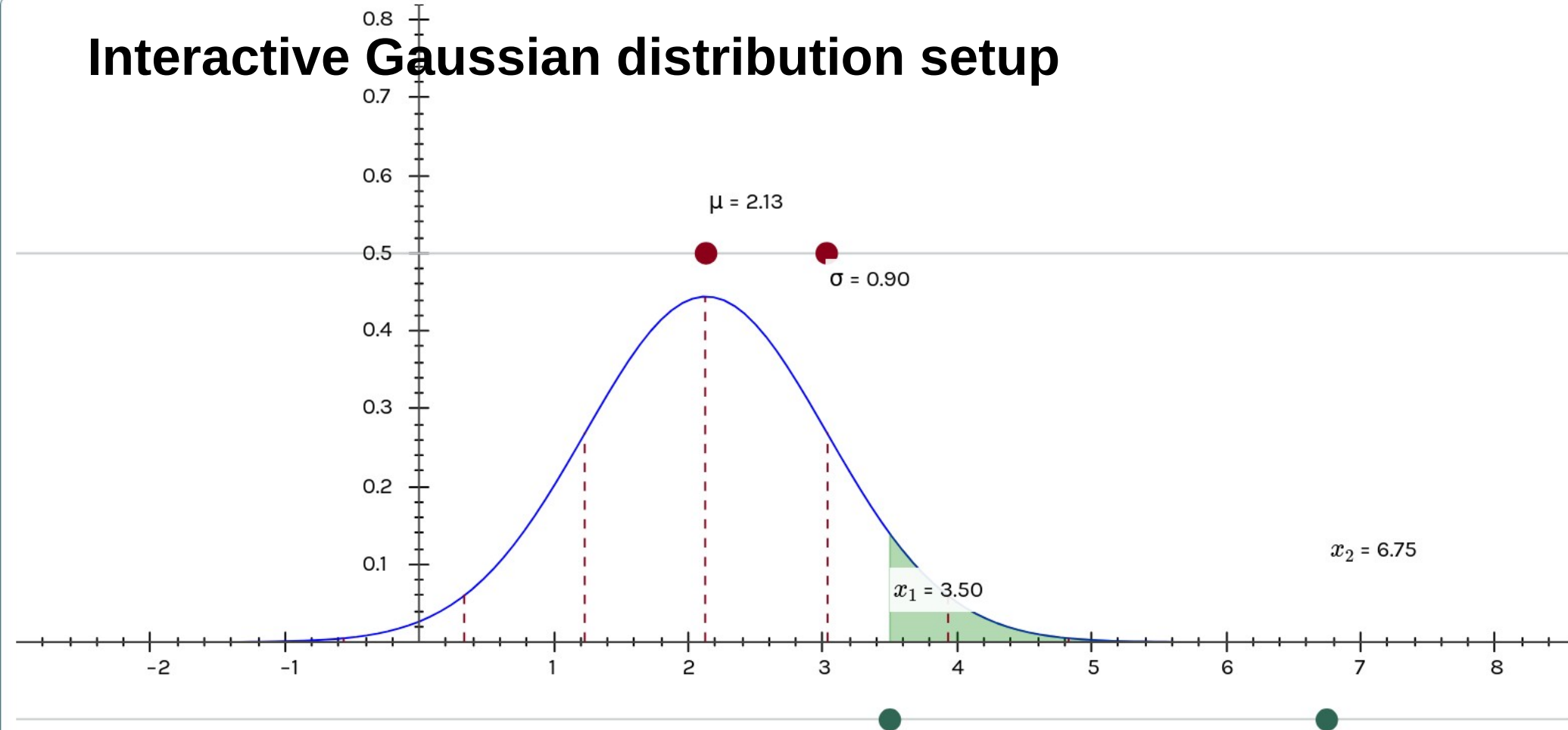
$$f(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2}$$



What value to use to measure how good each distribution is?

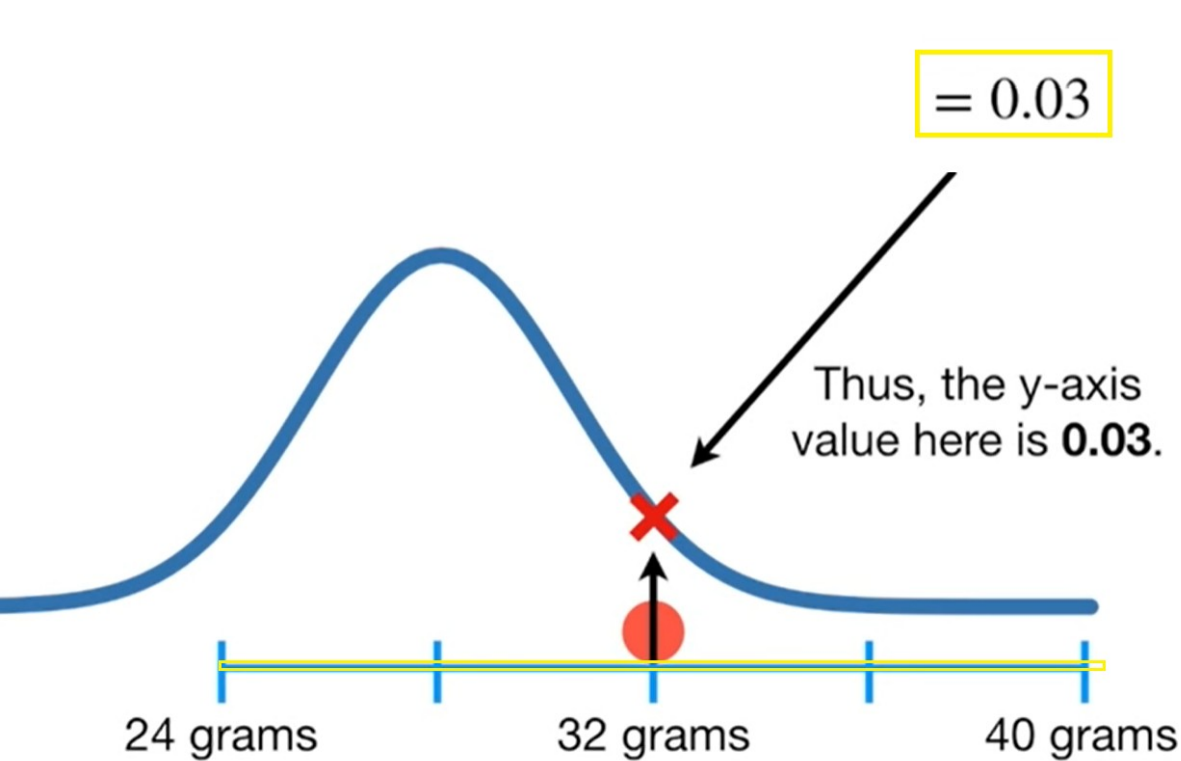


Interactive Gaussian distribution setup



<https://www.intmath.com/counting-probability/normal-distribution-graph-interactive.php>

Maximum Likelihood estimation (MLE)



$$f(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2}$$

- For every point – compute its probability based on density formula
- Multiply all point values to get the total value

$$\text{score}(\mu, \sigma) = \prod_i f(x_i | \mu, \sigma)$$

See full explanation:

https://didawiki.cli.di.unipi.it/lib/exe/fetch.php/dm/19_dm2_maximum_likelihood_estimation_2022_23.pdf

minimize $\text{score}(\mu, \sigma)$ *for* μ, σ

$$\text{score}(\mu, \sigma) = \prod_{x_1, x_2, \dots, x_m} f(x_i | \mu, \sigma)$$

$$f(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2}$$

Can anybody solve this analytically?

minimize $\text{score}(\mu, \sigma)$ for μ, σ

$$\text{score}(\mu, \sigma) = \prod_{x_1, x_2, \dots, x_m} f(x_i | \mu, \sigma)$$

$$f(x) = \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{1}{2} \left(\frac{x-\mu}{\sigma} \right)^2}$$

$$\log f(x) = -\log \sigma - \log \sqrt{2\pi} - \frac{1}{2} \left(\frac{x-\mu}{\sigma} \right)^2$$

Rewrite as:

$$\log \text{score}(\mu, \sigma) = \sum_i \log f(x_i | \mu, \sigma) = \sum_i -\log \sigma - \log \sqrt{2\pi} - \frac{1}{2} \left(\frac{x-\mu}{\sigma} \right)^2$$

Solve with first derivative conditions:

$$\frac{\partial}{\partial \mu} \log \text{score}(\mu, \sigma) = 0$$

$$\frac{\partial}{\partial \sigma} \log \text{score}(\mu, \sigma) = 0$$



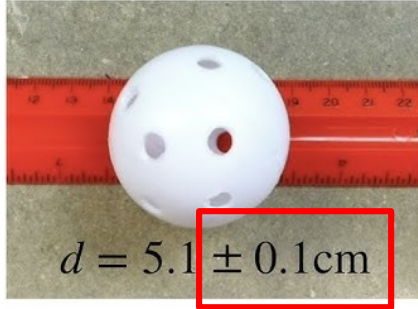
$$\mu = \frac{1}{m} \sum_i x_i$$

$$\sigma = \sqrt{\frac{1}{m} \sum_i (x_i - \mu)^2}$$

OR we can use something more familiar...

Variance is the quality of distribution for continuous variable

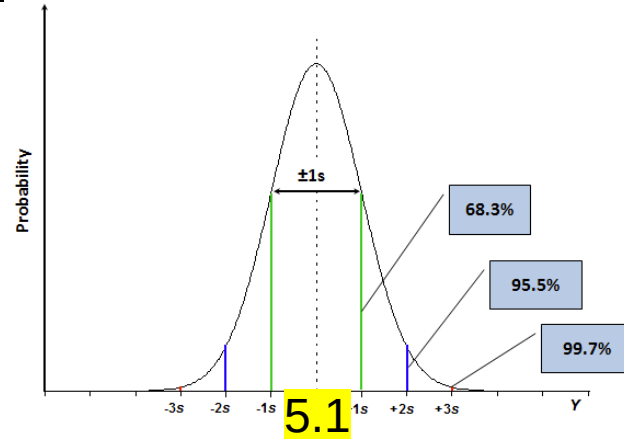
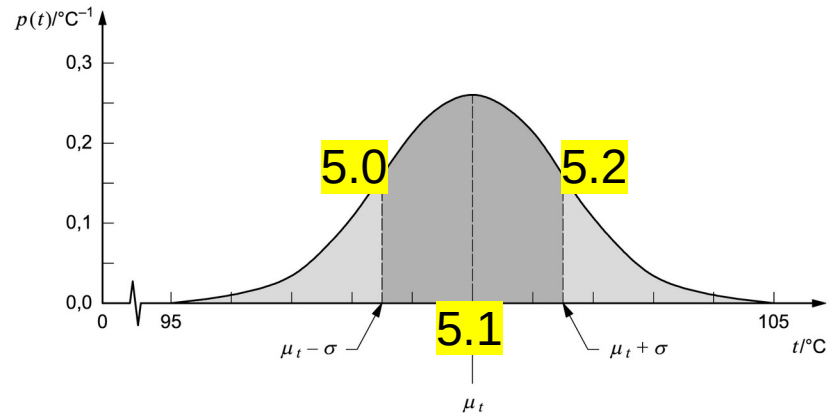
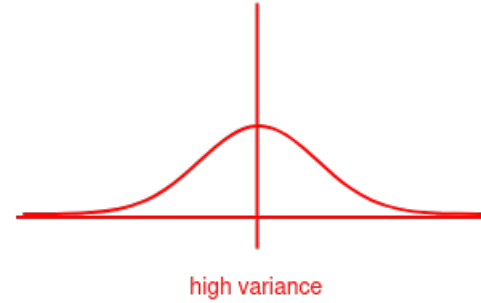
HOW DO YOU FIND
UNCERTAINTY?



Best = half smallest
Division
0.05 cm

$$d = 5.1 \pm 0.1 \text{ cm}$$

Estimate uncertainty



Minimum variance estimation

$$S = \sqrt{\frac{1}{m} \sum_{i=1}^m (x_i - \mu)^2} \quad \text{OR} \quad S^2 = \frac{1}{m} \sum_{i=1}^m (x_i - \mu)^2$$

minimize $S(\mu)$ for μ
 $x_1, x_2, \dots, x_m$

Can anybody solve this analytically?

Minimum variance estimation

$$S = \sqrt{\frac{1}{m} (x_i - \mu)^2} \quad \text{OR} \quad S^2 = \frac{1}{m} (x_i - \mu)^2$$

minimize $S(\mu)$ for μ
 $x_1, x_2, \dots, x_m$

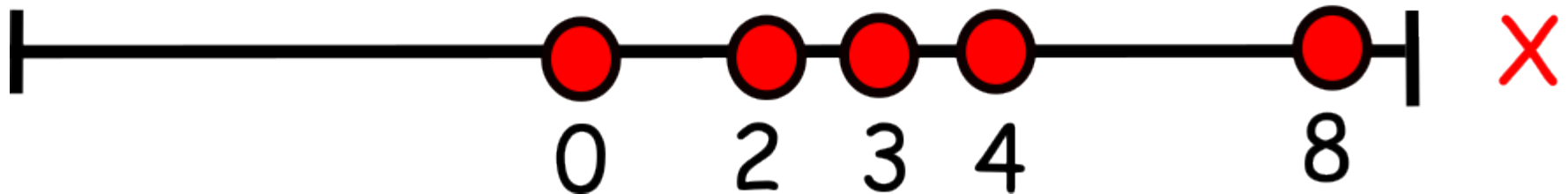
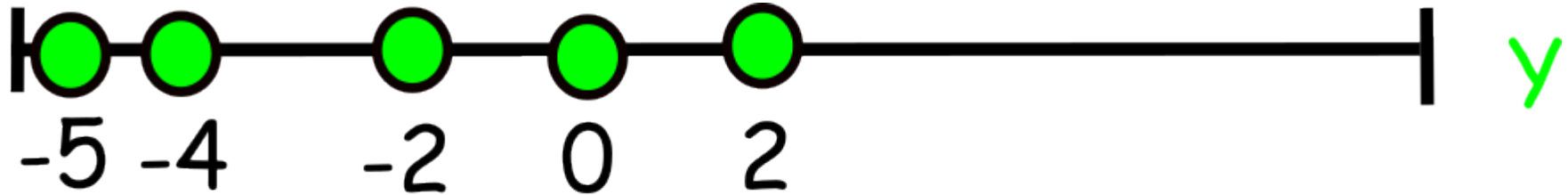
$$\begin{aligned} \frac{\partial}{\partial \mu} S^2 &= \frac{1}{m} \sum_i 2(x_i - \mu)(-1) = 0 \\ \sum_i (\mu - x_i) &= 0 \\ m\mu - \sum_i x_i &= 0 \\ \mu &= \frac{1}{m} \sum_i x_i \end{aligned}$$

Same as in MLE! But only for Gaussian Normal!
For now we will use this formula because of simplicity

Now we switch to multi-variable (multivariate) case

Consider the new problem:

- We get a sample of points for two variables
- We know that they both come from the Gaussian Normal distribution
- If they are independent, just solve every 1D problem as previously
- But what if we have a condition that they come from the same distribution?



Could you formulate the 2-variate optimization problem?

Could you formulate the 2-variate optimization problem?

$$S_X^2 = (x_i - \mu)^2 \quad \text{and} \quad S_Y^2 = (y_i - \mu)^2$$

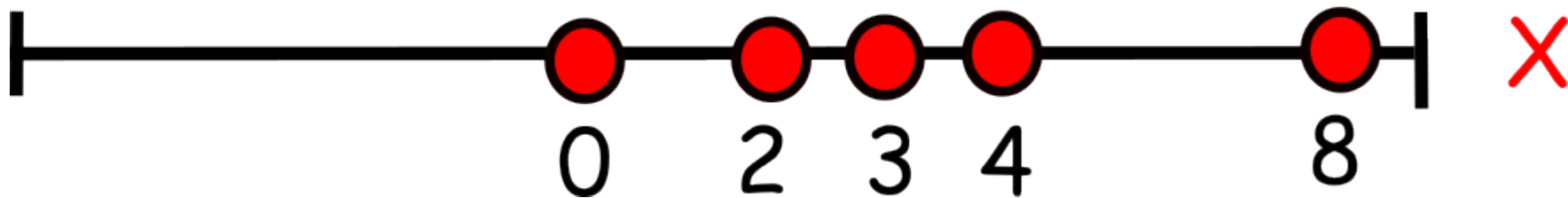
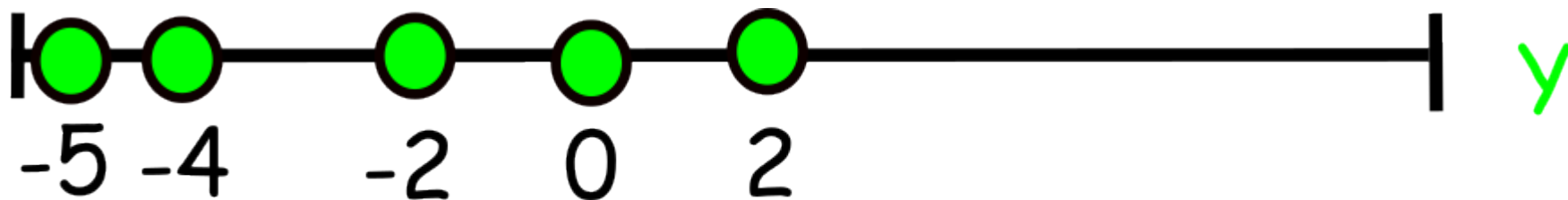
minimize $S_{total}^2(\mu) = S_X^2(\mu + 5) + S_Y^2(\mu)$ for μ
 $x_1, x_2, \dots, x_m, y_1, \dots, y_n$

$$\frac{\partial}{\partial \mu} S^2 = \sum_i 2(x_i - \mu)(-1) + \sum_i 2(y_i - \mu)(-1) = 0$$

.....

Consider a little modified new problem:

- We get a sample of points for two variables
- We know that they both come from the Gaussian Normal distribution
- If they are independent, just solve every 1D problem as previously
- But what if we have a condition that:
 - **Both share the same variance**
 - **But X has its mean shift by exactly 5 units**



Could you formulate the 2-variate optimization problem?

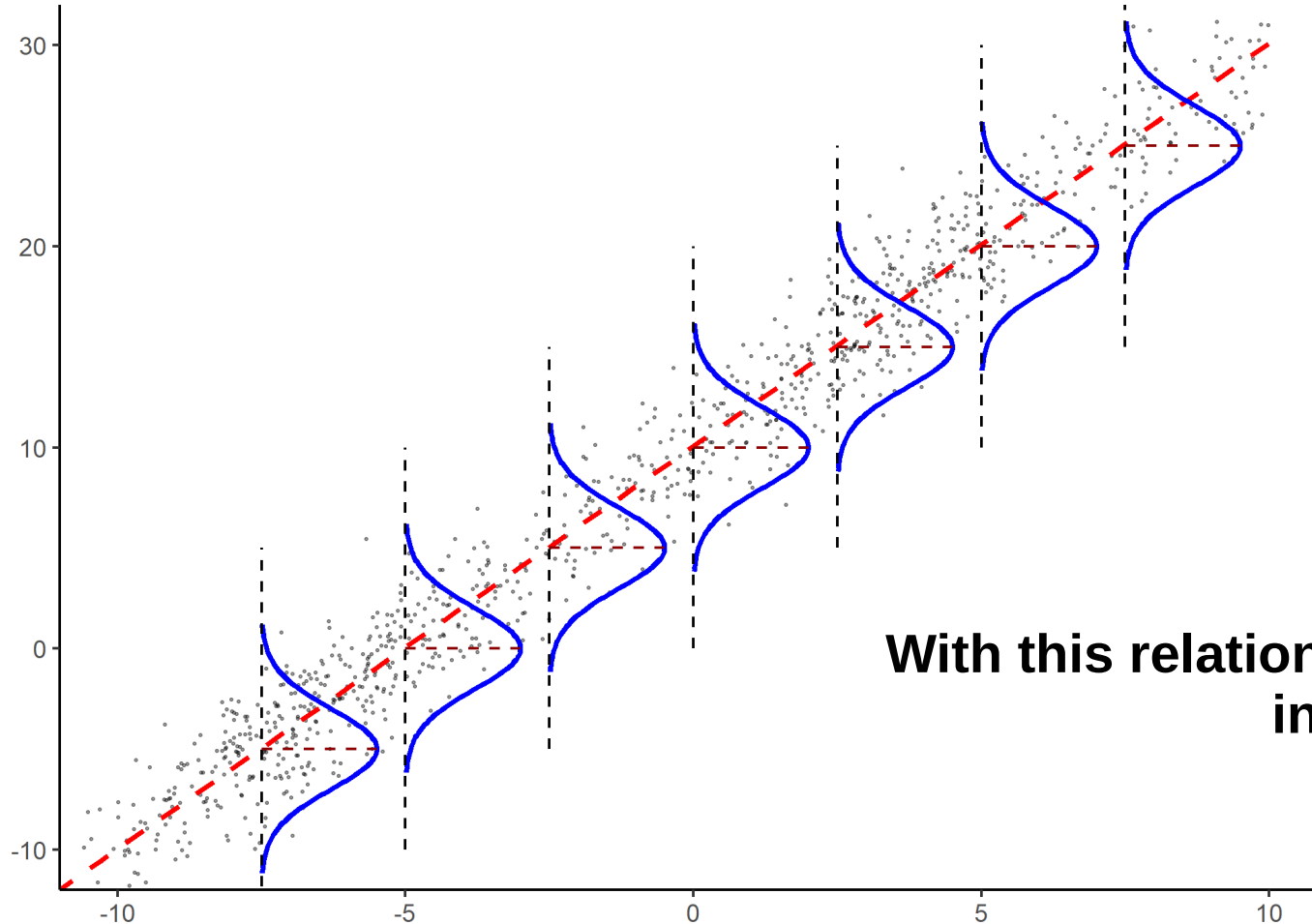
$$S_X^2 = (x_i - \mu)^2 \quad \text{and} \quad S_Y^2 = (y_i - \mu)^2$$

minimize $S_{total}^2(\mu) = S_X^2(\mu + 5) + S_Y^2(\mu)$ for μ
 $x_1, x_2, \dots, x_m, y_1, \dots, y_n$

$$\frac{\partial}{\partial \mu} S^2 = \sum_i 2(x_i - \mu - 5)(-1) + \sum_i 2(y_i - \mu)(-1) = 0$$

.....

Finally, the linear regression is just a set of such tasks



$$\mu_x = \beta_0 + X * \beta_1$$

**With this relationship between parameters
instead of +5**

And this is simply Minimum Variance objective

- we define the **residual sum of squares** (RSS) as

$$RSS = e_1^2 + e_2^2 + \cdots + e_m^2 = \sum_{i=1}^m (y_i - \hat{\beta}_0 - \hat{\beta}_1 x_i)^2$$

- the least squares approach chooses $\hat{\beta}_0$ and $\hat{\beta}_1$ to minimize the RSS

$$\frac{\partial RSS}{\partial \hat{\beta}_1} = 0 \rightarrow \hat{\beta}_1 = \frac{\sum_{i=1}^m (x_i - \bar{x})(y_i - \bar{y})}{\sum_{i=1}^m (x_i - \bar{x})^2}$$

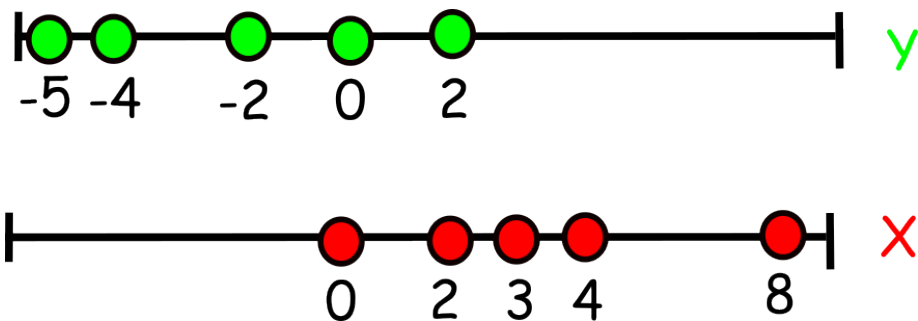
$$\frac{\partial RSS}{\partial \hat{\beta}_0} = 0 \rightarrow \hat{\beta}_0 = \bar{y} - \hat{\beta}_1 \bar{x}$$

Same derivative computation

What is the p-value of the coefficient?

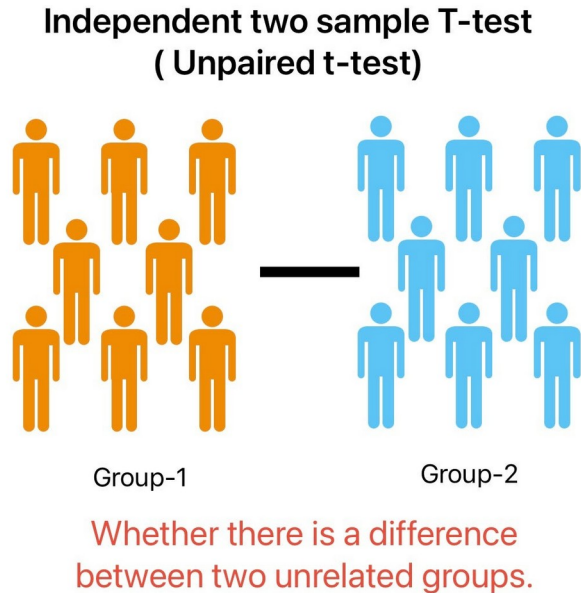
Lets recall how we defined a problem for 2 distributions:

- We get a sample of points for two variables
- We know that they both come from the Gaussian Normal distribution
- If they are independent, just solve every 1D problem as previously
- But what if we have a condition that:
 - **Both share the same variance**
 - **But X has its mean shift by exactly 5 units**
- **But what if the measured variable (whether X or Y) has no impact on the distribution?**
Is there a formal (hypothesis testing) way to compute a p-value to decide?



What is the p-value of the coefficient?

This is the same case as two sample t-test from statistics 101!



One Sample t Test

$$t_{cal} = \frac{(\bar{x} - \mu)}{s/\sqrt{n}}$$

Two Sample t Tests

❖ Is variance for two samples equal?

Yes

$$t_{cal} = \frac{(\bar{x}_1 - \bar{x}_2)}{s_p \sqrt{\left(\frac{1}{n_1} + \frac{1}{n_2}\right)}}$$
$$s_p^2 = \frac{(n_1 - 1)s_1^2 + (n_2 - 1)s_2^2}{n_1 + n_2 - 2}$$
$$df = n_1 + n_2 - 2$$

No

$$t_{cal} = \frac{(\bar{x}_1 - \bar{x}_2)}{\sqrt{\left(\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}\right)}}$$
$$df = \frac{\left[\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}\right]^2}{\frac{(s_1^2/n_1)^2}{(n_1 - 1)} + \frac{(s_2^2/n_2)^2}{(n_2 - 1)}}$$

QG

See: <https://www.qualitygurus.com/two-sample-t-test/>

See: <https://pub.aimind.so/t-test-essentials-understanding-one-sample-independent-two-sample-and-paired-test-methodologies-5dbbecd319b5>

What is the p-value of the coefficient?

- The most common hypothesis test
 - H_0 : there is no relationship between X and Y ,
 - H_A : there is some relationship between X and Y ,
- mathematically this corresponds to testing
 - $H_0 : \beta_1 = 0$ versus $H_A : \beta_1 \neq 0$,
- the test stems from the standard error and t-statistic given by

$$t = \frac{\hat{\beta}_1 - 0}{SE(\hat{\beta}_1)}$$

- the statistic will have a t-distribution with $m-2$ degrees of freedom,
- the corresponding p-value is the probability of observing any value equal to $|t|$ or larger,
- both under the H_0 assumption, i.e., assuming $\beta_1 = 0$.

What is R^2 (R-squared)?

- This is another way to measure how good we are in comparison to X-Y independent (baseline) model ($\beta_1 = 0$)
- But unlike previous p-value, this is based on metric value
- Recall the optimization problem for Variance minimization:

$$S_{x=1}^2 = \sum_i (y_{1i} - \mu)^2 \quad \text{and} \quad S_{x=2}^2 = \sum_i (y_{2i} - \mu_{x=2})^2$$

minimize $S_{total}^2(\mu) = S_{X=1}^2(\mu_{x=1}) + S_{x=2}^2(\mu_{x=2})$ for $\mu_x = \beta_0 + \beta_1 X$

$x=1 \rightarrow y_{11}, y_{12}, \dots$

$x=2 \rightarrow y_{21}, y_{22}, \dots$

What is R^2 (R-squared)?

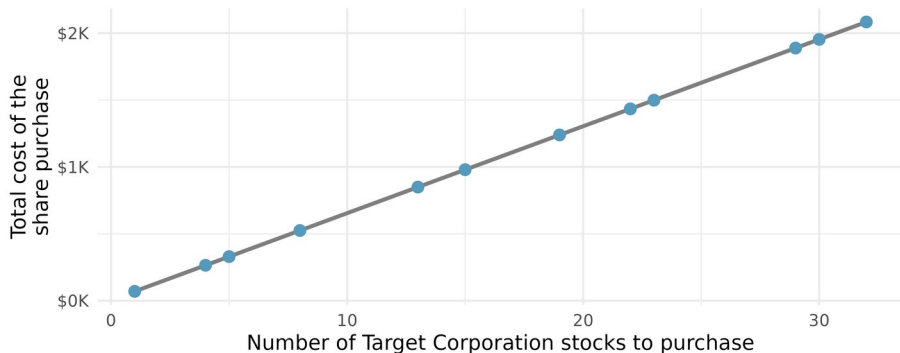
- This is another way to measure how good we are in comparison to X-Y independent (baseline) model ($\beta_1 = 0$)
- But unlike previous p-value, this is based on metric value
- We start with baseline model ($\beta_1 = 0$), it has a certain value of

$$S_{total}^2(\beta_0, \beta_1=0)$$

- We want to improve it for our derivative-based solution:

$$S_{total}^2(\beta_0, \hat{\beta}_1)$$

- And finally, if we were to ideally solve the problem:



$$S_{total}^2(perfect)=0$$

What is R^2 (R-squared)?

- This is another way to measure how good we are in comparison to X-Y independent (baseline) model ($\beta_1 = 0$)
- RSS = our variance
- TSS = variance of baseline model
- Decrease of minimization value in percentage

■ **R-squared** gives the fraction of variance explained by the model

$$R^2 = \frac{TSS - RSS}{TSS} = 1 - \frac{RSS}{TSS}$$

- where $TSS = \sum_{i=1}^m (y_i - \bar{y})^2$ stands for the total sum of squares,
- and $RSS = \sum_{i=1}^m (y_i - \hat{y}_i)^2$ stands for the residual sum of squares,

What is F-statistics and how to evaluate a total model?

- It is R-squared converted to hypothesis format with p-value
- Simply speaking – if $p\text{-value} < 0.05$, your model is useful

■ What would you say about the following model?

```
summary(lm(y ~ x,d))
```

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	3.0017	1.1239	2.671	0.02559	*
x	0.4999	0.1178	4.243	0.00216	**

>= 1 star is GOOD

Residual standard error: 1.236 on 9 degrees of freedom

Multiple R-squared: 0.6667, Adjusted R-squared: 0.6297 **> 0.5 is GOOD**

F-statistic: 18 on 1 and 9 DF, p-value: 0.002165 **< 0.05, GOOD**

Questions?

Rstudio/R basic programming

LOGICAL -> INTEGER -> NUMERIC -> CHARACTER

“boolean” “integer” “double” “string”

a = 1

class(a) = type of variable

vector = array of same type

....

https://www.w3schools.com/r/r_get_started.asp

<https://www.datacamp.com/courses/free-introduction-to-r>

Download today's tutorial script:

2	29.9.	JB, AA, JK	Simple linear regression.	 san_lreg.zip ,  pres_2.pdf
---	-------	---------------	---------------------------	---

Vector



- 1 column or row of data
- 1 type (numeric or text)

List



- 1 column or row of data
- 1 or more types

Matrix



- multiple columns and/or rows of data
- 1 type (numeric or text)

Data Frame



- multiple columns and/or rows of data
- multiple types

	dma	day	amount_of_deaths	new_deaths	weekly_deaths	gtrends
66	1	2021-05-02	66247	47	491	62
67	1	2021-05-09	66690	41	443	62
68	1	2021-05-16	67030	43	340	55
7481	2	2020-02-02	0	0	0	68
7482	2	2020-02-09	0	0	0	66
7483	2	2020-02-16	0	0	0	68
7484	2	2020-02-23	0	0	0	65
7485	2	2020-03-01	0	0	0	66
7486	2	2020-03-08	0	0	0	66
7487	2	2020-03-15	1	0	1	68
7488	2	2020-03-22	12	3	11	77
7489	2	2020-03-29	56	6	44	79
7490	2	2020-04-05	184	16	128	100
7491	2	2020-04-12	397	38	213	96
7492	2	2020-04-19	777	25	380	77
7493	2	2020-04-26	1170	21	393	77
7494	2	2020-05-03	1559	20	389	73
7495	2	2020-05-10	1945	21	386	69

Showing 66 to 84 of 14,212 entries, 6 total columns

<----- View(myDataFrame)

myDataFrame\$new_deaths = access column
as vector

$$H = \beta_0 + \beta_1 \cdot V + \beta_2 \cdot K + \epsilon, \quad \epsilon \in N(0, \sigma^2)$$

can be denoted shortly as (R notation): $H \sim V + K$

```
lm.fit <- lm(medv ~ lstat, data = Boston)
```

Also a useful notation: $\mathbf{H} \sim \blacksquare$ (H = regression of all other variables)

If you are not sure about something about function/data:

help(lm)
?lm

help(Boston)
?Boston

Today's homework (voluntary) for 1 point

2

29.9.

JB, AA,
JK

Simple linear regression.



 pres_2.pdf

Deadline: 06.10.2025 (next seminar)

Goals:

- *) Try to write some simple R code
- *) Understand basic concepts regarding the linear regression
- *) You could still answer with pseudocode and theoretical answers if you do not want to use R

