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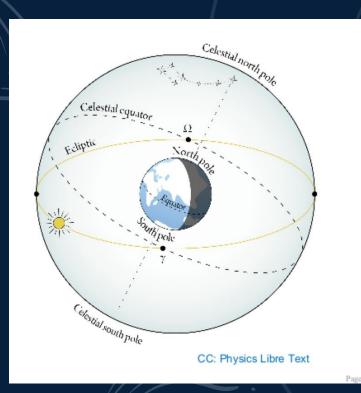
What's the project all About?
This is an engine for amateure astronomers by using it they can learn how to track celestial objects, predict astronomical events using previous data & filter astronomical images.

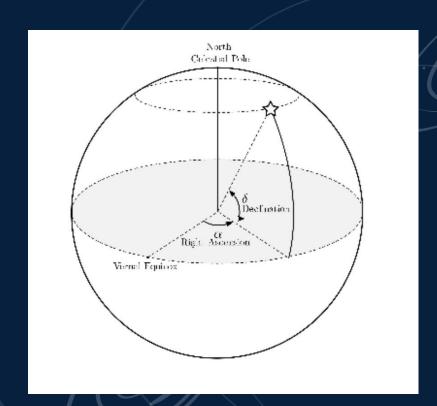
**Project motivation:** Computational astronomy is a rapidly expanding field. NASA, NRAO and numerous other institutes have begun to open-source their datasets for research purposes. However, astronomical data-analysis software can be quite complex for young learners. It's very challenging for them to understand and modify these tools for their use. In this project, I have implemented simplified algorithms to facilitate young learners' comprehension of how these systems function, enabling them to explore and learn about computational astronomy more easily.

### Key Features:

- Gradient descent
- Image Stacking
- Spherical coordinate Calculator

### Coordinate Calculations in Celestial-sphere:

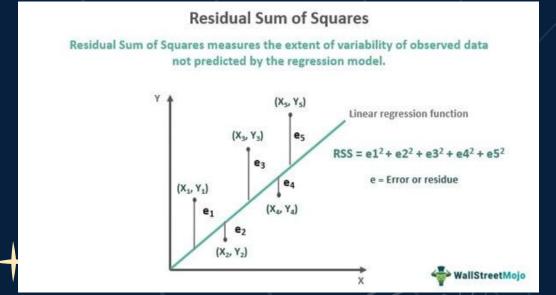






# Regression: A regression is a statistical technique that relates a dependent variable to one or more independent (explanatory) variables

RSS:The residual sum of squares measures the level of variance in the error term, or residuals, of a regression model.



## How can we get the best fit model?

By picking the coefficients of the parameters for which RSS gives the minimum value.

## Gradient Descent:

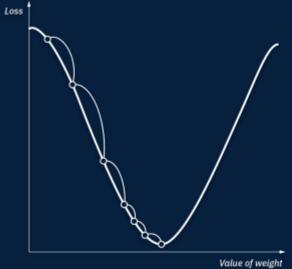
Gradient descent is an optimization algorithm that minimizes a function by iteratively adjusting the model parameters in the direction of the steepest decrease in the function's value.

$$\theta_j = \theta_j - \alpha \frac{\partial}{\partial \theta_j} J(\theta)$$

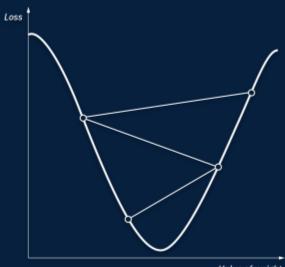
**2D**:



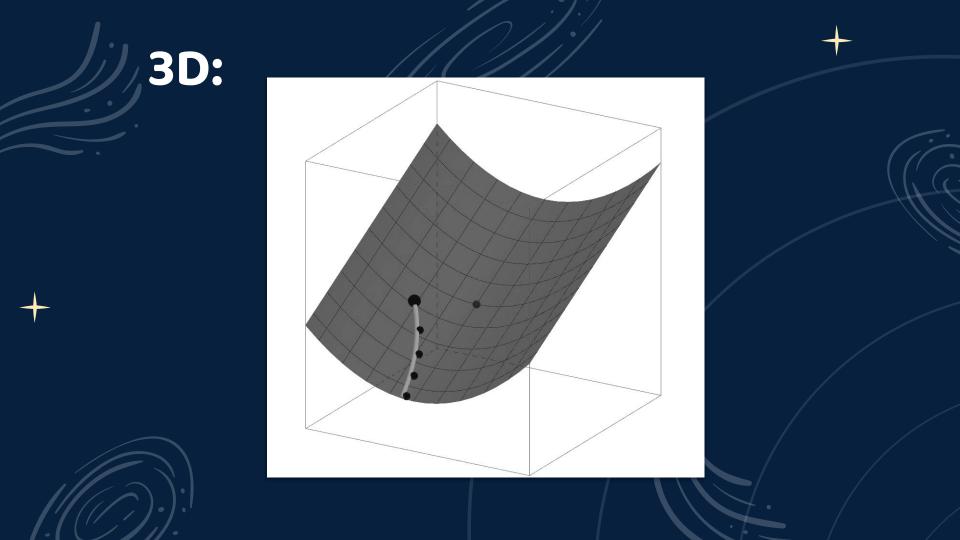




Large learning rate

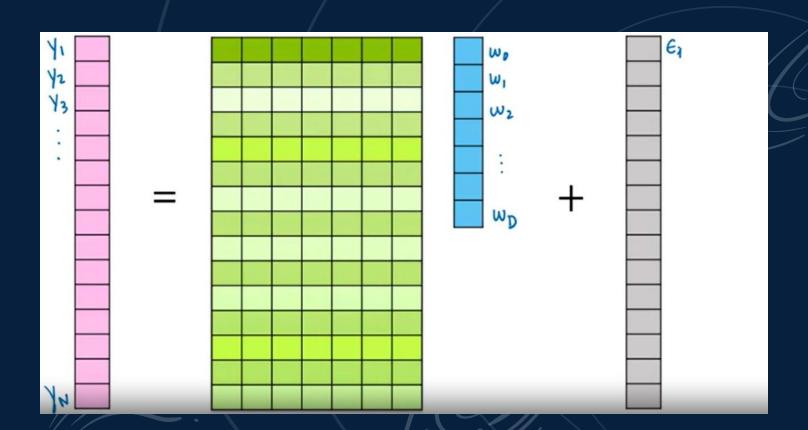


Value of weight





#### N-Dimensional:



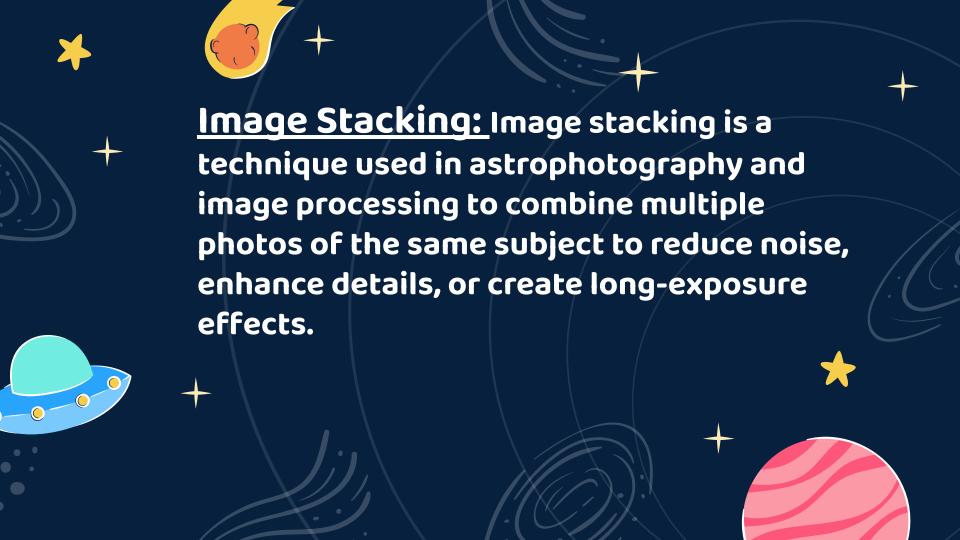
## Gradient of ND matrix form model:

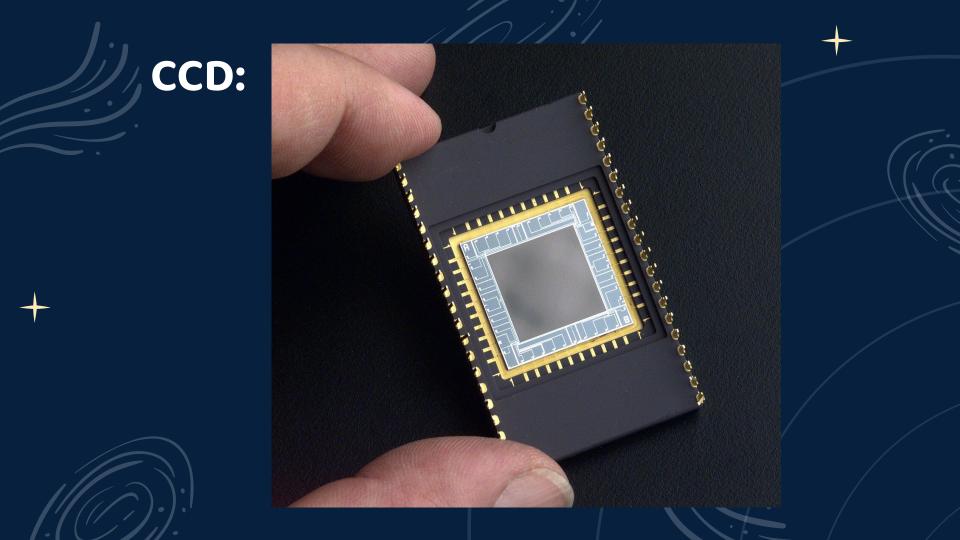
$$\nabla$$
RSS(w) =  $\nabla$ [(y-Hw)<sup>T</sup>(y-Hw)]  
= -2H<sup>T</sup>(y-Hw)

#### **Algorithm:**

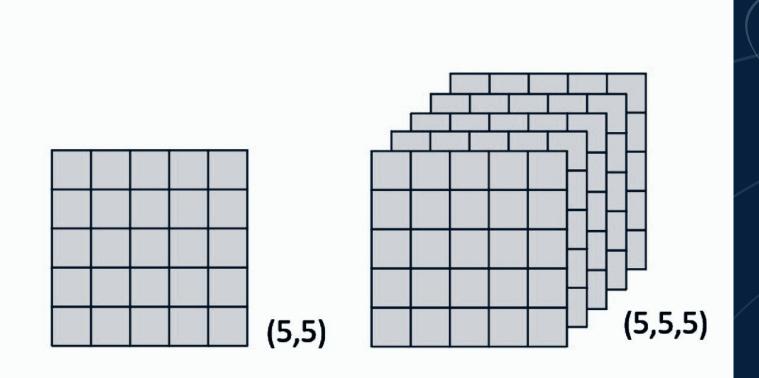
#### Gradient descent algorithm

repeat until convergence { 
$$\theta_j := \theta_j - \alpha \frac{\partial}{\partial \theta_j} J(\theta_0, \theta_1)$$
 (for  $j = 1$  and  $j = 0$ )



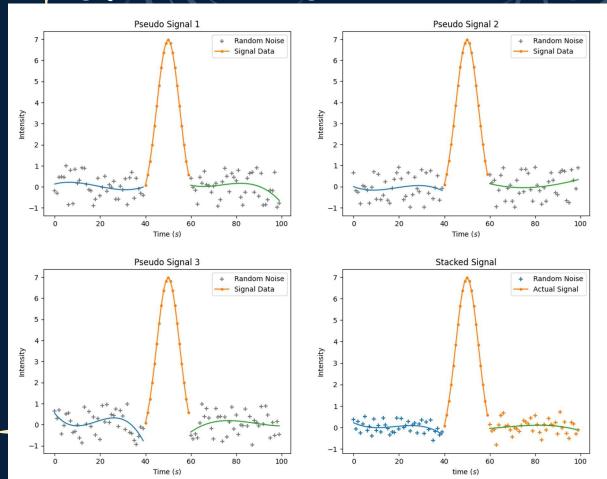


## Stacking grid of 5 images:



### Analogy with signals:







Before Stacking

After Stacking



## SPL Project's link:

https://github.com/hasin1415/SPL-01.git

