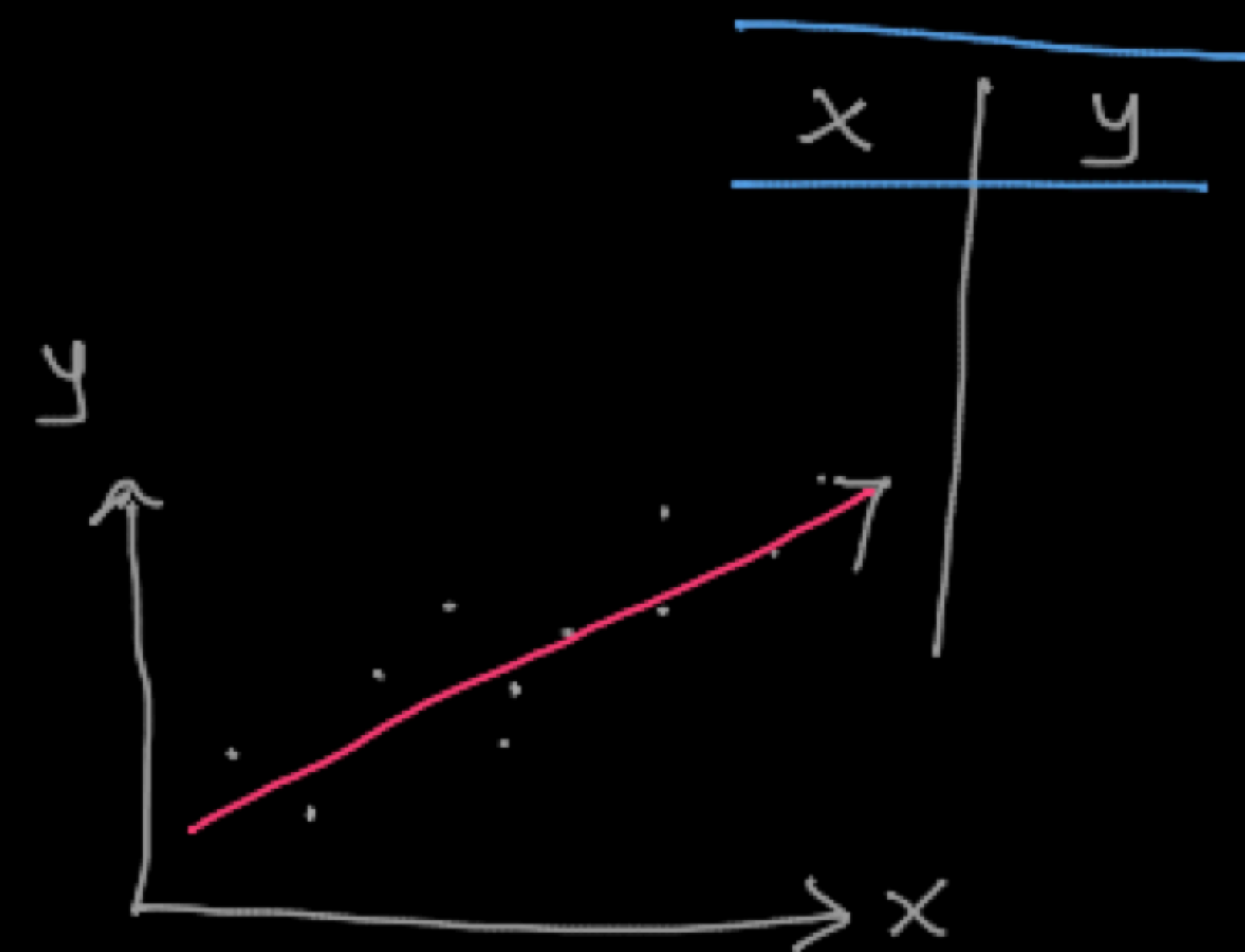


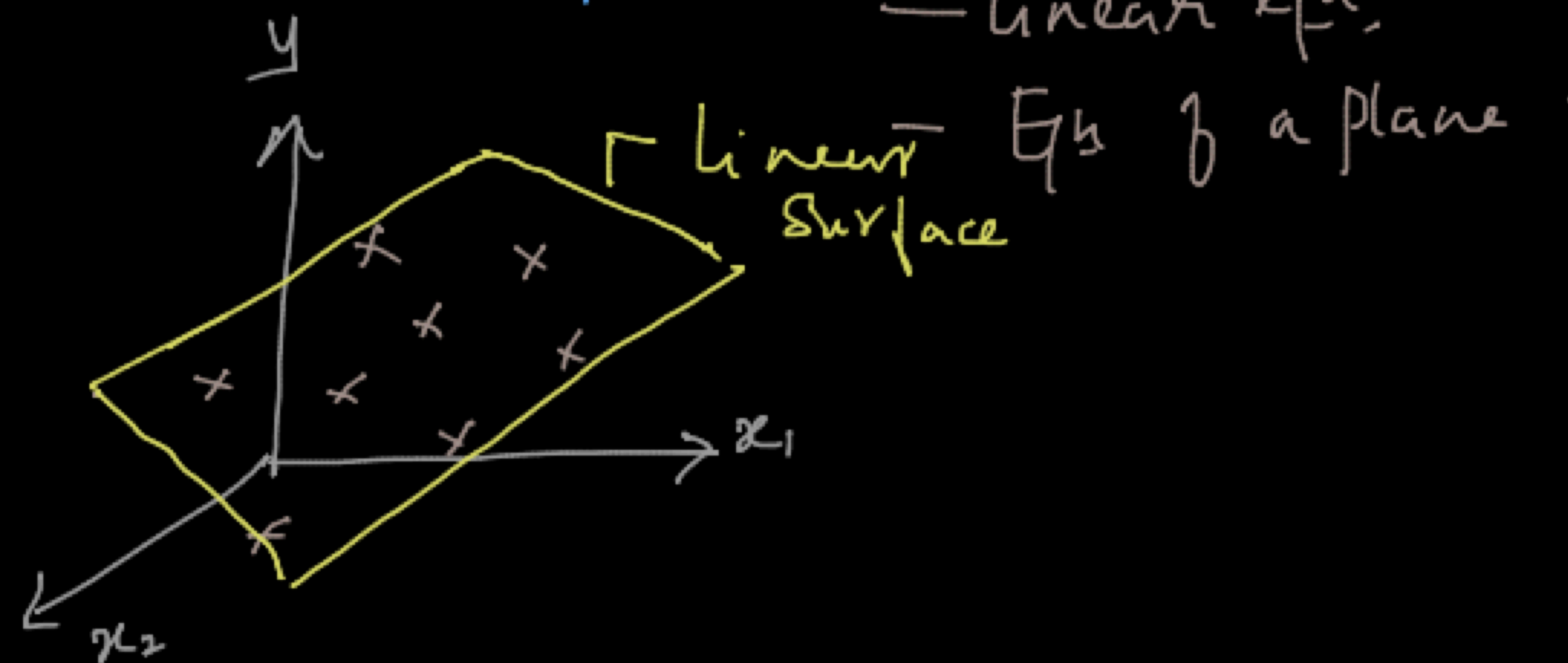
Multiple x -value \leftarrow Multi linear Regression \rightarrow Continuous y



$\hat{y} = \beta_0 + \beta_1 x$
 \rightarrow Eqn of a st. line.

x_1	x_2	y

$\hat{y} = \beta_0 + \beta_1 x_1 + \beta_2 x_2$
 \rightarrow linear eqn.



$\hat{y} = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \dots + \beta_d x_d$
 \rightarrow linear Eqn.

\leftarrow d-dimensional data \rightarrow

x_1	x_2	x_3	\dots	x_d	y

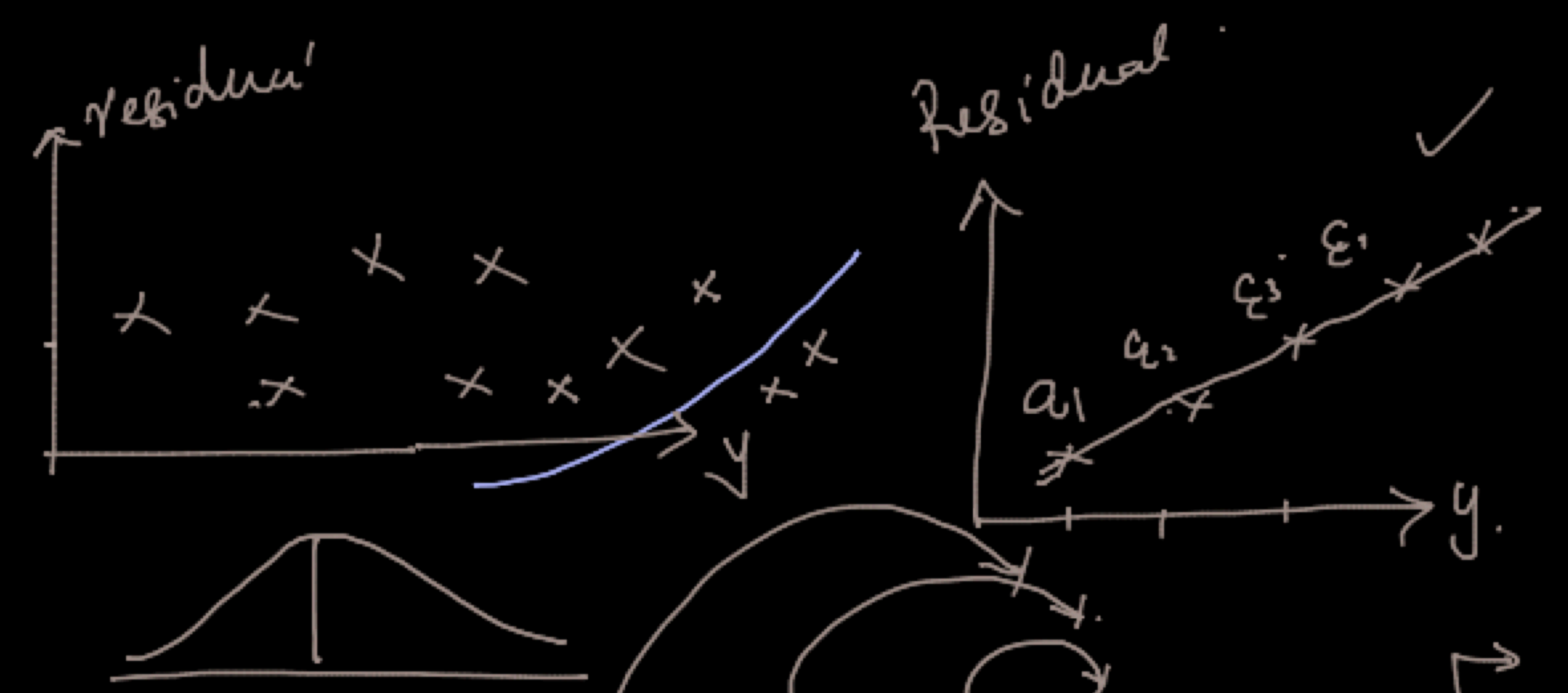
$\hat{y} = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \dots + \beta_d x_d$

- \rightarrow linear eqn
- \rightarrow linear surface.
- \rightarrow Hyperplane.

Assumptions:

1. Assumption on the parameters.

$$\hat{y} = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \dots$$



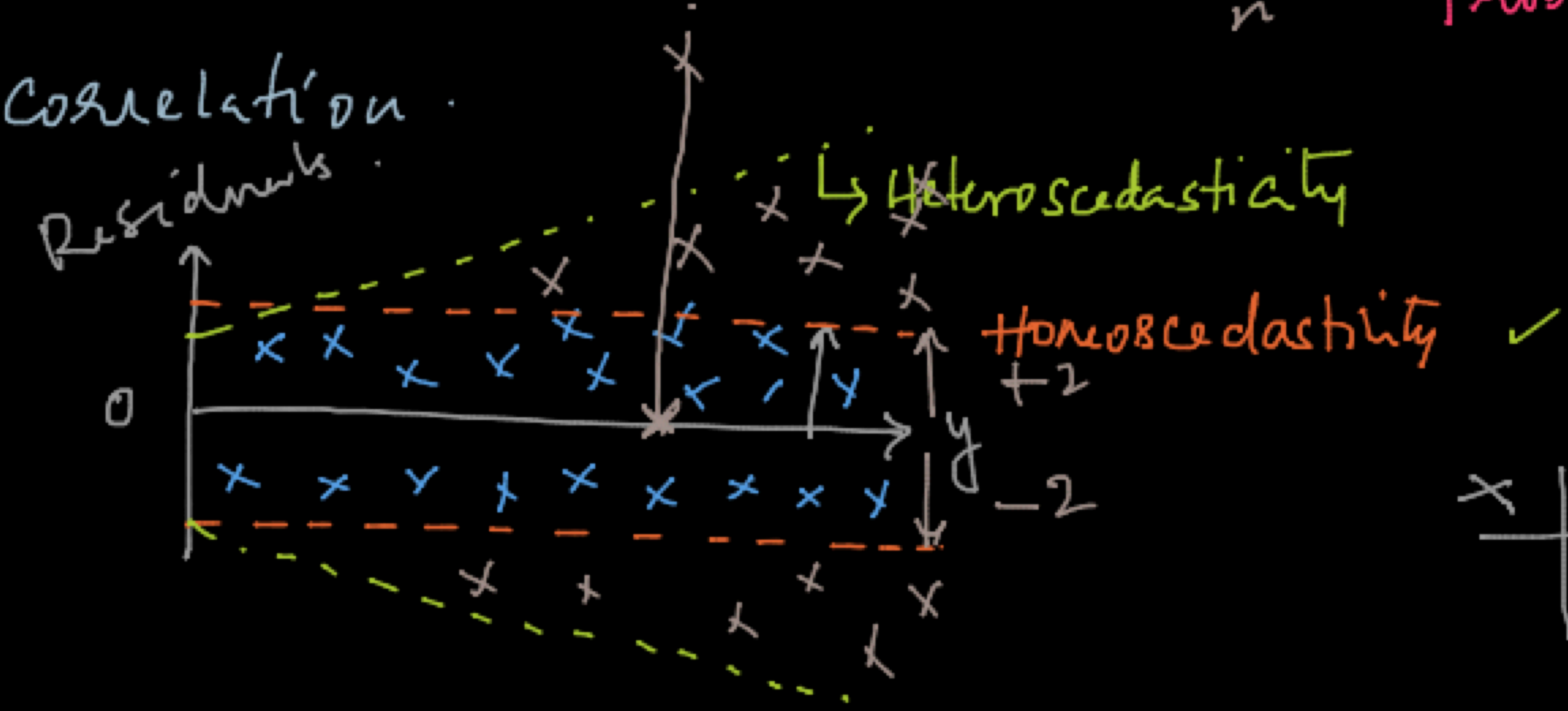
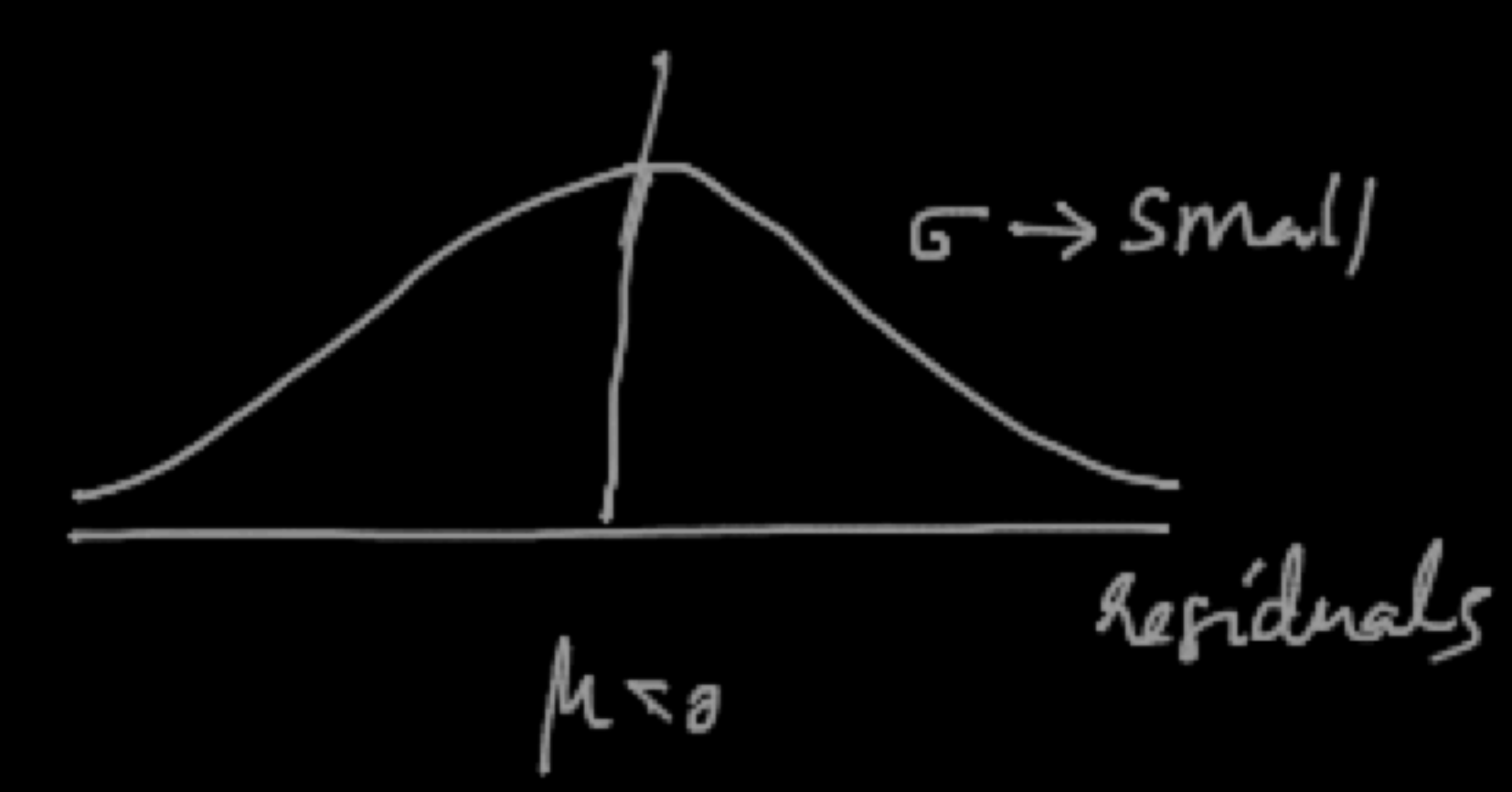
2. Residuals

- a. The mean should be zero.
- b. The residuals should be ND (random)
- c. There should be constant variance. (Homoscedasticity)
- d. No autocorrelation.

	x_1	x_2	x_3	y	\hat{y}	$y - \hat{y}$
1	-	-	-	-	-	$+ve \epsilon_1$
2	-	-	-	-	-	$-ve \epsilon_2$
3	-	-	-	-	-	$+ve \epsilon_3$
\vdots						$-ve$
n						$-ve$

Residuals

Multi collinearity Problem.



$\pm 2\sigma$

$x | y$

Errors
 ϵ_1
 ϵ_2
 ϵ_3

3. Independent Variables

— the x-values should be independent of each other

x	y	\hat{y}	Error
1200	53	55	-2
1500	75	80	-5
1800	86	100	-14

y	\hat{y}	Error
53	55	-2
75	73	+2
86	87	-1
93	94	-1
		-2
		+2.6
		-2.5

→ 24

