

Advanced Python for Neuroscientists

Lecture 3: Regression, Classification (Decoding)

2022/07/05

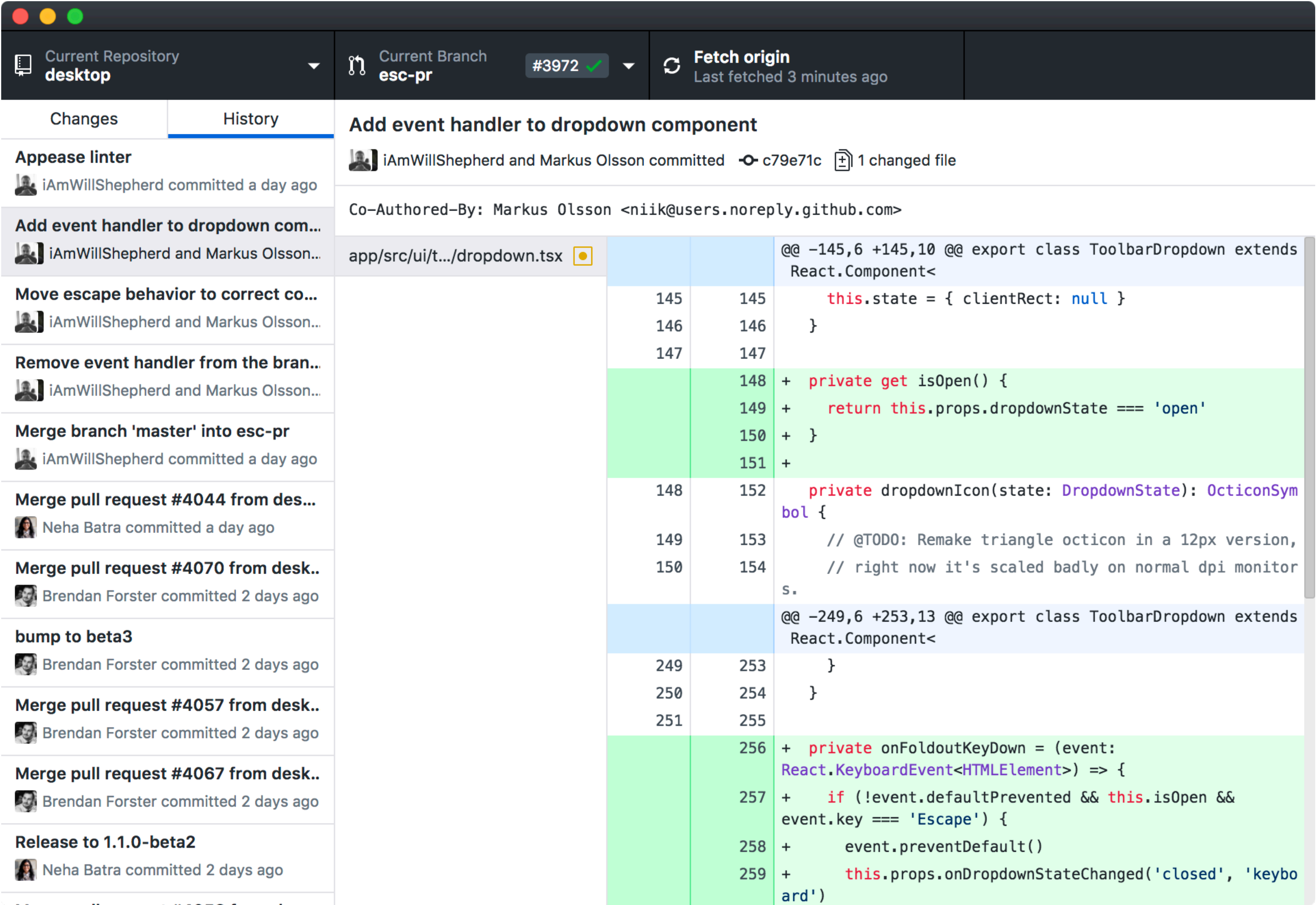
Git Command

Git(hub)

- Git - software for tracking changes in any set of files, usually used for coordinating work among programmers collaboratively developing source code during software development. `conda install numpy git`
- GitHub, Inc. - a provider of Internet hosting for software development and version control using Git
- Versions with new changes - commits
- Upload - push; download - pull

User interface

- Github desktop



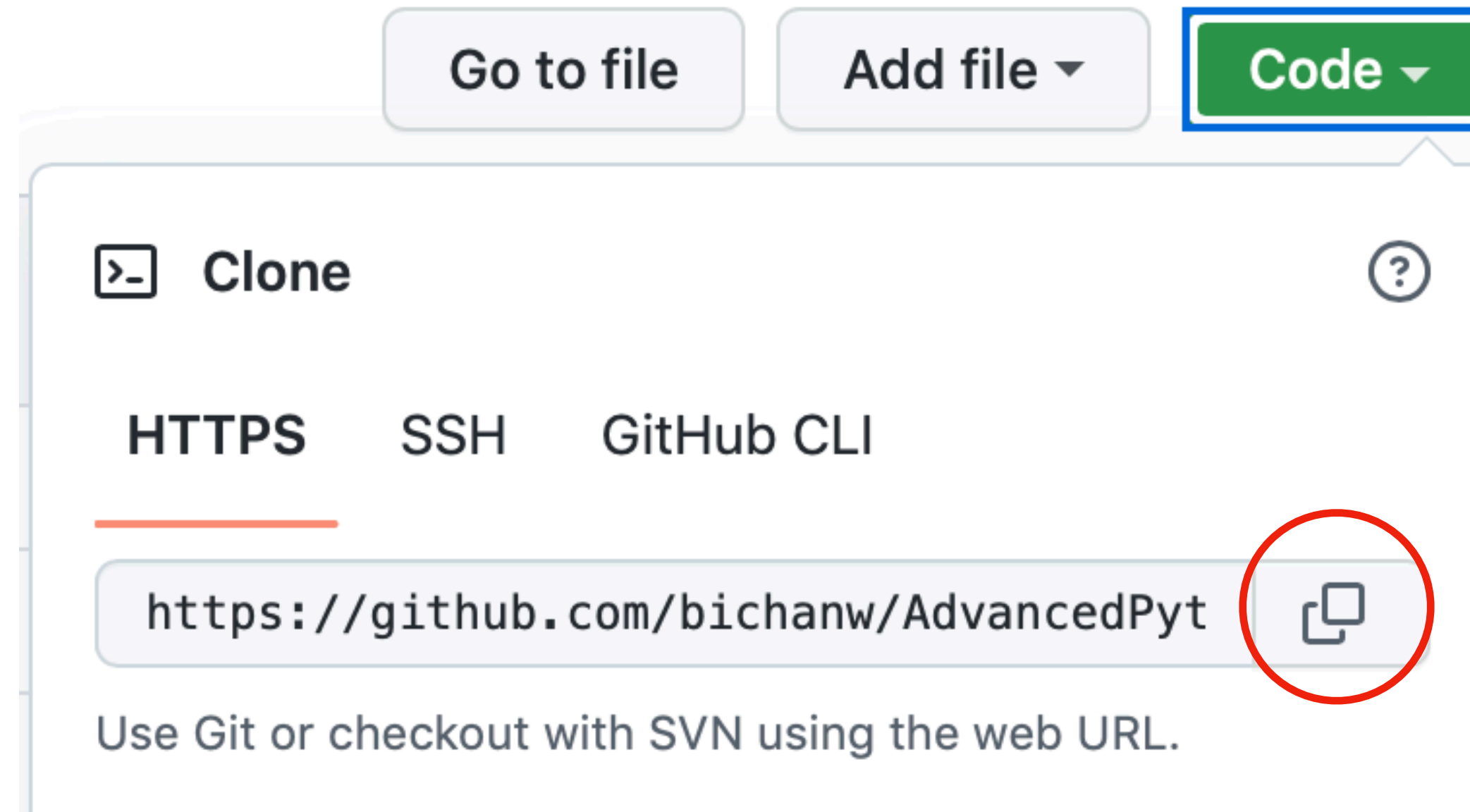
User interface

- Github desktop
- Sublime merge

User interface

Command lines

- First download: git clone



User interface

Command lines

- First download: `git clone + repo location`
- Update new changes to local: `git pull`
- Update local changes to GitHub:
 - Add local changes to a new commit: `git add -A / file name`
 - Specify what changes are made: `git commit -m "name the changes"`
 - Upload changes: `git push`

User interface

Command lines

- Cancel local changes: `git checkout filename`
- Hard reset:
 - `git fetch --all`
 - `git branch backup-master`
 - `git reset --hard origin/master`
- Put them all in a bash script (example)

Object-oriented Programming

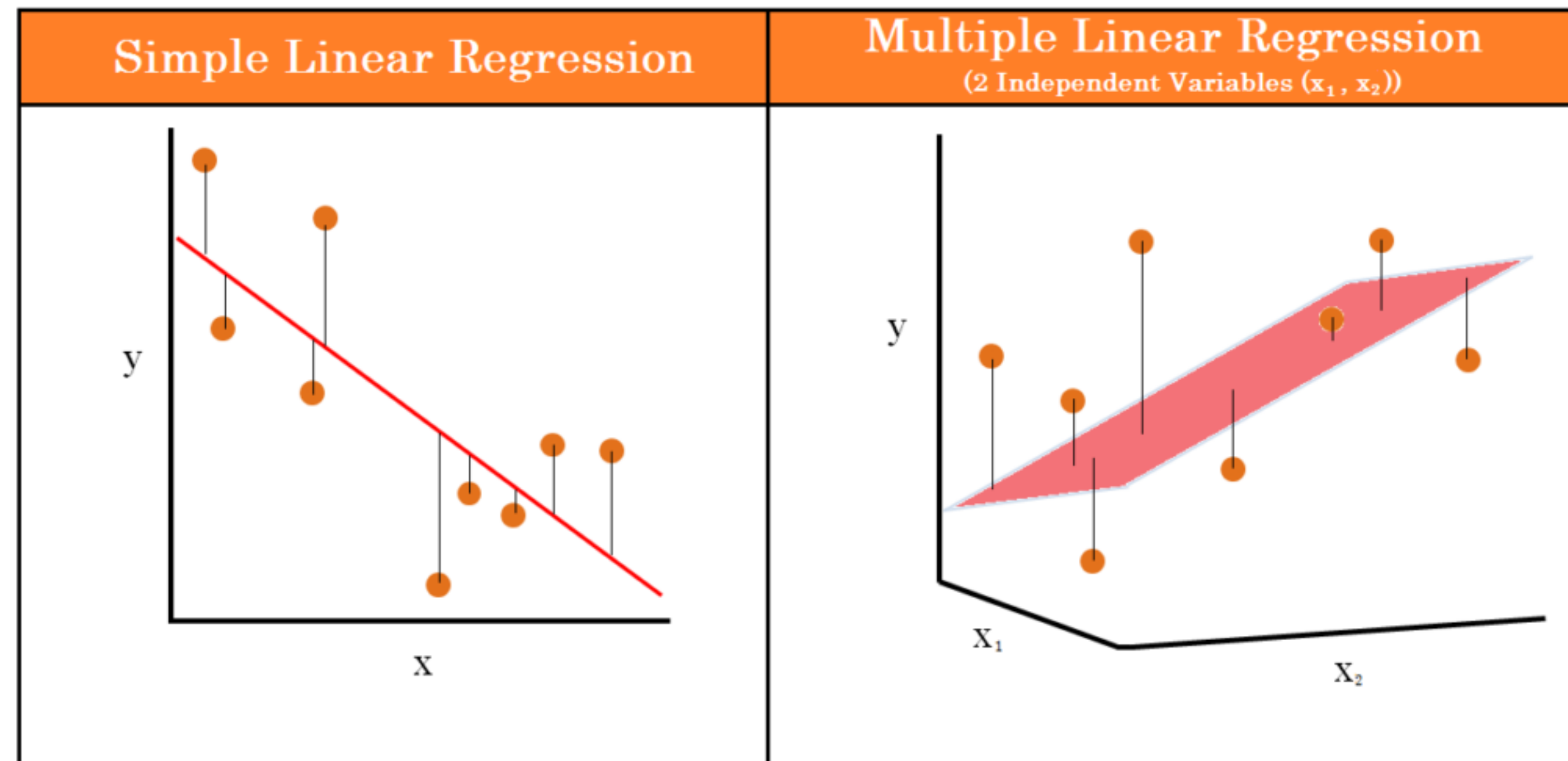
Object-oriented programming

- An object with packs of values + codes
 - Dictionary (struct) + functions
- Easier for organizing and naming functions
- `dog_name = dog()`
- `dog_name.bark(Y)`

Regression

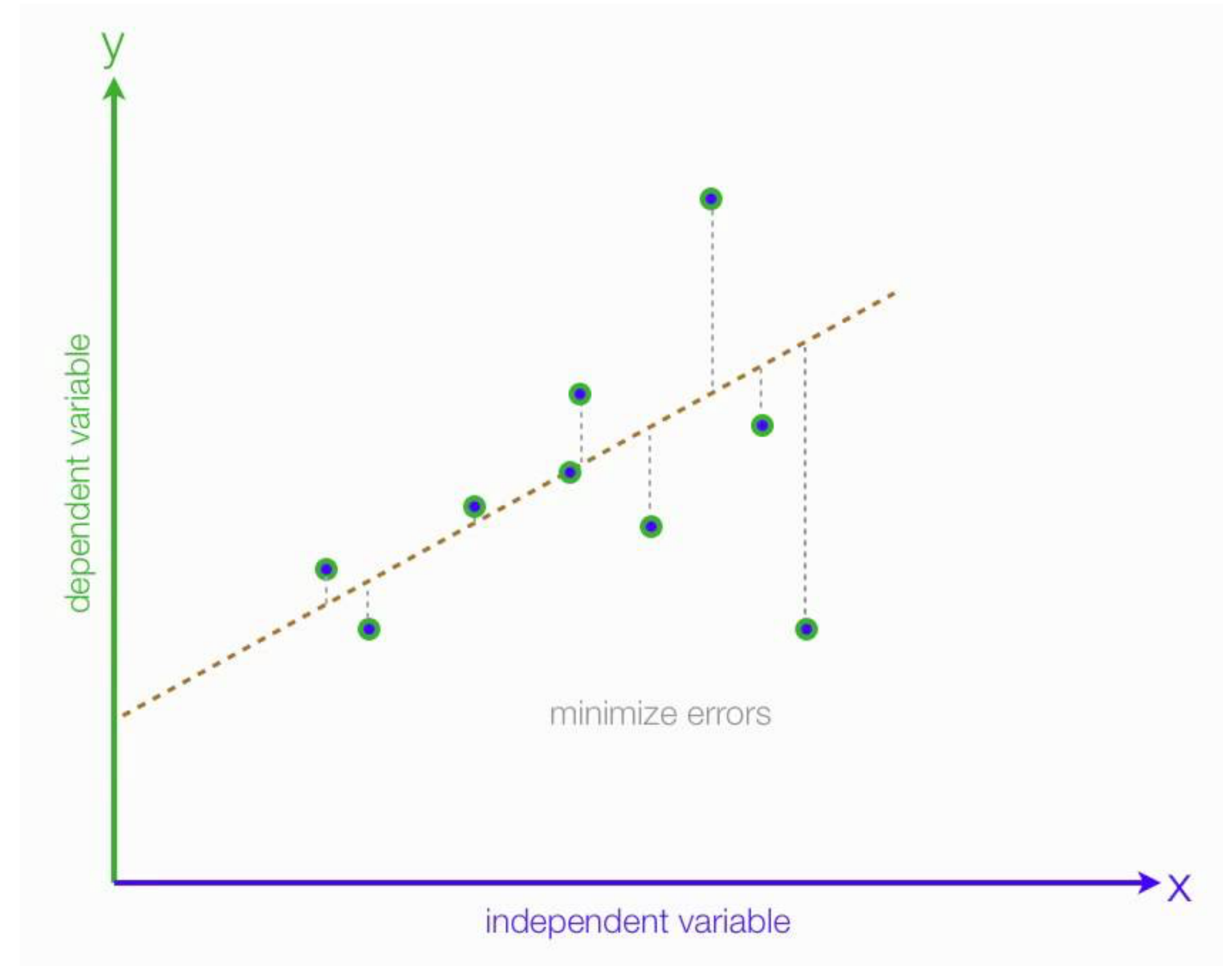
Linear Regression

- Predicting a real-valued output y
- given a vector of real valued inputs x_1, \dots, x_n
- Assuming their relationship is linear: $y = w_1x_1 + \dots w_nx_n + w_0$



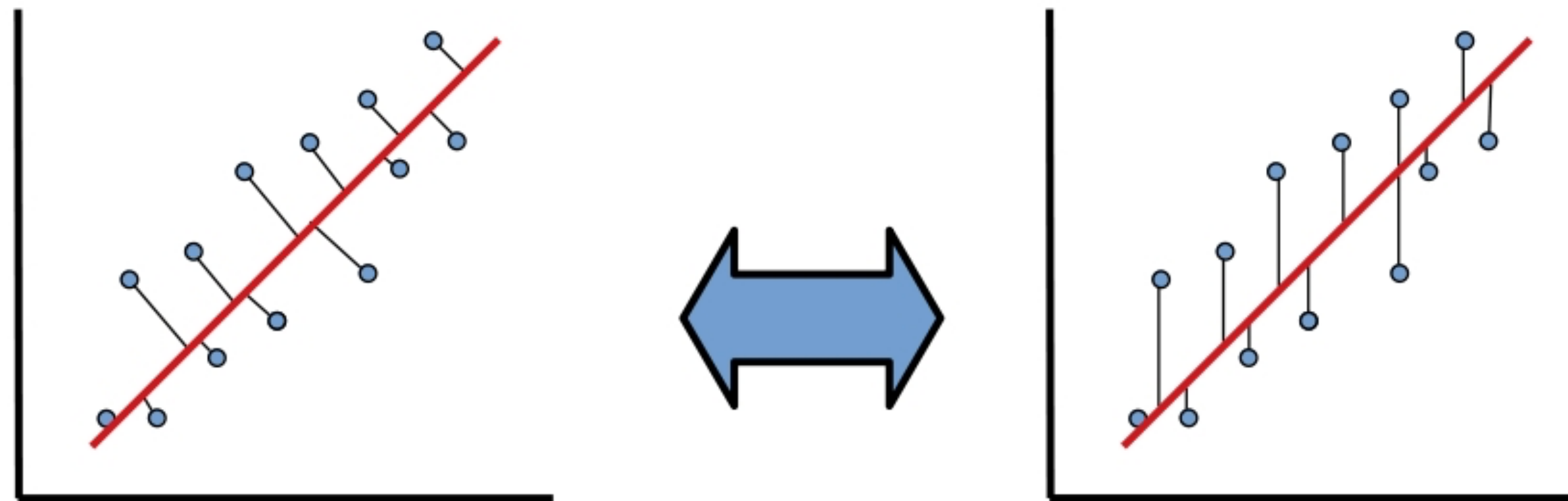
How to find w

- Find the line the best fits the data
- Predict y with x
- Minimize difference between y and $y_{\text{predicted}}$



PCA vs linear regression

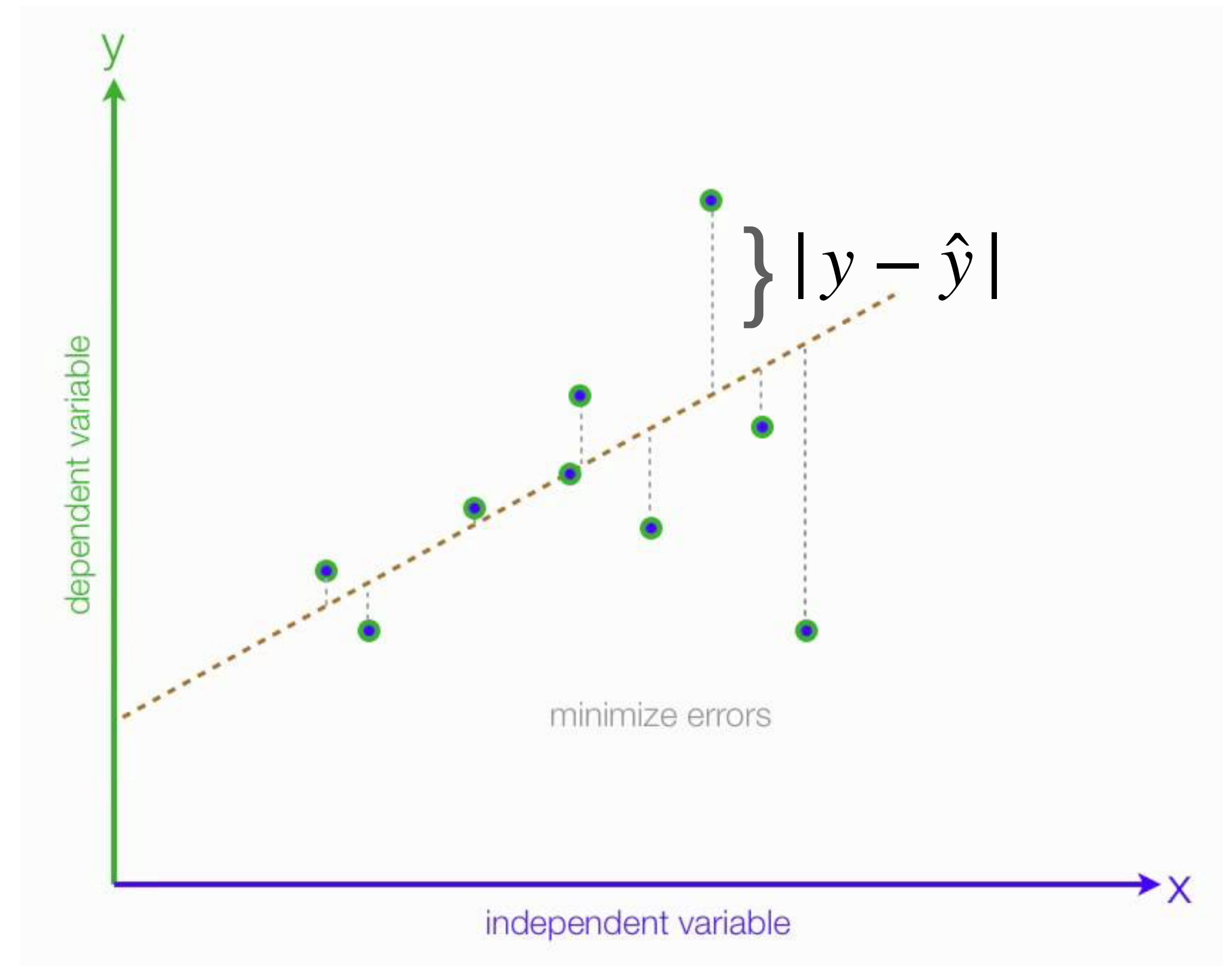
- Distance to the principal axis vs. distance between y, \hat{y}
- Treat 2 variables as the same vs. input & output



- estimation
- $\mathbf{X}\mathbf{B} = \mathbf{Y}$
- $\mathbf{X}^T\mathbf{X}\mathbf{B} = \mathbf{X}^T\mathbf{Y}$

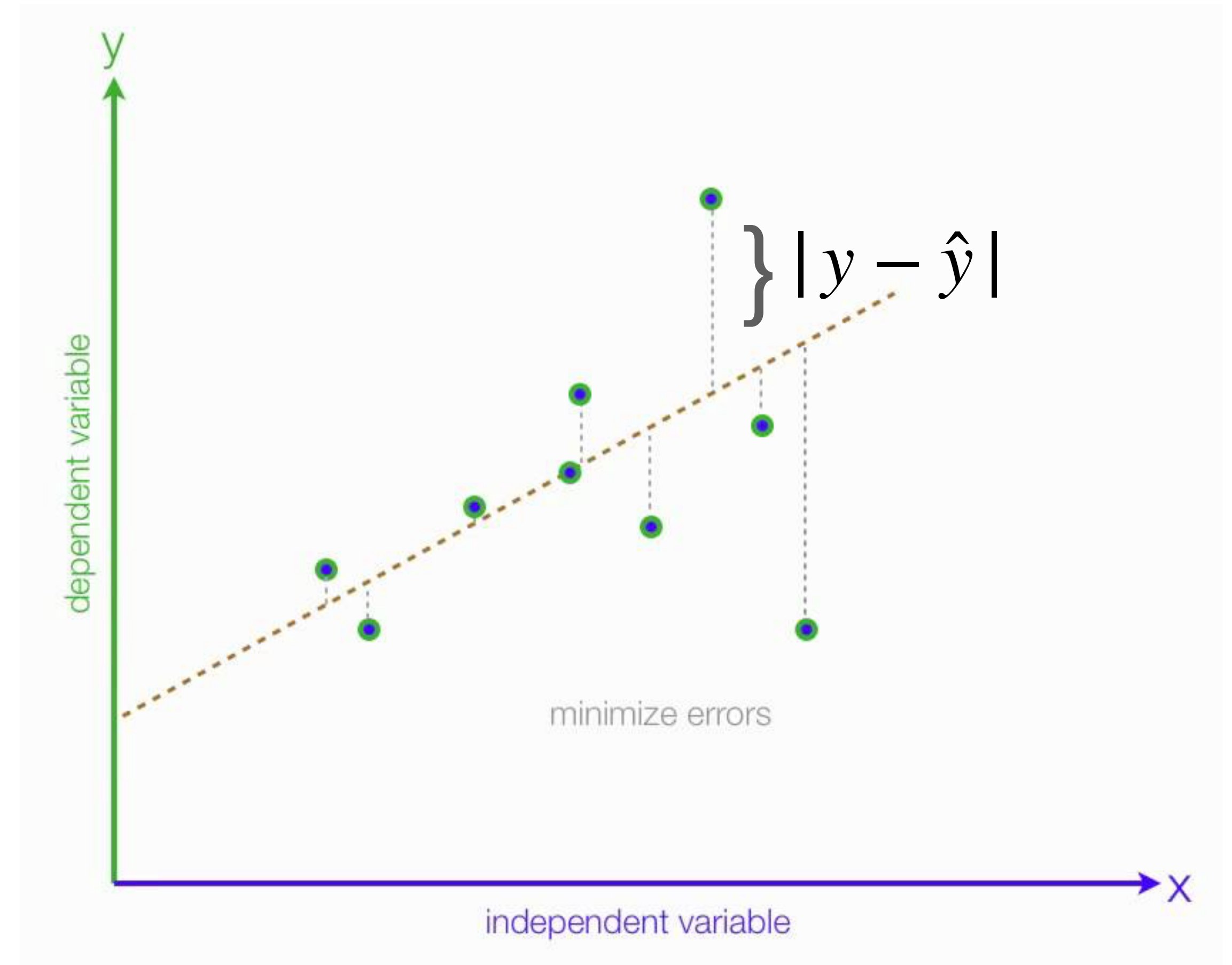
$$\text{NLL}(\mathbf{w}, \sigma^2) = - \sum_{n=1}^N \log \left[\left(\frac{1}{2\pi\sigma^2} \right)^{\frac{1}{2}} \exp \left(-\frac{1}{2\sigma^2} (y_n - \mathbf{w}^T \mathbf{x}_n)^2 \right) \right]$$

$$(\mathbf{X}^T \mathbf{X})^{-1} \mathbf{X}^T \mathbf{Y}$$



How good is w

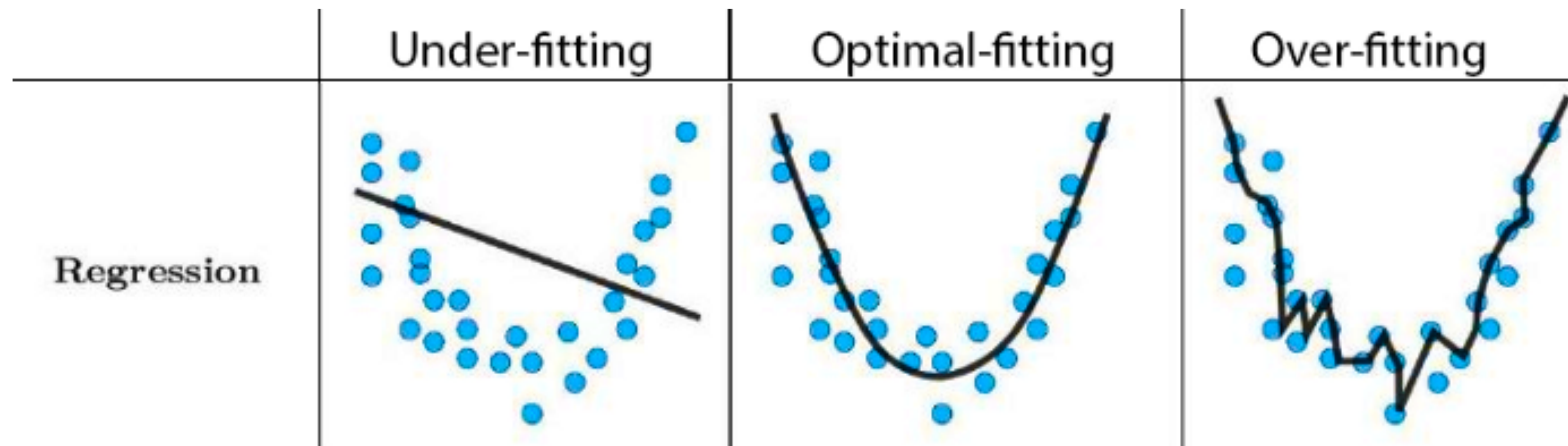
- $\sum (y - \hat{y})^2$ - least square estimation
- $1 - \frac{\sum (y - \hat{y})^2}{y^2}$
- Coefficient of determination
- Correlation coefficient (1d)
- R^2
- 1 - root mean squared error



Linear Regression Expansion

Add regularization

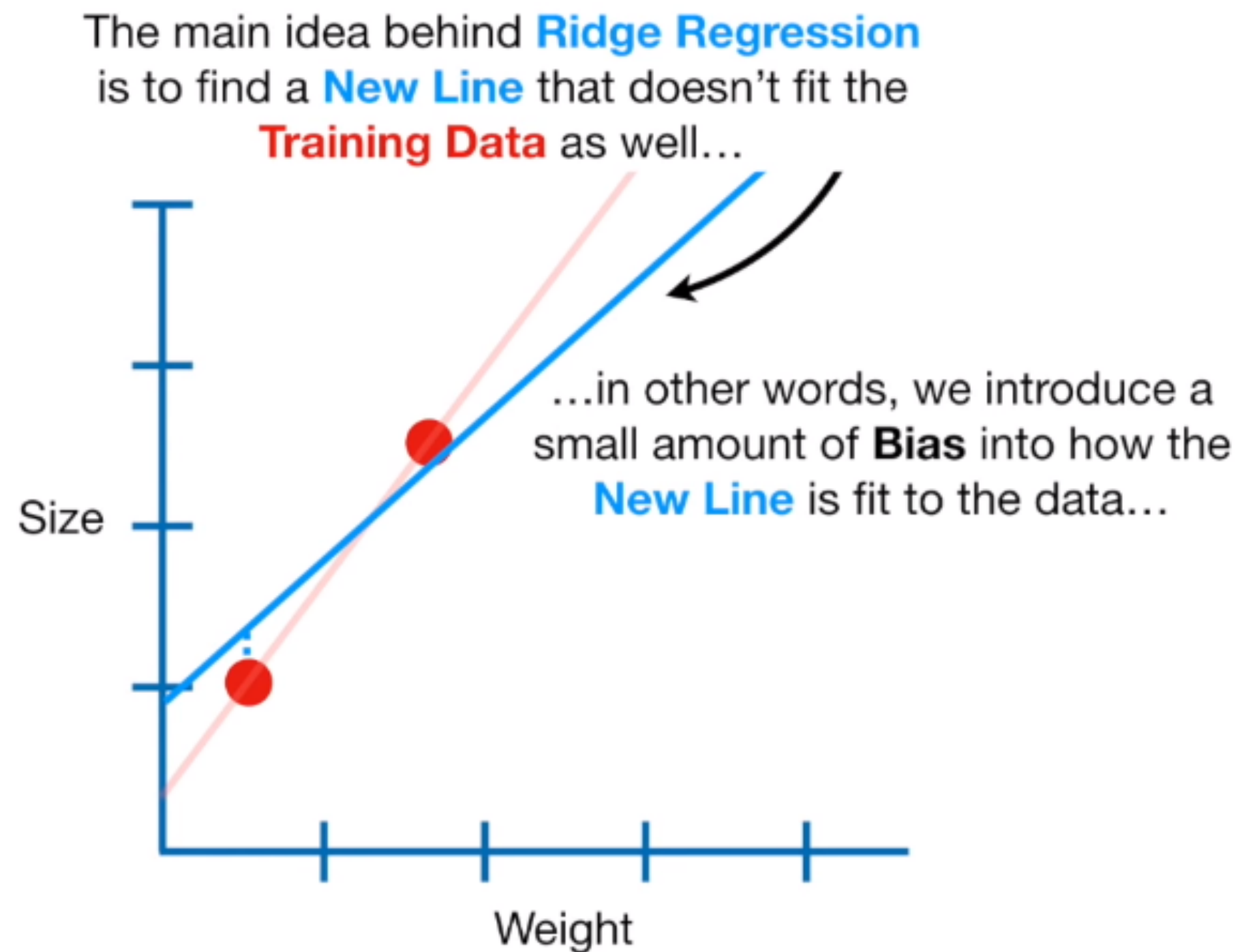
- Overfitting - the production of an analysis that corresponds too closely or exactly to a particular set of data, and may therefore fail to fit to additional data or predict future observations reliably



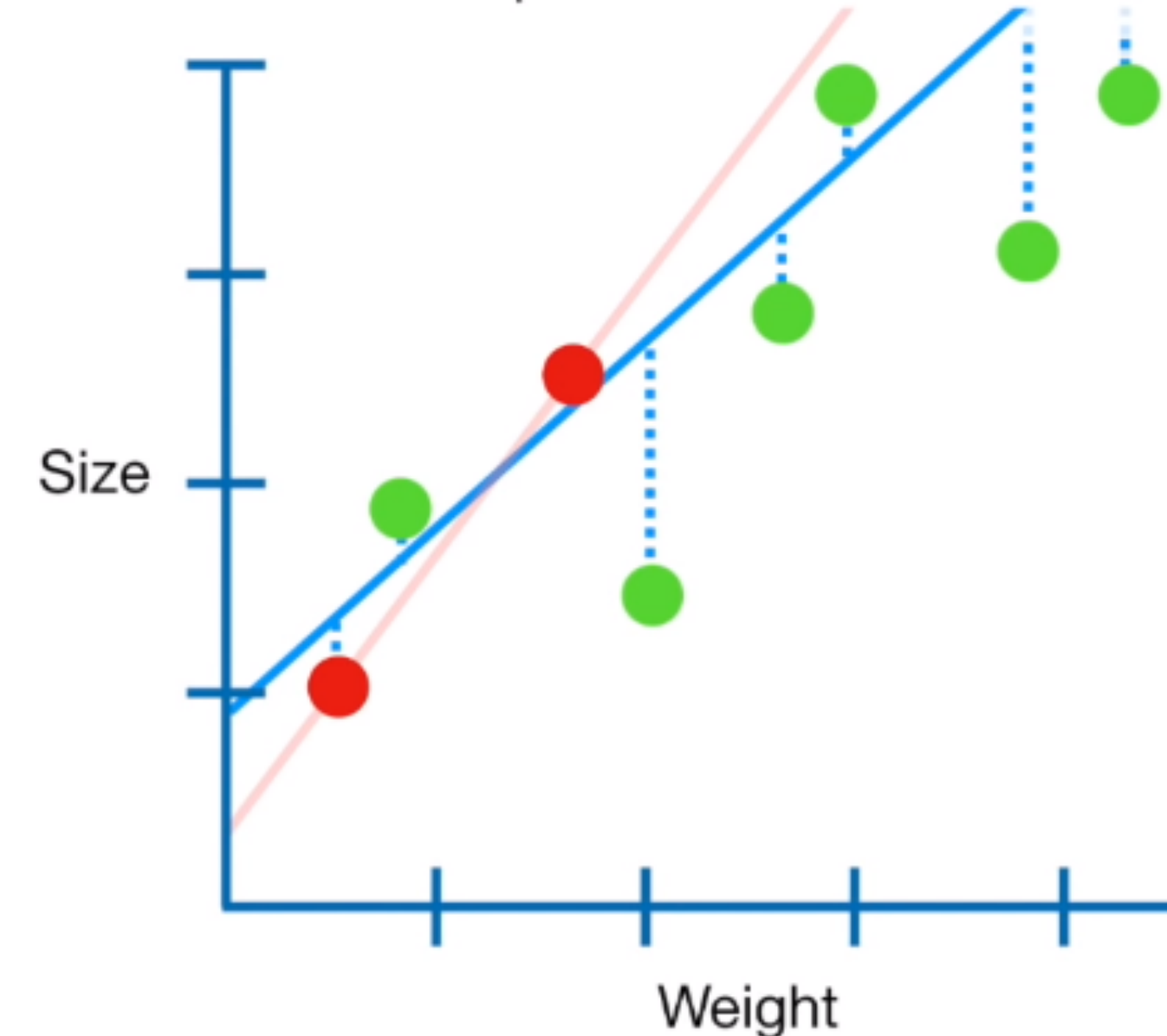
Linear Regression Expansion

Add regularization

- Overfitting



In other words, by starting with a slightly worse fit, **Ridge Regression** can provide better long term predictions.



Linear Regression Expansion

Add regularization

- Overfitting
- Regularization - make results “simpler”
 - Penalty term assumption: coefficients are not all large

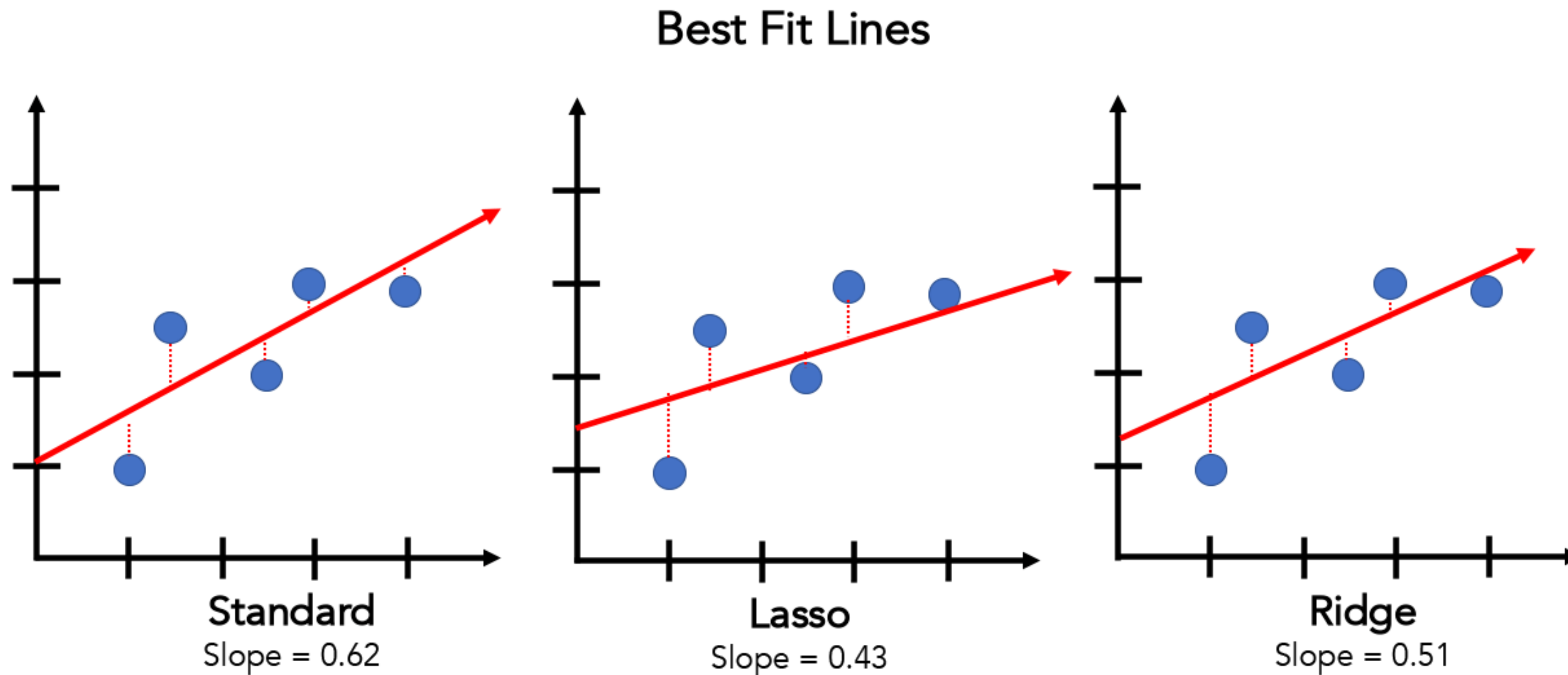
- Ridge (L2) regression: $\lambda \sum_i w_i^2 + \sum (y - \hat{y})^2$

- Lasso (L1) regression: $\text{RSE} + \lambda \sum_i |w_i|$

Linear Regression Expansion

Add regularization

- Overfitting



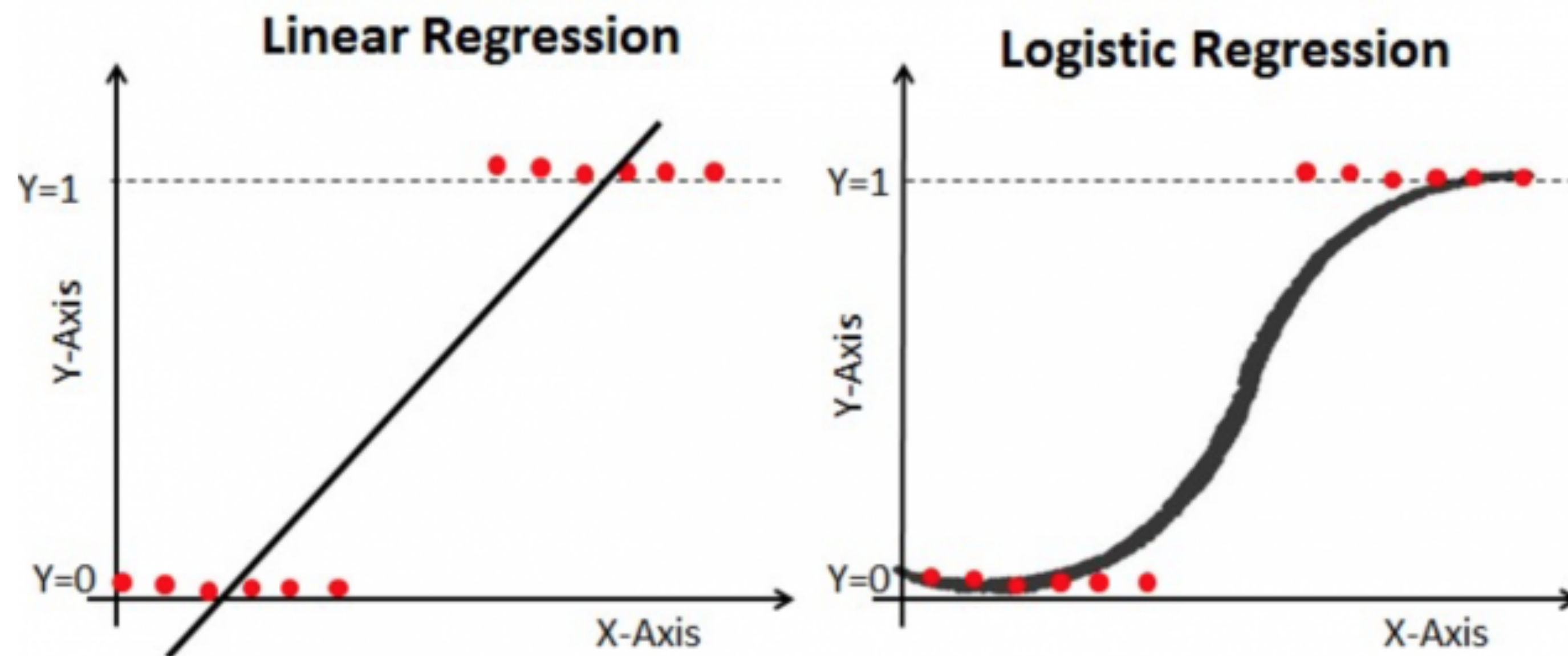
Linear Regression Expansion

Add nonlinear terms

- $y = w_1x_1 + \dots w_nx_n + w_0$
- What if true relationship is slightly more complicated: $y = x_1 + 3x_2^2$
- Simply calculate $x_3 = x_2^2$, run linear regression on x_1, x_2, x_3
- You need to have an idea what the nonlinear term is (e.g. $\sin x, x_1x_2$)

Logistic Regression

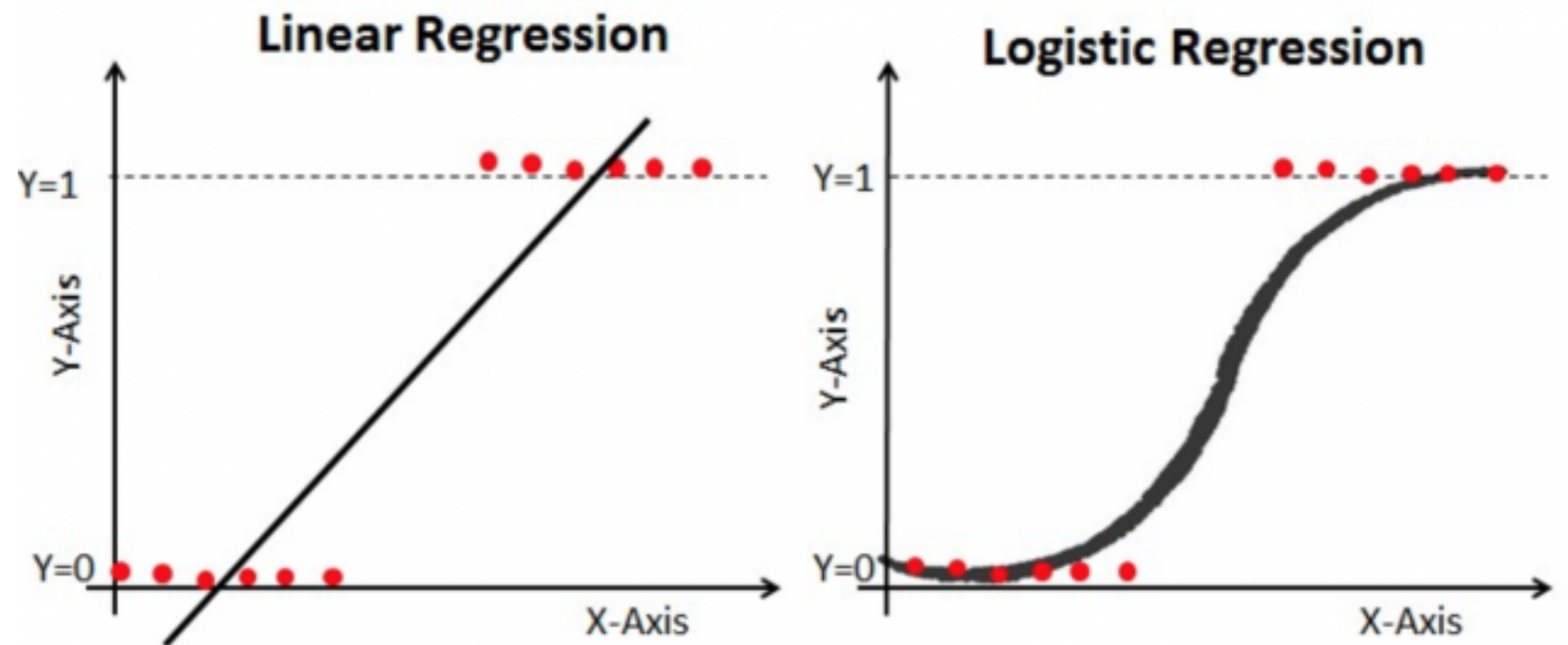
- Predicting a binary y



Logistic Regression

- Predicting a binary y
- Add a transformation to compress y - logistic function

- $$y = \frac{1}{1 + e^{w_1x_1 + w_2x_2 \dots w_0}}$$



Logistic Regression

- Predicting a binary y
- Add a transformation to compress y - logistic function
 - $$y = \frac{1}{1 + e^{w_1x_1 + w_2x_2 \dots w_0}}$$
- Commonly used in binary outcome tasks (selecting left / right)