

Tile EM Model

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1 Problem Formulation

$\mathcal{T} = \{t_k\}$ is the set of all non-overlapping tiles for an object i , where T is the ground truth tile set and T' is some combination of tiles chosen from \mathcal{T} .

We define the ground truth as the tile combination T' that maximizes the probability that it would be ground truth, given the set of worker qualities Q_j and tile indicator labels l_{kj} . The indicator label l_{kj} is one when worker votes on the tile t_{kj} (i.e. the bounding box that he draws contains t_{kj}), and zero otherwise. Given these definitions, the objective function of the segmentation problem is formulated as:

$$T = \arg \max_{T' \subseteq \mathcal{T}} P(T = T' | l_{kj}, Q_j) \quad (1)$$

Using Bayes rule we can rewrite this in terms of the posterior probability of the tile-based values(l_{kj}) or worker-based values(Q_j), which we can use for the E and M step equations respectively.

2 Inference

For the E step, we fix T and estimate the Q_j parameters. We can rewrite Eq.1 as:

$$p(T' | Q_j, l_{kj}) = \frac{p(Q_j | l_{kj}, T') P(T')}{p(Q_j)} \quad (2)$$

Our worker error model follows a Bernoulli distribution by describing worker quality $Q_j = q_j$ as the probability that the worker makes a correct vote for a tile ($w=1$), otherwise $Q_j = 1 - q_j$ for an incorrect guess.

$$p(w) = q_j p(w = 1) + (1 - q_j) p(w = 0) \quad (3)$$

$$(4)$$

There is a closed form of the maximum likelihood solution for the Bernoulli distribution.

$$\mathcal{L} = \ln p(w) = p(w = 1) \ln q_j + p(w = 0) \ln(1 - q_j) \quad (5)$$

$$\frac{\partial \mathcal{L}}{\partial q_j} = \frac{p(w = 1)}{\hat{q}_j} - \frac{p(w = 0)}{1 - \hat{q}_j} = 0 \quad (6)$$

$$(7)$$

Solving for \hat{q}_j :

$$\hat{q}_j = \frac{p(w = 1)}{p(w = 0) + p(w = 1)} = \frac{\frac{n_{w=1}}{n_{total}}}{\frac{n_{w=0}}{n_{total}} + \frac{n_{w=1}}{n_{total}}} \quad (8)$$

$$\boxed{\hat{q}_j = \frac{n_{correct}}{n_{total}}} \quad (9)$$

For the M step, we maximize the likelihood of the tile combination T' for a fixed set of worker qualities, $\{Q_j\}$. Following Eq. 1,

$$p(T'|Q_j, l_{kj}) = \frac{p(l_{kj}|Q_j, l_k)P(\cancel{T'})}{\cancel{p(l_{kj})}} = p(l_{kj}|Q_j, l_k) \quad (10)$$

We can decompose T' into its tile components:

$$= \prod_{t_k \in T'} p(t_k \in T|Q_j, l_k) \prod_{t_k \notin T'} p(t_k \notin T|Q_j, l_k) \quad (11)$$

We can evaluate these quantities by the observed indicator labels. The worker makes a correct response in the scenario where they vote for a tile that actually lies within T or they don't vote for a tile that is excluded from T. In that case, the probability of the tile being in T is the probability that the worker had made the correct response: $p(l_k) = q_j$. Likewise, a worker makes an incorrect response when their vote contradicts with the inclusion of the tile in T ($\{t_k \in T \& l_{kj} = 0\}, \{t_k \notin T \& l_{kj} = 1\}$), in which case the tile is assigned a probability of $p(l_k) = 1 - q_j$.