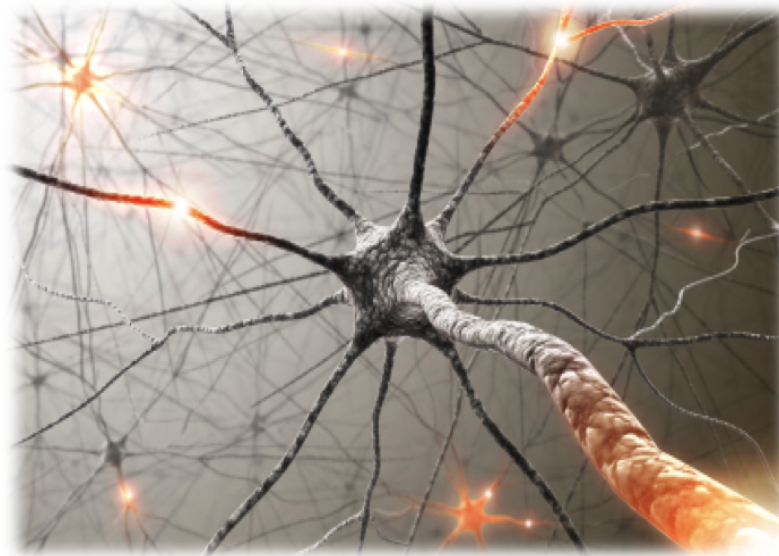

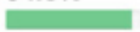



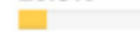














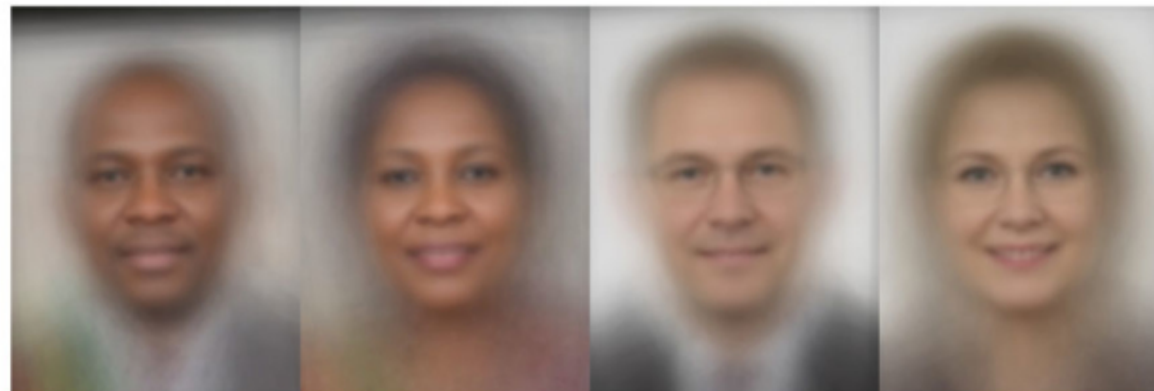


Machine learning: group DRO



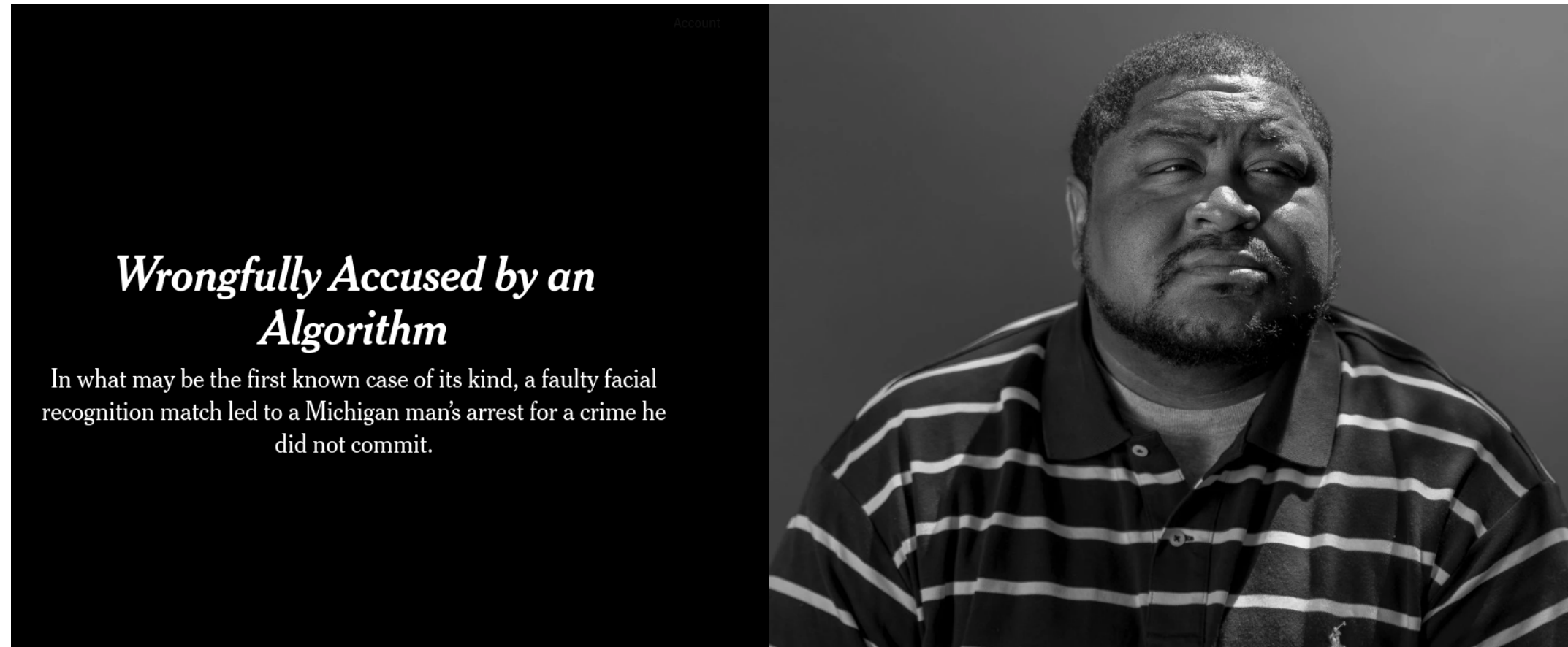
Gender Shades

Gender Classifier	Darker Male	Darker Female	Lighter Male	Lighter Female	Largest Gap
 Microsoft	94.0% 	79.2% 	100% 	98.3% 	20.8% 
 FACE++	99.3% 	65.5% 	99.2% 	94.0% 	33.8% 
 IBM	88.0% 	65.3% 	99.7% 	92.9% 	34.4% 



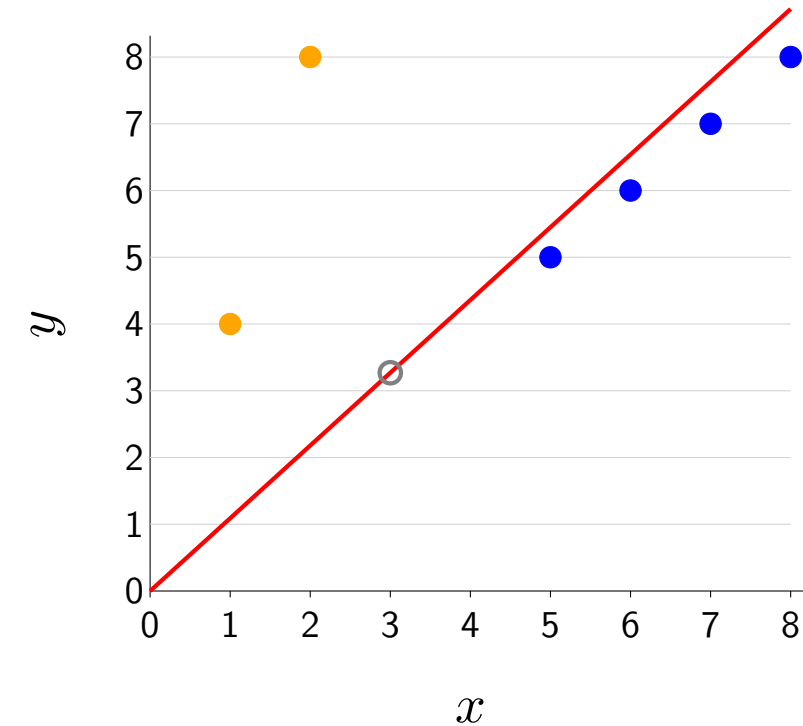
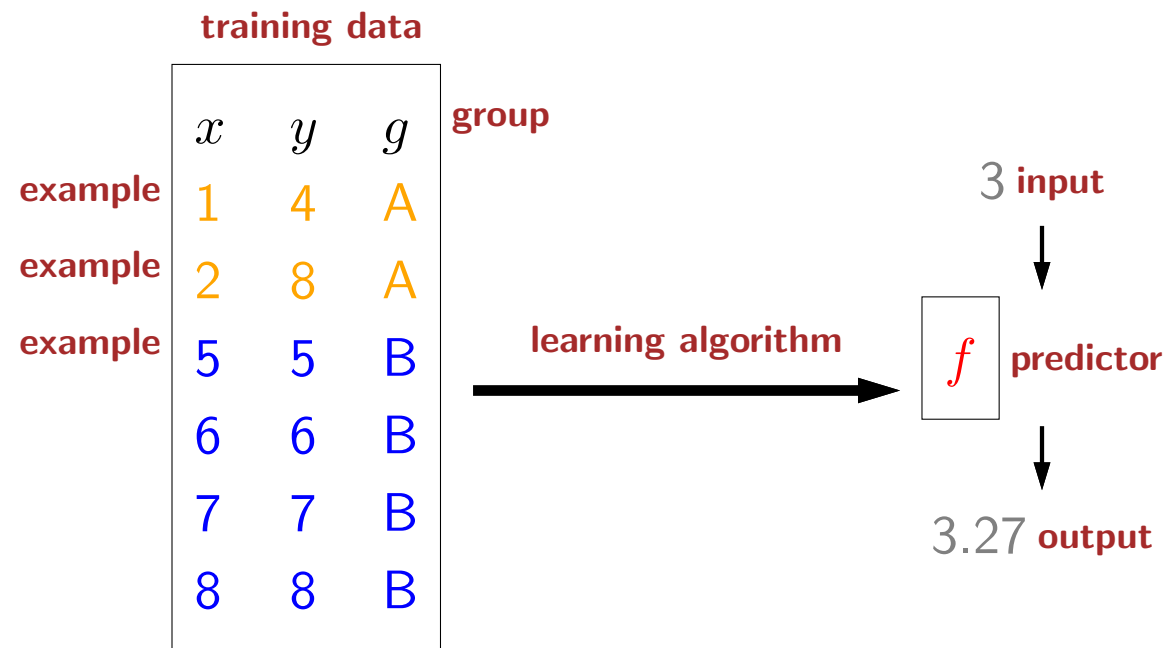
Inequalities arise in machine learning

False arrest due to facial recognition



Real-life consequences

Linear regression with groups



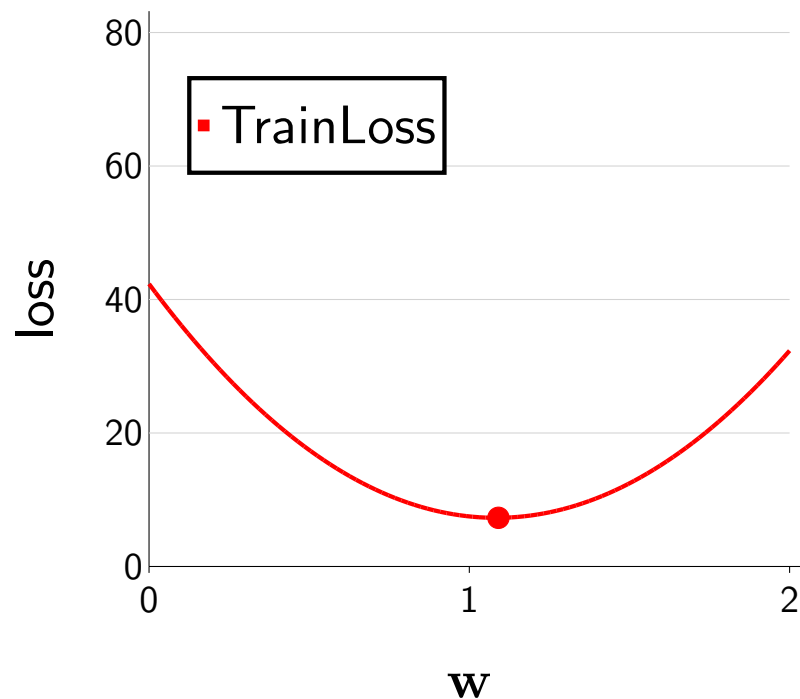
$$f_{\mathbf{w}}(x) = \mathbf{w} \cdot \phi(x) \quad \mathbf{w} = [w] \quad \phi(x) = [x]$$

Note: predictor $f_{\mathbf{w}}$ does not use group information g

Average loss

$$\text{Loss}(x, y, \mathbf{w}) = (f_{\mathbf{w}}(x) - y)^2$$

x	y	g
1	4	A
2	8	A
5	5	B
6	6	B
7	7	B
8	8	B

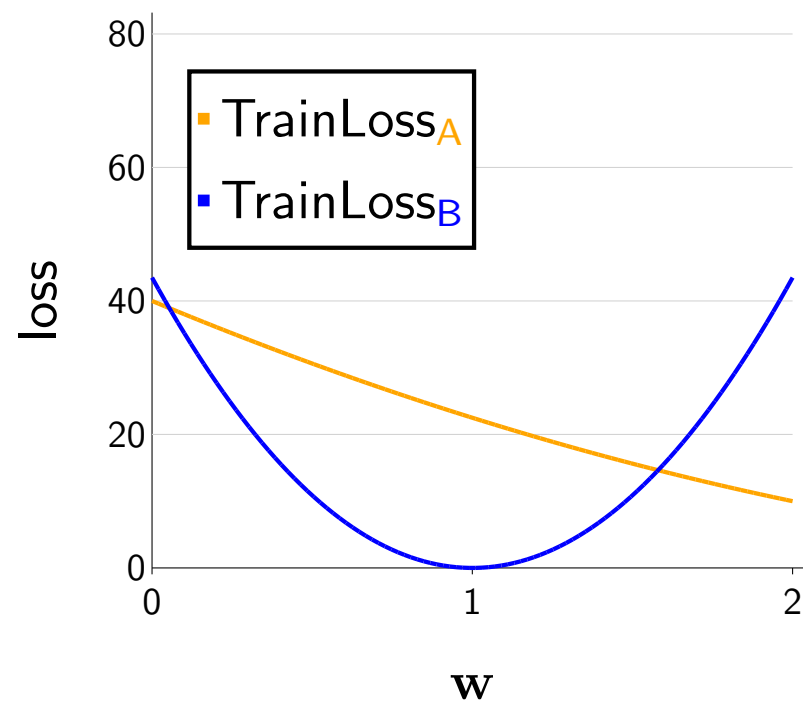


$$\text{TrainLoss}(\mathbf{w}) = \frac{1}{|\mathcal{D}_{\text{train}}|} \sum_{(x, y) \in \mathcal{D}_{\text{train}}} \text{Loss}(x, y, \mathbf{w})$$

$$\text{TrainLoss}(1) = \frac{1}{6}((1 - 4)^2 + (2 - 8)^2 + (5 - 5)^2 + (6 - 6)^2 + (7 - 7)^2 + (8 - 8)^2) = 7.5$$

Per-group loss

x	y	g
1	4	A
2	8	A
5	5	B
6	6	B
7	7	B
8	8	B



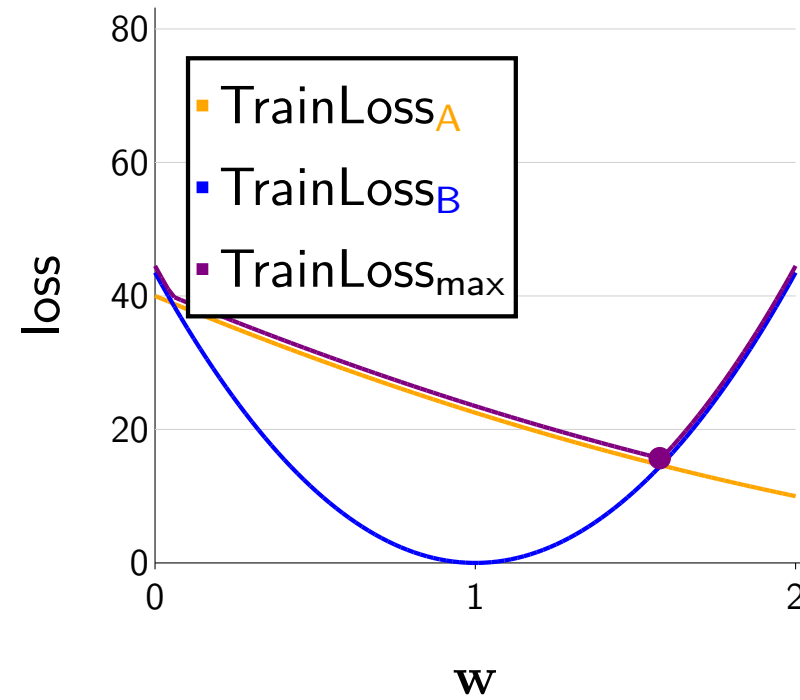
$$\text{TrainLoss}_g(\mathbf{w}) = \frac{1}{|\mathcal{D}_{\text{train}}(g)|} \sum_{(x,y) \in \mathcal{D}_{\text{train}}(g)} \text{Loss}(x, y, \mathbf{w})$$

$$\text{TrainLoss}_A(1) = \frac{1}{2}((1-4)^2 + (2-8)^2) = 22.5$$

$$\text{TrainLoss}_B(1) = \frac{1}{4}((5-5)^2 + (6-6)^2 + (7-7)^2 + (8-8)^2) = 0$$

Maximum group loss

x	y	g
1	4	A
2	8	A
5	5	B
6	6	B
7	7	B
8	8	B



$$\text{TrainLoss}_{\max}(\mathbf{w}) = \max_g \text{TrainLoss}_g(\mathbf{w})$$

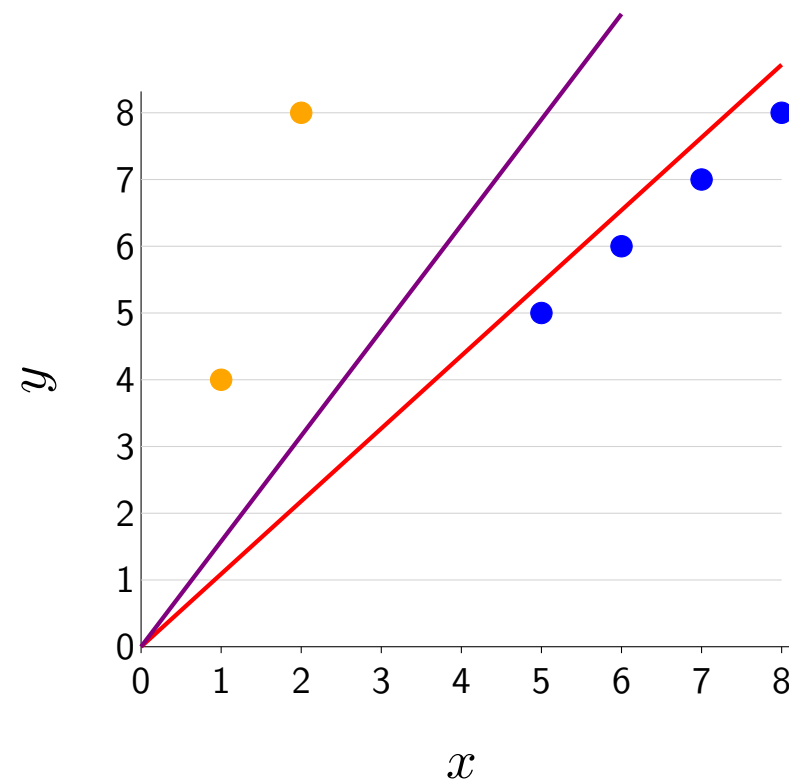
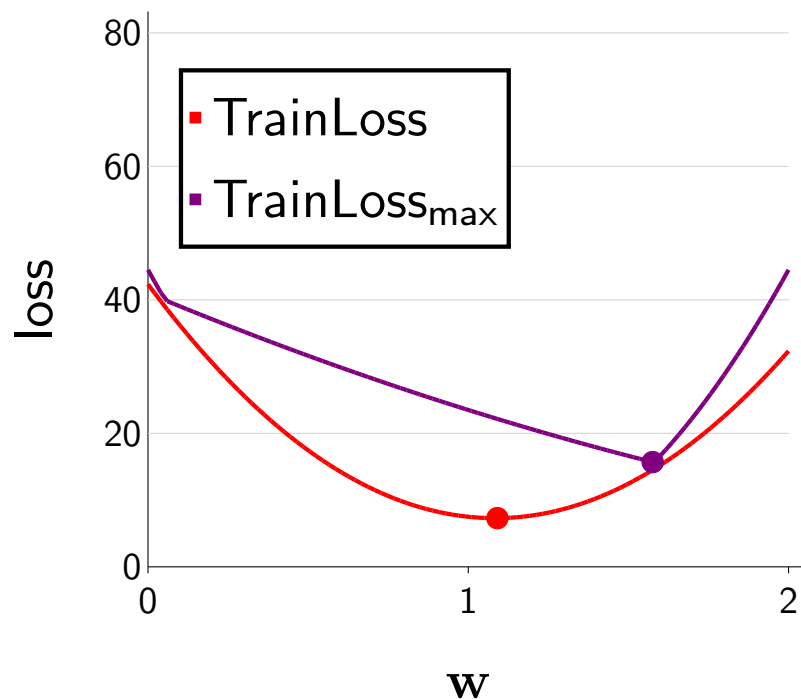
$$\text{TrainLoss}_A(1) = 22.5$$

$$\text{TrainLoss}_B(1) = 0$$

$$\text{TrainLoss}_{\max}(1) = \max(22.5, 0) = 22.5$$

Average loss versus maximum group loss

x	y	g
1	4	A
2	8	A
5	5	B
6	6	B
7	7	B
8	8	B



Standard learning:

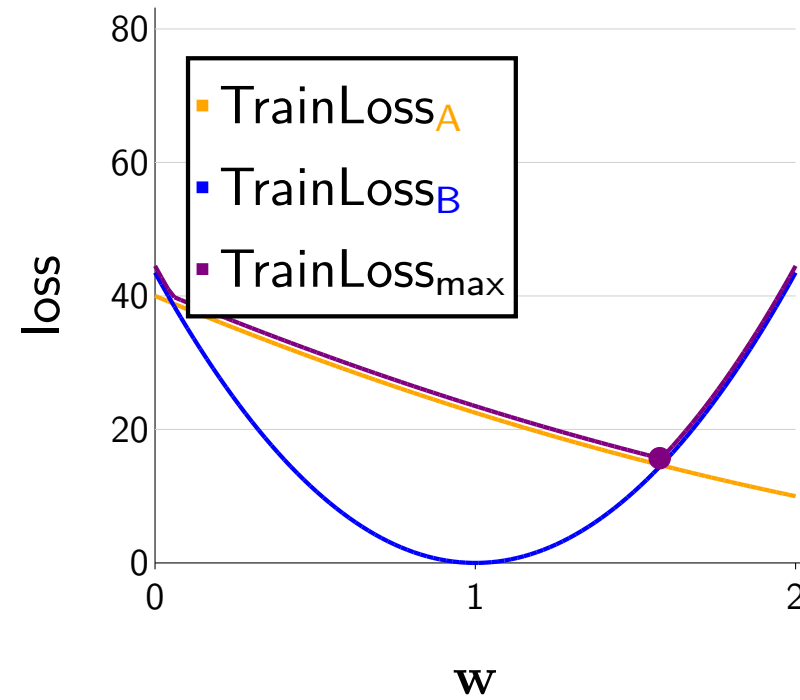
minimizer of average loss: $w = 1.09$

Group distributionally robust optimization (group DRO):

minimizer of maximum group loss: $w = 1.58$

Training via gradient descent

x	y	g
1	4	A
2	8	A
5	5	B
6	6	B
7	7	B
8	8	B



$$\text{TrainLoss}_{\max}(\mathbf{w}) = \max_g \text{TrainLoss}_g(\mathbf{w})$$

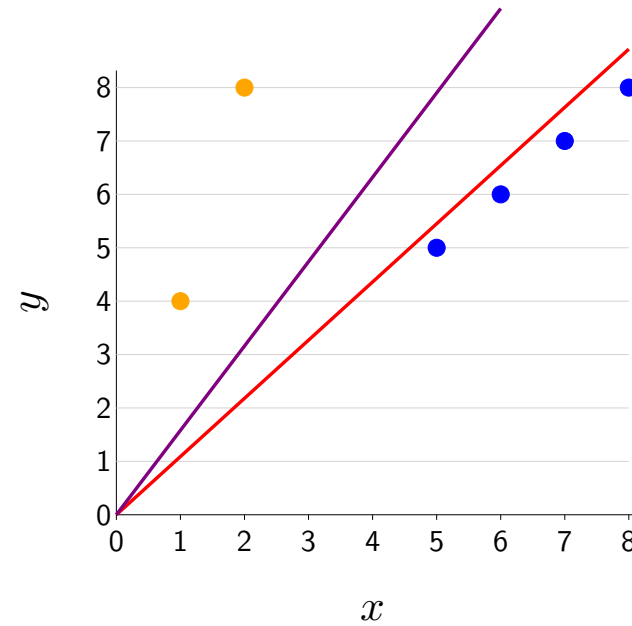
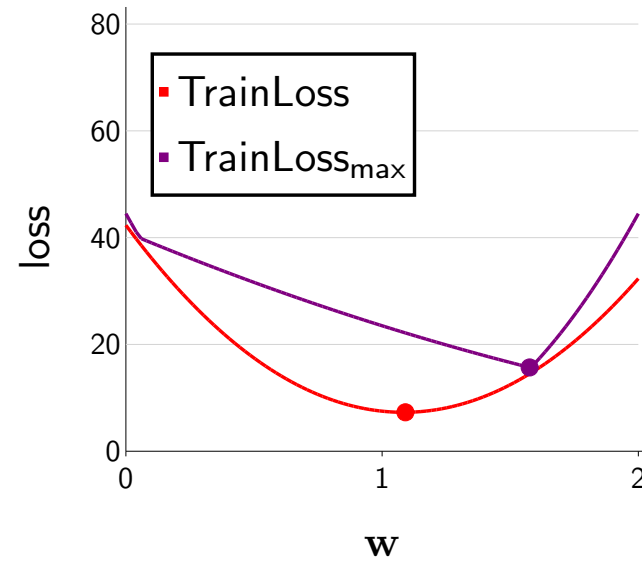
$$\nabla \text{TrainLoss}_{\max}(\mathbf{w}) = \nabla \text{TrainLoss}_{g^*}(\mathbf{w})$$

$$\text{where } g^* = \arg \max_g \text{TrainLoss}_g(\mathbf{w})$$



Summary

x	y	g
1	4	A
2	8	A
5	5	B
6	6	B
7	7	B
8	8	B



- Maximum group loss \neq average loss
- Group DRO: minimize the maximum group loss
- Many more nuances: intersectionality? don't know groups? overfitting?