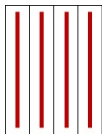
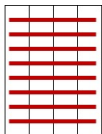


Advanced Machine Learning

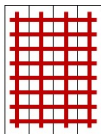
Loss Functions for Multi-Target Prediction



macro



instance-wise



micro

Learning goals

- Get to know loss functions for multi-target prediction problems
- Know the Bayes predictor for Hamming loss and subset 0/1 loss
- Understand the difference between macro-, micro-, and instance-wise-losses

MULTIVARIATE LOSS FUNCTIONS

- In multi-target prediction we want to the following: For a feature vector \mathbf{x} , predict a vector of scores $\mathbf{y} = (y_1, y_2, \dots, y_m)^\top$ by means of a function (hypothesis) f :

$$\mathbf{x} = (x_1, x_2, \dots, x_p)^\top \xrightarrow{f(\mathbf{x})} \hat{\mathbf{y}} = (\hat{y}_1, \hat{y}_2, \dots, \hat{y}_m)^\top$$

- If we want to follow the machine learning paradigm based on loss minimization, we need a *multivariate loss functions*

$$\ell : \mathcal{Y}^m \times \mathcal{Y}^m \rightarrow \mathbb{R}.$$

Compared to single-target prediction, a broad spectrum of such multivariate loss functions is conceivable.

- In case we have an appropriate multivariate loss function ℓ , we want to find a (Bayes) predictor f^* that minimizes expected loss with regard to ℓ :

$$\begin{aligned} f^* &= \arg \min_{f: \mathcal{X} \rightarrow \mathcal{Y}^m} \mathcal{R}_\ell(f) = \arg \min_{f: \mathcal{X} \rightarrow \mathcal{Y}^m} \mathbb{E}_{xy} [\ell(y, f(\mathbf{x}))] \\ &= \arg \min_{f: \mathcal{X} \rightarrow \mathcal{Y}^m} \int \ell(y, f(\mathbf{x})) d\mathbb{P}_{xy}. \end{aligned}$$

EXAMPLES OF MTP LOSS FUNCTIONS

- *Squared error loss* (typically used in multivariate regression):

$$\ell(\mathbf{y}, \hat{\mathbf{y}}) = \sum_{j=1}^m (y_j - \hat{y}_j)^2,$$

where $\mathbf{y}, \hat{\mathbf{y}} \in \mathbb{R}^m$.

- The *Hamming loss* averages over mistakes on individual scores:

$$\ell_H(\mathbf{y}, \hat{\mathbf{y}}) = \frac{1}{m} \sum_{j=1}^m \mathbb{1}_{[y_j \neq \hat{y}_j]}$$

- The *subset 0/1 loss* simply checks for entire correctness:

$$\ell_{0/1}(\mathbf{y}, \hat{\mathbf{y}}) = \mathbb{1}_{[\mathbf{y} \neq \hat{\mathbf{y}}]} = \max_j \mathbb{1}_{[y_j \neq \hat{y}_j]}$$

HAMMING VS. SUBSET 0/1 LOSS

- The risk minimizer for the Hamming loss is the *marginal mode*:

$$f_j^*(\mathbf{x}) = \arg \max_{y_j \in \{0,1\}} \Pr(y_j | \mathbf{x}), \quad j = 1, \dots, m,$$

while for the subset 0/1 loss it is the *joint mode*:

$$\mathbf{f}^*(\mathbf{x}) = \arg \max_{\mathbf{y} \in \mathcal{Y}^m} \Pr(\mathbf{y} | \mathbf{x}).$$

- Marginal mode vs. joint mode:

\mathbf{y}	$\Pr(\mathbf{y})$
0 0 0 0	0.30
0 1 1 1	0.17
1 0 1 1	0.18
1 1 0 1	0.17
1 1 1 0	0.18

Marginal mode: 1 1 1 1

Joint mode: 0 0 0 0

MULTIVARIATE LOSS FUNCTIONS

- A loss L (on test data) is decomposable over examples if it can be written in the form

$$L = \sum_{i=1}^n \ell(\mathbf{y}^{(i)}, f(\mathbf{x}^{(i)})),$$

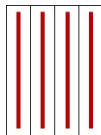
i.e., as a sum of losses over all (test) examples.

- A multivariate loss ℓ is decomposable over targets if it can be written as

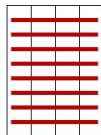
$$\ell(\mathbf{y}, f(\mathbf{x})) = \sum_{j=1}^m \ell_j(y_j, f_j(\mathbf{x}))$$

with suitable single-target losses ℓ_j .

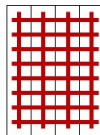
- In general, we distinguish between three categories of losses: macro-, micro-, and instance-wise-losses.



macro



instance-wise



micro

MACRO- AND MICRO-LOSSES

- Macro-losses: The overall loss corresponds to aggregating the losses over the targets.

True scores				Predicted scores			
y_{11}	y_{12}	y_{13}	y_{14}	\hat{y}_{11}	\hat{y}_{12}	\hat{y}_{13}	\hat{y}_{14}
y_{21}	y_{22}	y_{23}	y_{24}	\hat{y}_{21}	\hat{y}_{22}	\hat{y}_{23}	\hat{y}_{24}
y_{31}	y_{32}	y_{33}	y_{34}	\hat{y}_{31}	\hat{y}_{32}	\hat{y}_{33}	\hat{y}_{34}
y_{41}	y_{42}	y_{43}	y_{44}	\hat{y}_{41}	\hat{y}_{42}	\hat{y}_{43}	\hat{y}_{44}
y_{51}	y_{52}	y_{53}	y_{54}	\hat{y}_{51}	\hat{y}_{52}	\hat{y}_{53}	\hat{y}_{54}
y_{61}	y_{62}	y_{63}	y_{64}	\hat{y}_{61}	\hat{y}_{62}	\hat{y}_{63}	\hat{y}_{64}

- Example: Averaging the target losses.

$$L = \frac{1}{4} (L_1 + L_2 + L_3 + L_4)$$

MACRO- AND MICRO-LOSSES

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True scores				Predicted scores			
y_{11}	y_{12}	y_{13}	y_{14}	\hat{y}_{11}	\hat{y}_{12}	\hat{y}_{13}	\hat{y}_{14}
y_{21}	y_{22}	y_{23}	y_{24}	\hat{y}_{21}	\hat{y}_{22}	\hat{y}_{23}	\hat{y}_{24}
y_{31}	y_{32}	y_{33}	y_{34}	\hat{y}_{31}	\hat{y}_{32}	\hat{y}_{33}	\hat{y}_{34}
y_{41}	y_{42}	y_{43}	y_{44}	\hat{y}_{41}	\hat{y}_{42}	\hat{y}_{43}	\hat{y}_{44}
y_{51}	y_{52}	y_{53}	y_{54}	\hat{y}_{51}	\hat{y}_{52}	\hat{y}_{53}	\hat{y}_{54}
y_{61}	y_{62}	y_{63}	y_{64}	\hat{y}_{61}	\hat{y}_{62}	\hat{y}_{63}	\hat{y}_{64}

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y_{31}	y_{32}	y_{33}	y_{34}	\hat{y}_{31}	\hat{y}_{32}	\hat{y}_{33}	\hat{y}_{34}
y_{41}	y_{42}	y_{43}	y_{44}	\hat{y}_{41}	\hat{y}_{42}	\hat{y}_{43}	\hat{y}_{44}
y_{51}	y_{52}	y_{53}	y_{54}	\hat{y}_{51}	\hat{y}_{52}	\hat{y}_{53}	\hat{y}_{54}
y_{61}	y_{62}	y_{63}	y_{64}	\hat{y}_{61}	\hat{y}_{62}	\hat{y}_{63}	\hat{y}_{64}

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y_{21}	y_{22}	y_{23}	y_{24}	\hat{y}_{21}	\hat{y}_{22}	\hat{y}_{23}	\hat{y}_{24}
y_{31}	y_{32}	y_{33}	y_{34}	\hat{y}_{31}	\hat{y}_{32}	\hat{y}_{33}	\hat{y}_{34}
y_{41}	y_{42}	y_{43}	y_{44}	\hat{y}_{41}	\hat{y}_{42}	\hat{y}_{43}	\hat{y}_{44}
y_{51}	y_{52}	y_{53}	y_{54}	\hat{y}_{51}	\hat{y}_{52}	\hat{y}_{53}	\hat{y}_{54}
y_{61}	y_{62}	y_{63}	y_{64}	\hat{y}_{61}	\hat{y}_{62}	\hat{y}_{63}	\hat{y}_{64}

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y_{11}	y_{12}	y_{13}	y_{14}	\hat{y}_{11}	\hat{y}_{12}	\hat{y}_{13}	\hat{y}_{14}
y_{21}	y_{22}	y_{23}	y_{24}	\hat{y}_{21}	\hat{y}_{22}	\hat{y}_{23}	\hat{y}_{24}
y_{31}	y_{32}	y_{33}	y_{34}	\hat{y}_{31}	\hat{y}_{32}	\hat{y}_{33}	\hat{y}_{34}
y_{41}	y_{42}	y_{43}	y_{44}	\hat{y}_{41}	\hat{y}_{42}	\hat{y}_{43}	\hat{y}_{44}
y_{51}	y_{52}	y_{53}	y_{54}	\hat{y}_{51}	\hat{y}_{52}	\hat{y}_{53}	\hat{y}_{54}
y_{61}	y_{62}	y_{63}	y_{64}	\hat{y}_{61}	\hat{y}_{62}	\hat{y}_{63}	\hat{y}_{64}

- Example: Averaging the target losses.

$$L = \frac{1}{4} (L_1 + L_2 + L_3 + L_4)$$

MACRO- AND MICRO-LOSSES

- Micro-losses: The overall loss corresponds to aggregating the pointwise losses over the targets and the instances.

True scores				Predicted scores			
y_{11}	y_{12}	y_{13}	y_{14}	\hat{y}_{11}	\hat{y}_{12}	\hat{y}_{13}	\hat{y}_{14}
y_{21}	y_{22}	y_{23}	y_{24}	\hat{y}_{21}	\hat{y}_{22}	\hat{y}_{23}	\hat{y}_{24}
y_{31}	y_{32}	y_{33}	y_{34}	\hat{y}_{31}	\hat{y}_{32}	\hat{y}_{33}	\hat{y}_{34}
y_{41}	y_{42}	y_{43}	y_{44}	\hat{y}_{41}	\hat{y}_{42}	\hat{y}_{43}	\hat{y}_{44}
y_{51}	y_{52}	y_{53}	y_{54}	\hat{y}_{51}	\hat{y}_{52}	\hat{y}_{53}	\hat{y}_{54}
y_{61}	y_{62}	y_{63}	y_{64}	\hat{y}_{61}	\hat{y}_{62}	\hat{y}_{63}	\hat{y}_{64}

- Thus, we have

$$L = \sum_{i,j} \ell(y_{ij}, \hat{y}_{ij}),$$

where $\ell : \mathcal{Y} \times \mathcal{Y} \rightarrow \mathbb{R}$ in this case.

MACRO- AND MICRO-LOSSES

- Micro-losses: The overall loss corresponds to averaging the pointwise losses over the targets and the instances.

True scores				Predicted scores			
y_{11}	y_{12}		y_{14}	\hat{y}_{11}	\hat{y}_{12}		\hat{y}_{14}
y_{21}		y_{23}	y_{24}	\hat{y}_{21}		\hat{y}_{23}	\hat{y}_{24}
y_{31}	y_{32}	y_{33}	y_{34}	\hat{y}_{31}	\hat{y}_{32}	\hat{y}_{33}	\hat{y}_{34}
y_{41}		y_{43}	y_{44}	\hat{y}_{41}		\hat{y}_{43}	\hat{y}_{44}
y_{51}	y_{52}	y_{53}	y_{54}	\hat{y}_{51}	\hat{y}_{52}	\hat{y}_{53}	\hat{y}_{54}
	y_{62}	y_{63}			\hat{y}_{62}	\hat{y}_{63}	

- Thus, we have

$$L = \sum_{i,j} \ell(y_{ij}, \hat{y}_{ij}),$$

where $\ell : \mathcal{Y} \times \mathcal{Y} \rightarrow \mathbb{R}$ in this case.

- Can be used also for cases with missing entries.

INSTANCE-WISE LOSSES

- Instance-wise losses: Aggregating the losses over the instances.

True scores				Predicted scores			
y_{11}	y_{12}	y_{13}	y_{14}	\hat{y}_{11}	\hat{y}_{12}	\hat{y}_{13}	\hat{y}_{14}
y_{21}	y_{22}	y_{23}	y_{24}	\hat{y}_{21}	\hat{y}_{22}	\hat{y}_{23}	\hat{y}_{24}
y_{31}	y_{32}	y_{33}	y_{34}	\hat{y}_{31}	\hat{y}_{32}	\hat{y}_{33}	\hat{y}_{34}
y_{41}	y_{42}	y_{43}	y_{44}	\hat{y}_{41}	\hat{y}_{42}	\hat{y}_{43}	\hat{y}_{44}
y_{51}	y_{52}	y_{53}	y_{54}	\hat{y}_{51}	\hat{y}_{52}	\hat{y}_{53}	\hat{y}_{54}
y_{61}	y_{62}	y_{63}	y_{64}	\hat{y}_{61}	\hat{y}_{62}	\hat{y}_{63}	\hat{y}_{64}

- Example: Averaging over the instance-losses.

$$L = \frac{1}{6} \left(\ell(\mathbf{y}^{(1)}, \hat{\mathbf{y}}^{(1)}) + \ell(\mathbf{y}^{(2)}, \hat{\mathbf{y}}^{(2)}) + \ell(\mathbf{y}^{(3)}, \hat{\mathbf{y}}^{(3)}) + \right. \\ \left. \ell(\mathbf{y}^{(4)}, \hat{\mathbf{y}}^{(4)}) + \ell(\mathbf{y}^{(5)}, \hat{\mathbf{y}}^{(5)}) + \ell(\mathbf{y}^{(6)}, \hat{\mathbf{y}}^{(6)}) \right)$$

INSTANCE-WISE LOSSES

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True scores				Predicted scores			
y_{11}	y_{12}	y_{13}	y_{14}	\hat{y}_{11}	\hat{y}_{12}	\hat{y}_{13}	\hat{y}_{14}
y_{21}	y_{22}	y_{23}	y_{24}	\hat{y}_{21}	\hat{y}_{22}	\hat{y}_{23}	\hat{y}_{24}
y_{31}	y_{32}	y_{33}	y_{34}	\hat{y}_{31}	\hat{y}_{32}	\hat{y}_{33}	\hat{y}_{34}
y_{41}	y_{42}	y_{43}	y_{44}	\hat{y}_{41}	\hat{y}_{42}	\hat{y}_{43}	\hat{y}_{44}
y_{51}	y_{52}	y_{53}	y_{54}	\hat{y}_{51}	\hat{y}_{52}	\hat{y}_{53}	\hat{y}_{54}
y_{61}	y_{62}	y_{63}	y_{64}	\hat{y}_{61}	\hat{y}_{62}	\hat{y}_{63}	\hat{y}_{64}

- Example: Averaging over the instance-losses.

$$L = \frac{1}{6} \left(\ell(\mathbf{y}^{(1)}, \hat{\mathbf{y}}^{(1)}) + \ell(\mathbf{y}^{(2)}, \hat{\mathbf{y}}^{(2)}) + \ell(\mathbf{y}^{(3)}, \hat{\mathbf{y}}^{(3)}) + \right. \\ \left. \ell(\mathbf{y}^{(4)}, \hat{\mathbf{y}}^{(4)}) + \ell(\mathbf{y}^{(5)}, \hat{\mathbf{y}}^{(5)}) + \ell(\mathbf{y}^{(6)}, \hat{\mathbf{y}}^{(6)}) \right)$$

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True scores				Predicted scores			
y_{11}	y_{12}	y_{13}	y_{14}	\hat{y}_{11}	\hat{y}_{12}	\hat{y}_{13}	\hat{y}_{14}
y_{21}	y_{22}	y_{23}	y_{24}	\hat{y}_{21}	\hat{y}_{22}	\hat{y}_{23}	\hat{y}_{24}
y_{31}	y_{32}	y_{33}	y_{34}	\hat{y}_{31}	\hat{y}_{32}	\hat{y}_{33}	\hat{y}_{34}
y_{41}	y_{42}	y_{43}	y_{44}	\hat{y}_{41}	\hat{y}_{42}	\hat{y}_{43}	\hat{y}_{44}
y_{51}	y_{52}	y_{53}	y_{54}	\hat{y}_{51}	\hat{y}_{52}	\hat{y}_{53}	\hat{y}_{54}
y_{61}	y_{62}	y_{63}	y_{64}	\hat{y}_{61}	\hat{y}_{62}	\hat{y}_{63}	\hat{y}_{64}

- Example: Averaging over the instance-losses.

$$L = \frac{1}{6} \left(\ell(\mathbf{y}^{(1)}, \hat{\mathbf{y}}^{(1)}) + \ell(\mathbf{y}^{(2)}, \hat{\mathbf{y}}^{(2)}) + \ell(\mathbf{y}^{(3)}, \hat{\mathbf{y}}^{(3)}) + \right. \\ \left. \ell(\mathbf{y}^{(4)}, \hat{\mathbf{y}}^{(4)}) + \ell(\mathbf{y}^{(5)}, \hat{\mathbf{y}}^{(5)}) + \ell(\mathbf{y}^{(6)}, \hat{\mathbf{y}}^{(6)}) \right)$$

INSTANCE-WISE LOSSES

- Instance-wise losses: Aggregating the losses over the instances.

True scores				Predicted scores			
y_{11}	y_{12}	y_{13}	y_{14}	\hat{y}_{11}	\hat{y}_{12}	\hat{y}_{13}	\hat{y}_{14}
y_{21}	y_{22}	y_{23}	y_{24}	\hat{y}_{21}	\hat{y}_{22}	\hat{y}_{23}	\hat{y}_{24}
y_{31}	y_{32}	y_{33}	y_{34}	\hat{y}_{31}	\hat{y}_{32}	\hat{y}_{33}	\hat{y}_{34}
y_{41}	y_{42}	y_{43}	y_{44}	\hat{y}_{41}	\hat{y}_{42}	\hat{y}_{43}	\hat{y}_{44}
y_{51}	y_{52}	y_{53}	y_{54}	\hat{y}_{51}	\hat{y}_{52}	\hat{y}_{53}	\hat{y}_{54}
y_{61}	y_{62}	y_{63}	y_{64}	\hat{y}_{61}	\hat{y}_{62}	\hat{y}_{63}	\hat{y}_{64}

- Example: Averaging over the instance-losses.

$$L = \frac{1}{6} \left(\ell(\mathbf{y}^{(1)}, \hat{\mathbf{y}}^{(1)}) + \ell(\mathbf{y}^{(2)}, \hat{\mathbf{y}}^{(2)}) + \ell(\mathbf{y}^{(3)}, \hat{\mathbf{y}}^{(3)}) + \right. \\ \left. \ell(\mathbf{y}^{(4)}, \hat{\mathbf{y}}^{(4)}) + \ell(\mathbf{y}^{(5)}, \hat{\mathbf{y}}^{(5)}) + \ell(\mathbf{y}^{(6)}, \hat{\mathbf{y}}^{(6)}) \right)$$

INSTANCE-WISE LOSSES

- Instance-wise losses: Aggregating the losses over the instances.

True scores				Predicted scores			
y_{11}	y_{12}	y_{13}	y_{14}	\hat{y}_{11}	\hat{y}_{12}	\hat{y}_{13}	\hat{y}_{14}
y_{21}	y_{22}	y_{23}	y_{24}	\hat{y}_{21}	\hat{y}_{22}	\hat{y}_{23}	\hat{y}_{24}
y_{31}	y_{32}	y_{33}	y_{34}	\hat{y}_{31}	\hat{y}_{32}	\hat{y}_{33}	\hat{y}_{34}
y_{41}	y_{42}	y_{43}	y_{44}	\hat{y}_{41}	\hat{y}_{42}	\hat{y}_{43}	\hat{y}_{44}
y_{51}	y_{52}	y_{53}	y_{54}	\hat{y}_{51}	\hat{y}_{52}	\hat{y}_{53}	\hat{y}_{54}
y_{61}	y_{62}	y_{63}	y_{64}	\hat{y}_{61}	\hat{y}_{62}	\hat{y}_{63}	\hat{y}_{64}

- Example: Averaging over the instance-losses.

$$L = \frac{1}{6} \left(\ell(\mathbf{y}^{(1)}, \hat{\mathbf{y}}^{(1)}) + \ell(\mathbf{y}^{(2)}, \hat{\mathbf{y}}^{(2)}) + \ell(\mathbf{y}^{(3)}, \hat{\mathbf{y}}^{(3)}) + \right. \\ \left. \ell(\mathbf{y}^{(4)}, \hat{\mathbf{y}}^{(4)}) + \ell(\mathbf{y}^{(5)}, \hat{\mathbf{y}}^{(5)}) + \ell(\mathbf{y}^{(6)}, \hat{\mathbf{y}}^{(6)}) \right)$$

INSTANCE-WISE LOSSES

- Instance-wise losses: Aggregating the losses over the instances.

True scores				Predicted scores			
y_{11}	y_{12}	y_{13}	y_{14}	\hat{y}_{11}	\hat{y}_{12}	\hat{y}_{13}	\hat{y}_{14}
y_{21}	y_{22}	y_{23}	y_{24}	\hat{y}_{21}	\hat{y}_{22}	\hat{y}_{23}	\hat{y}_{24}
y_{31}	y_{32}	y_{33}	y_{34}	\hat{y}_{31}	\hat{y}_{32}	\hat{y}_{33}	\hat{y}_{34}
y_{41}	y_{42}	y_{43}	y_{44}	\hat{y}_{41}	\hat{y}_{42}	\hat{y}_{43}	\hat{y}_{44}
y_{51}	y_{52}	y_{53}	y_{54}	\hat{y}_{51}	\hat{y}_{52}	\hat{y}_{53}	\hat{y}_{54}
y_{61}	y_{62}	y_{63}	y_{64}	\hat{y}_{61}	\hat{y}_{62}	\hat{y}_{63}	\hat{y}_{64}

- Example: Averaging over the instance-losses.

$$L = \frac{1}{6} \left(\ell(\mathbf{y}^{(1)}, \hat{\mathbf{y}}^{(1)}) + \ell(\mathbf{y}^{(2)}, \hat{\mathbf{y}}^{(2)}) + \ell(\mathbf{y}^{(3)}, \hat{\mathbf{y}}^{(3)}) + \right. \\ \left. \ell(\mathbf{y}^{(4)}, \hat{\mathbf{y}}^{(4)}) + \ell(\mathbf{y}^{(5)}, \hat{\mathbf{y}}^{(5)}) + \ell(\mathbf{y}^{(6)}, \hat{\mathbf{y}}^{(6)}) \right)$$

INSTANCE-WISE LOSSES

- Instance-wise losses: Aggregating the losses over the instances.

True scores				Predicted scores			
y_{11}	y_{12}	y_{13}	y_{14}	\hat{y}_{11}	\hat{y}_{12}	\hat{y}_{13}	\hat{y}_{14}
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y_{31}	y_{32}	y_{33}	y_{34}	\hat{y}_{31}	\hat{y}_{32}	\hat{y}_{33}	\hat{y}_{34}
y_{41}	y_{42}	y_{43}	y_{44}	\hat{y}_{41}	\hat{y}_{42}	\hat{y}_{43}	\hat{y}_{44}
y_{51}	y_{52}	y_{53}	y_{54}	\hat{y}_{51}	\hat{y}_{52}	\hat{y}_{53}	\hat{y}_{54}
y_{61}	y_{62}	y_{63}	y_{64}	\hat{y}_{61}	\hat{y}_{62}	\hat{y}_{63}	\hat{y}_{64}

- Example: Averaging over the instance-losses.

$$L = \frac{1}{6} \left(\ell(\mathbf{y}^{(1)}, \hat{\mathbf{y}}^{(1)}) + \ell(\mathbf{y}^{(2)}, \hat{\mathbf{y}}^{(2)}) + \ell(\mathbf{y}^{(3)}, \hat{\mathbf{y}}^{(3)}) + \right. \\ \left. \ell(\mathbf{y}^{(4)}, \hat{\mathbf{y}}^{(4)}) + \ell(\mathbf{y}^{(5)}, \hat{\mathbf{y}}^{(5)}) + \ell(\mathbf{y}^{(6)}, \hat{\mathbf{y}}^{(6)}) \right)$$