# Preliminary Work

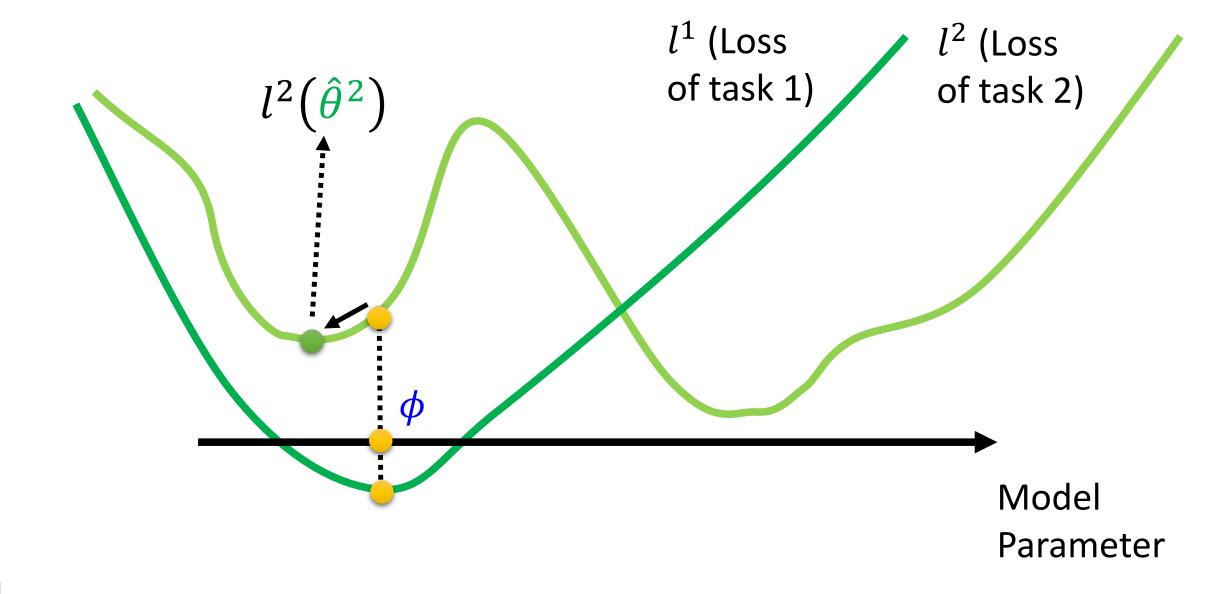
## **MAML**

### **Model Pre-training**

$$L(\boldsymbol{\phi}) = \sum_{n=1}^{N} l^n(\boldsymbol{\phi})$$

找尋在所有 task 都最好的  $\phi$ 

並不保證拿 $\phi$  去訓練以後會 得到好的 $\hat{\theta}^n$ 

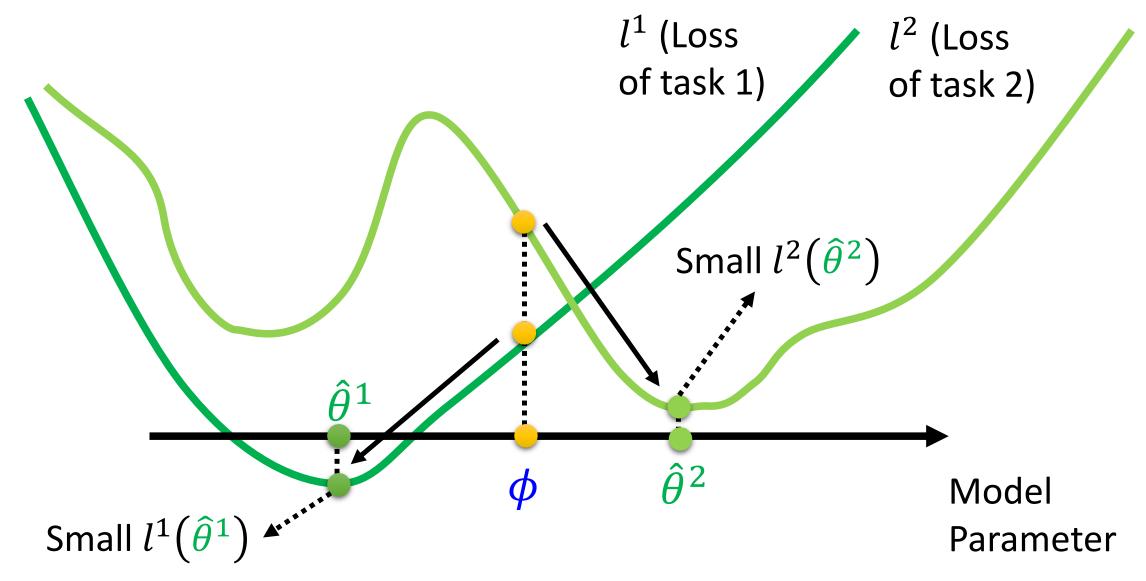


#### **MAML**

$$L(\boldsymbol{\phi}) = \sum_{n=1}^{N} l^n(\hat{\boldsymbol{\theta}}^n)$$

我們不在意  $\phi$  在 training task 上表現如何

我們在意用  $\phi$  訓練出來的  $\hat{\theta}^n$  表現如何





## Preliminary Work **MAML**

## MAML

 $\hat{\theta}^n$ : model learned from task n $\hat{\theta}^n$  depends on  $\phi$ 

Loss Function:

$$L(\phi) = \sum_{n=1}^{N} l^n(\hat{\theta}^n)$$

 $l^n(\widehat{\theta}^n)$ : loss of task n on the testing set of task n

How to minimize  $L(\phi)$ ? Gradient Descent

$$\phi \leftarrow \phi - \eta \nabla_{\phi} L(\phi)$$

Find  $\phi$  achieving good performance after training

潛力

## Model Pre-training

Widely used in transfer learning Loss Function:

$$L(\phi) = \sum_{n=1}^{N} l^n(\phi)$$

Find  $\phi$  achieving good performance 現在表現如何

