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Instance Segmentation of Dense Objects via Deep Pixel Embedding



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Loss Function: Cartesian

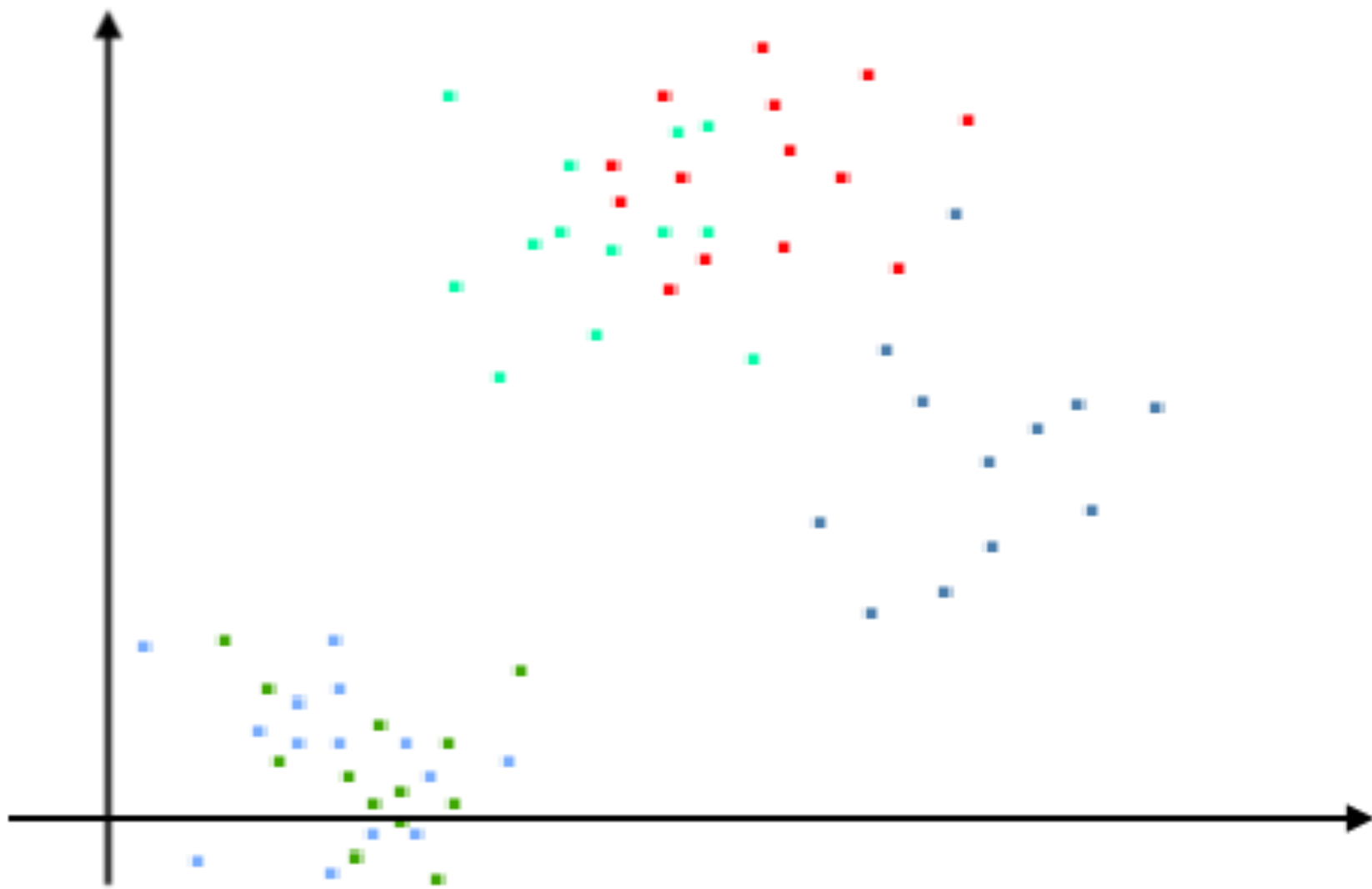
Deep Pixel Embedding



De Brabandere, B., Neven, D., Van Gool, L., & ESAT-PSI. Kl. Semantic Instance Segmentation with a Discriminative Loss Function

$$L_{intra} = \frac{1}{C} \sum_{c=1}^C \frac{1}{E_c} \sum_{i=1}^{E_c} \left[\|\mu_c - e_i\| - \delta_2 \right]_+^2$$

$$L_{inter} = \frac{1}{C(C-1)} \sum_{\substack{c_A=1 \\ c_A \neq c_B}}^C \sum_{c_B=1}^C \left[\|\mu_{c_A} - \mu_{c_B}\| - 2\delta_1 \right]_+^2$$



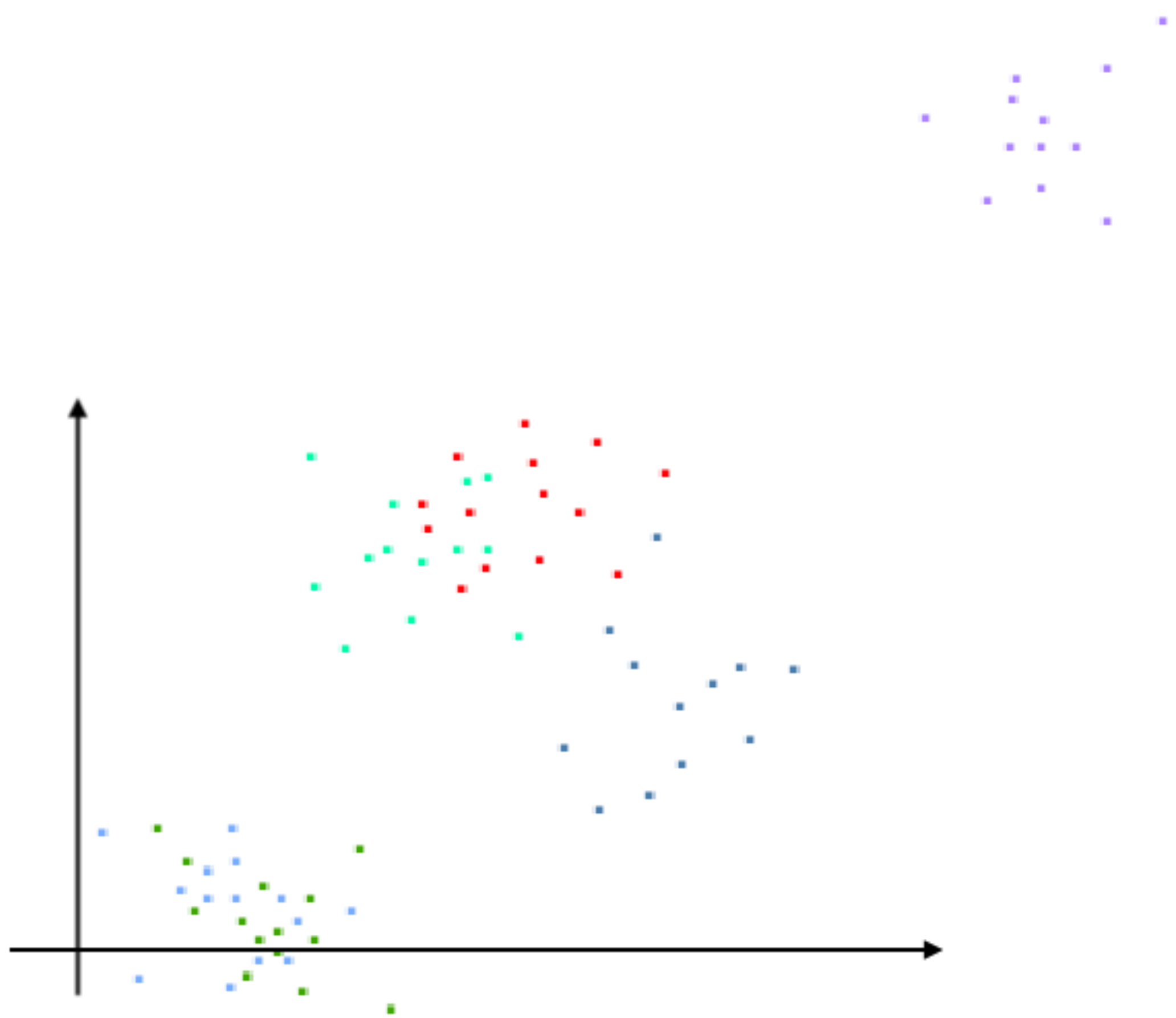
Centres





Embedding





Loss Function: Cartesian

Deep Pixel Embedding

$$L_{inter} = \frac{1}{C(C-1)} \sum_{\substack{c_A=1 \\ c_A \neq c_B}}^C \sum_{c_B=1}^C \left[\|\mu_{c_A} - \mu_{c_B}\| - 2\delta_1 \right]_+^2$$

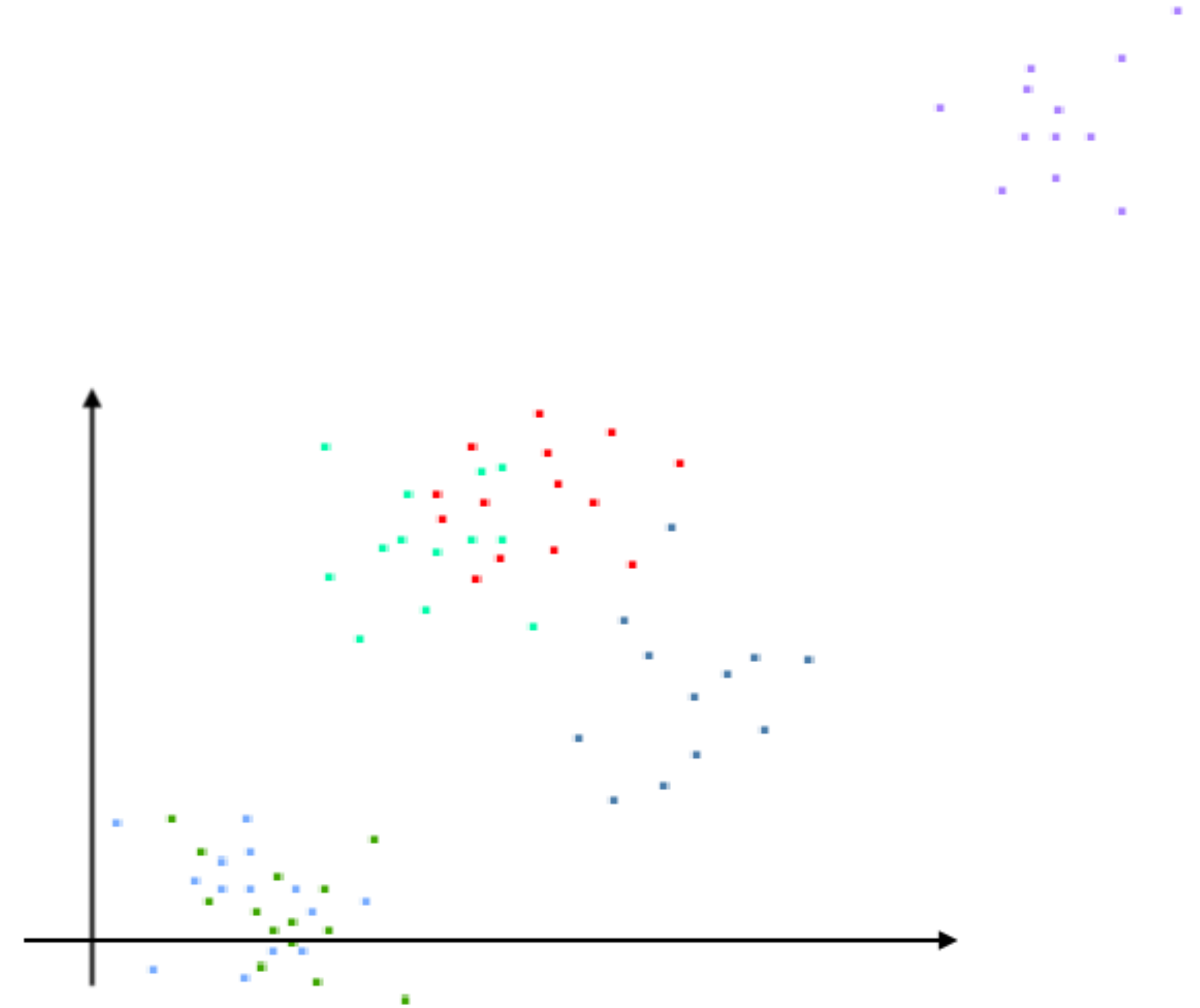
Centres

(Arrows point from 'Centres' to μ_{c_A} and μ_{c_B} in the formula)

$$L_{intra} = \frac{1}{C} \sum_{c=1}^C \frac{1}{E_c} \sum_{i=1}^{E_c} \left[\|\mu_c - e_i\| - \delta_2 \right]_+^2$$

Embedding

(Arrow points from 'Embedding' to e_i in the formula)



De Brabandere, B., Neven, D., Van Gool, L., & ESAT-PSI, K. U. *Semantic Instance Segmentation with a Discriminative Loss Function.*

Neighbours

$$L_{inter} = \frac{1}{C} \sum_{c_A=1}^C \frac{1}{|\mathbf{N}_{c_A}|} \sum_{c_B \in \mathbf{N}_{c_A}} \left[\text{CosS}(\mu_{c_A}, \mu_{c_B}) \right]$$

$$L_{intra} = \frac{1}{C} \sum_{c=1}^C \frac{1}{E_c} \sum_{i=1}^{E_c} \left[1 - \text{CosS}(e_i, \mu_c) \right]$$

$$\text{CosS}(a, b) = \frac{a \cdot b}{\|a\|_2 \|b\|_2}$$

Chen, L., Strauch, M., & Merhof, D. *Instance Segmentation of Biomedical Images with an Object-Aware Embedding Learned with Local Constraints*.