

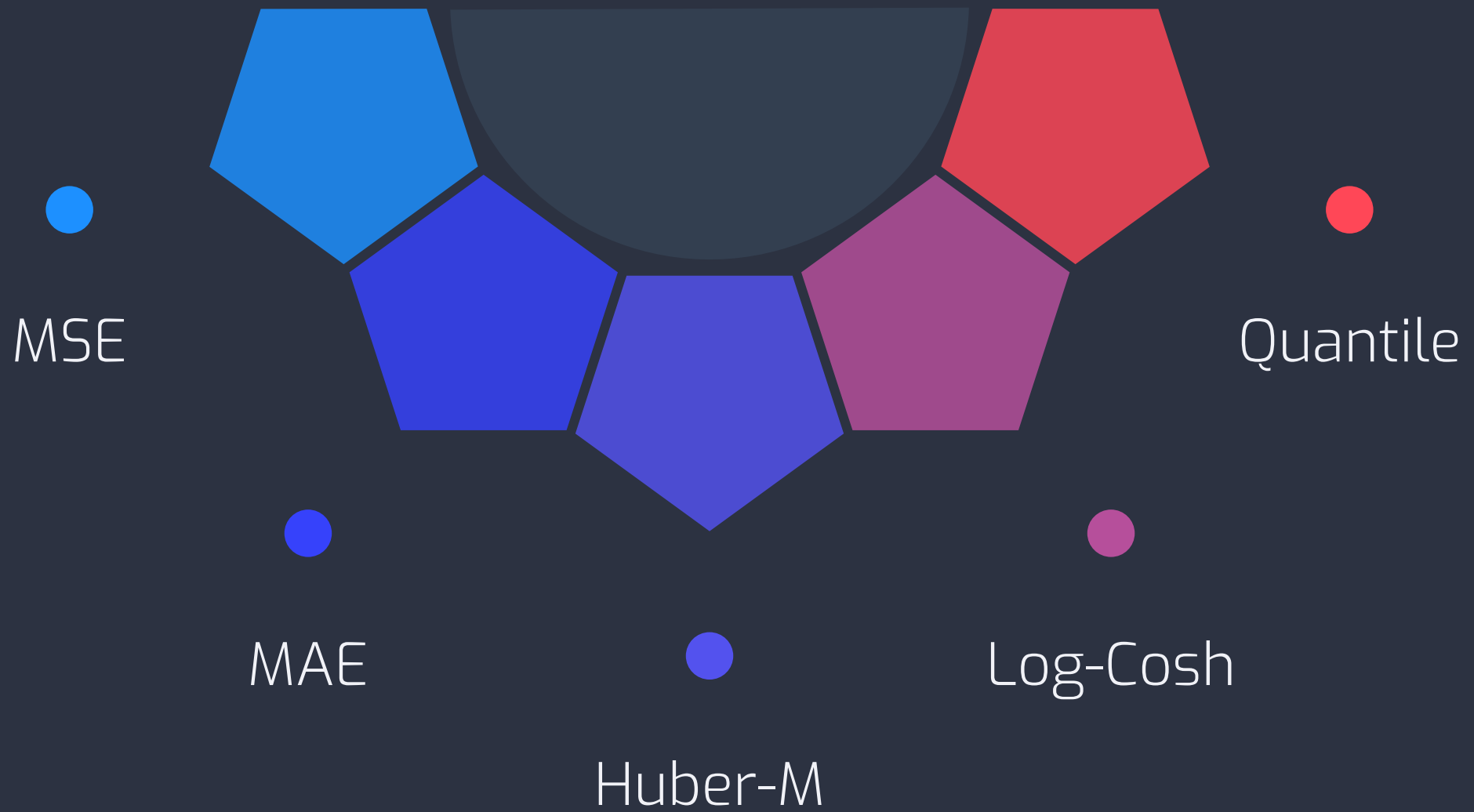
COMP2261 ARTIFICIAL INTELLIGENCE / MACHINE LEARNING

Cost Functions For Regression Models -- Huber-M

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🎯 Learning Objectives

- Understand what is Huber-M
- Understand how Huber-M cost function works
- Understand the hyperparameter δ

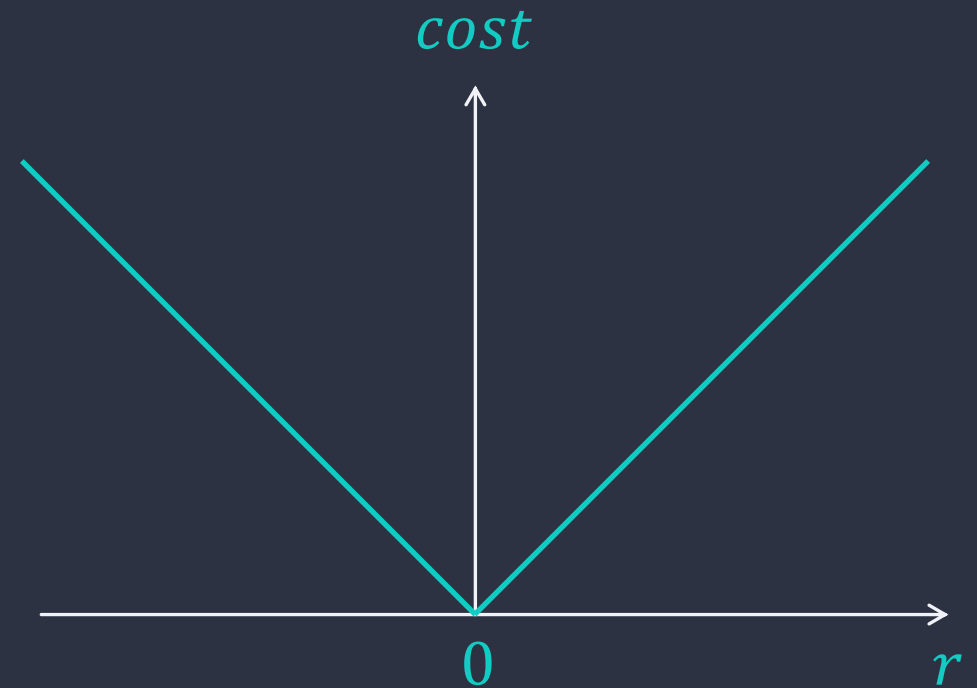


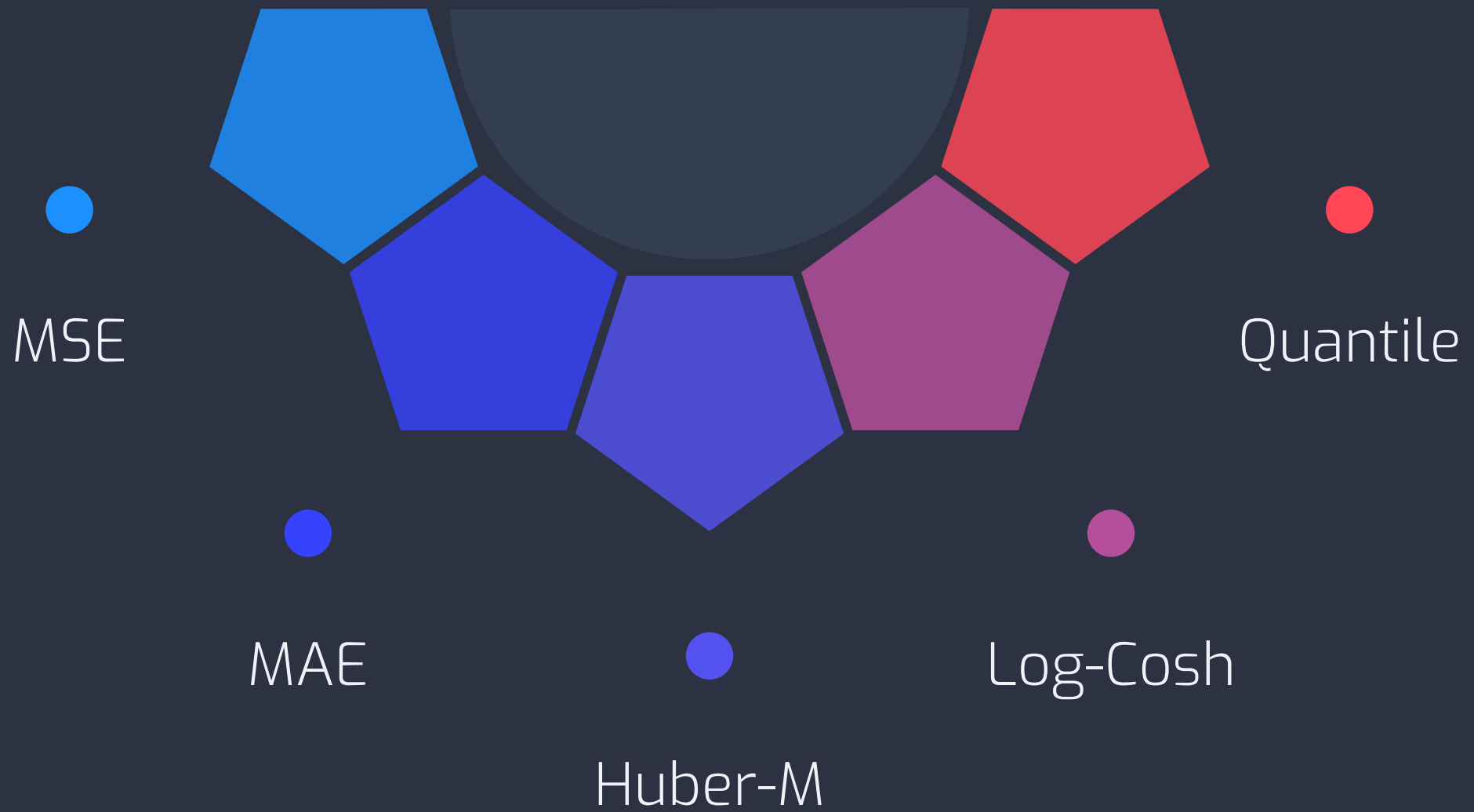
Mean Squared Error (MSE) vs Mean Absolute Error (MAE)

$$J = \frac{1}{m} \sum_{i=1}^m (y^{(i)} - \hat{y}^{(i)})^2$$



$$J = \frac{1}{m} \sum_{i=1}^m |y^{(i)} - \hat{y}^{(i)}|$$





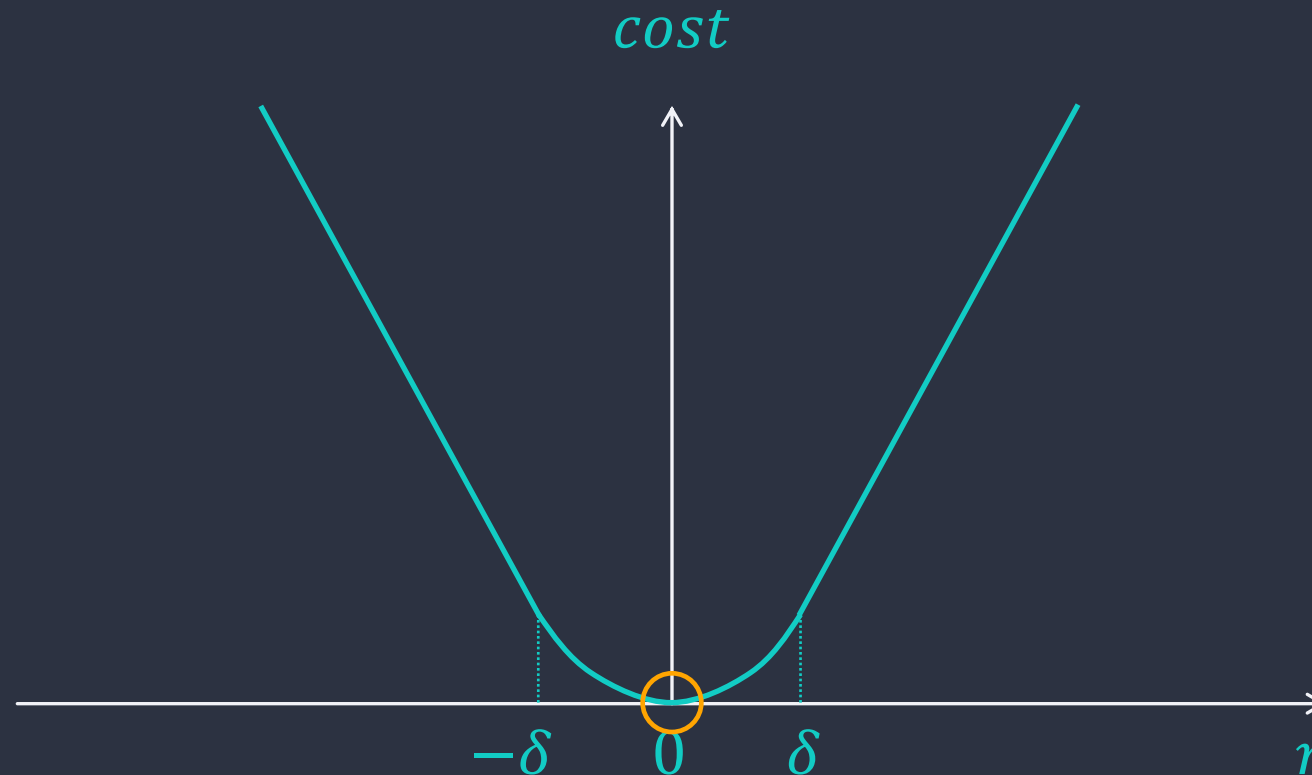
Huber-M

Huber-M

$$J = \frac{1}{m} \sum_{i=1}^m \begin{cases} \frac{1}{2} (y^{(i)} - \hat{y}^{(i)})^2, & |y^{(i)} - \hat{y}^{(i)}| \leq \delta \\ \delta (|y^{(i)} - \hat{y}^{(i)}| - \frac{1}{2} \delta), & \textit{otherwise} \end{cases}$$

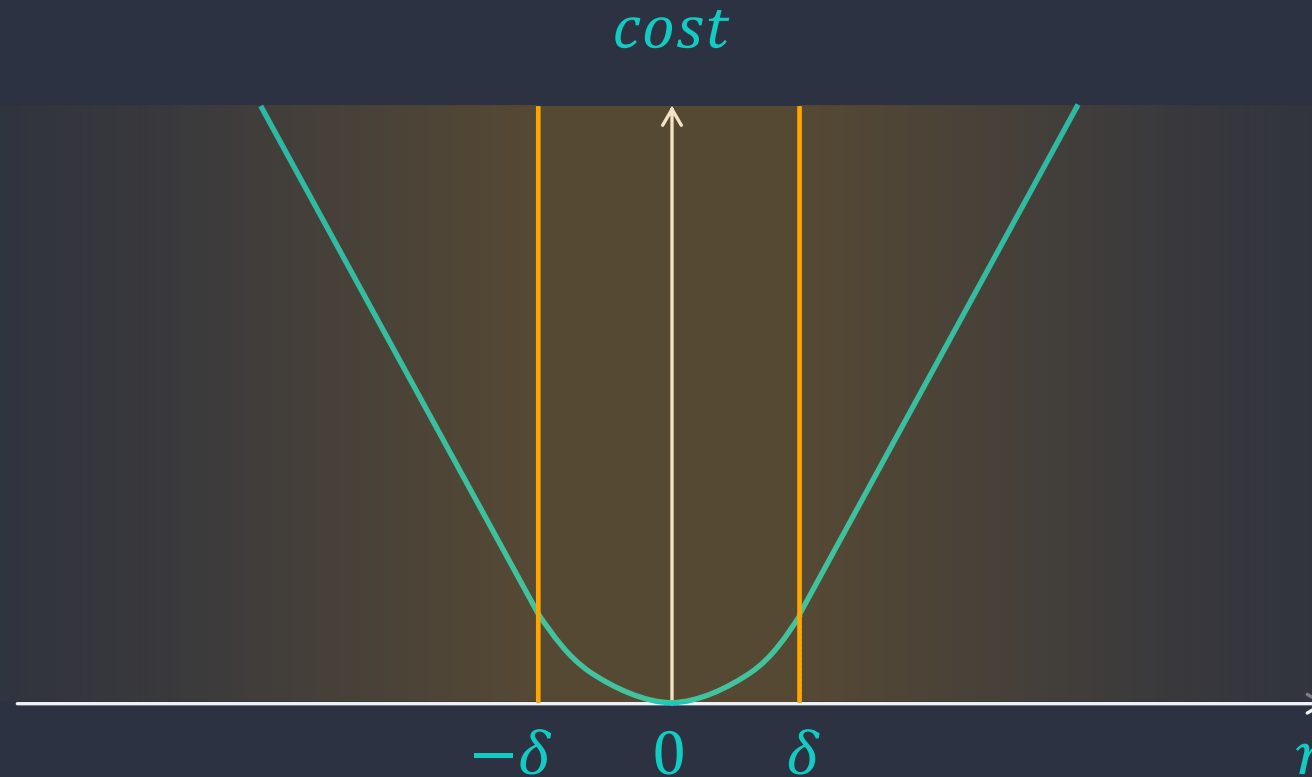
Huber-M

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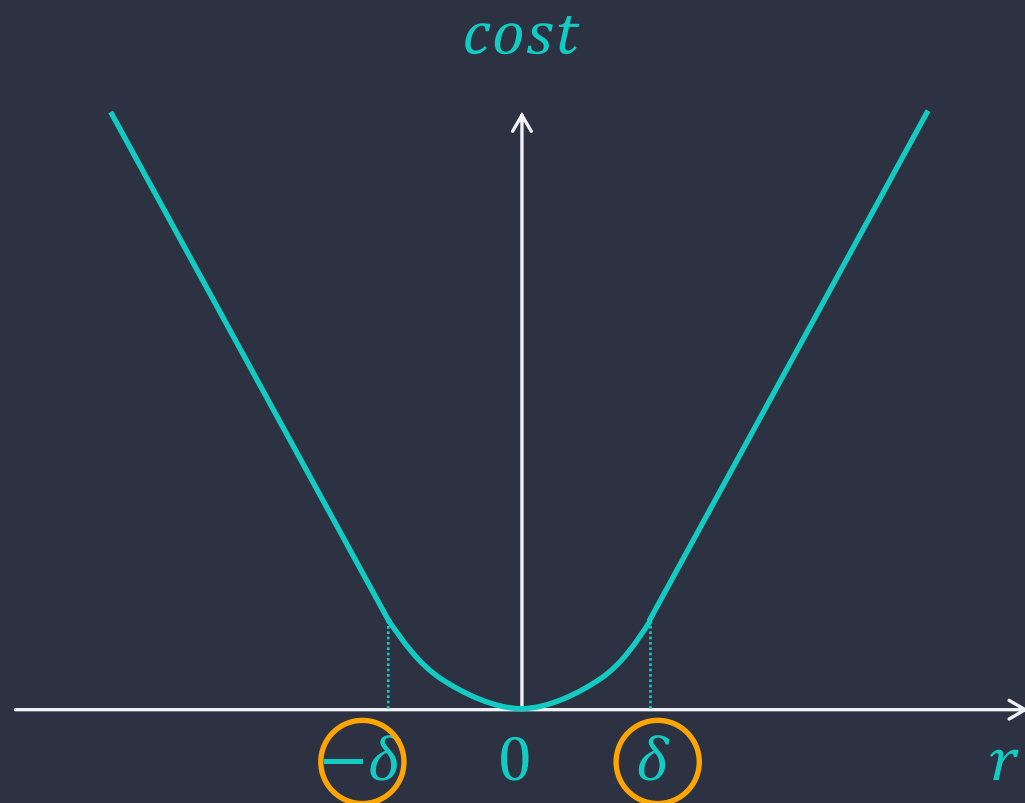
Huber-M

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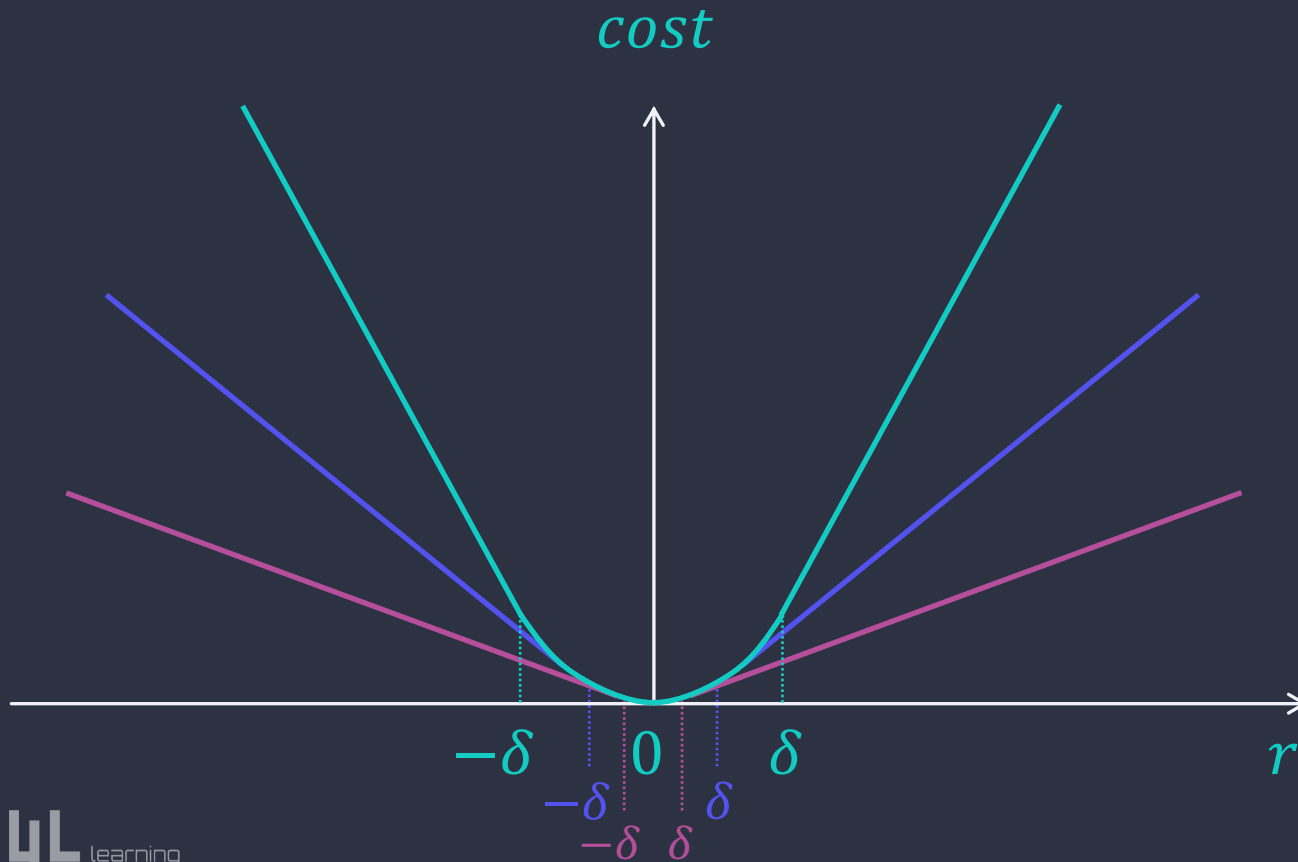


label (actual) label (predicted)

	1	2	3	4	5	...	n	y	\hat{y}	$ r $	
1	■	■	■	■	■	...	■	●	○	::	$\max r $
2	■	■	■	■	■	...	■	●	○	::	
3	■	■	■	■	■	...	■	●	○	::	10% δ
4	■	■	■	■	■	...	■	●	○	::	
5	■	■	■	■	■	...	■	●	○	::	
...	
m	■	■	■	■	■	...	■	●	○	::	$\min r $

Huber-M

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Choice of δ is very important, as it determines what are considered as outliers and how they are dealt with.

✓ Takeaway Points

- Huber-M combines good properties from MSE and MAE.
- Curving around minima – reducing gradient during training.
- More robust with outliers.
- Important to choose hyperparameter δ as it determines what are considered as outliers and how to deal with them.