Let

- $\mu, \nu \in \Delta^n$ be the source and the target measures,
- $C \in \mathbb{R}^{n \times n}_+$ be the cost matrix,
- $\gamma > 0$ be the entropy regularization parameter,
- H(P) denote the negative entropy for a plan $P \in \mathbb{R}^{n \times n}$, i.e., $H(P) := \sum_{i,j} P_{ij} \ln P_{ij}$.

Consider the OT problem:

$$\min_{P \in \mathbb{R}^{n \times n}} \langle C, P \rangle + \gamma H(P)
\text{s.t. } P \ge 0,
P \mathbf{1} = \mu,
P^{\top} \mathbf{1} = \nu.$$
(1)

Proposition 0.1. Let $\beta^* \in \mathbb{R}^n$ be the solution to the dual problem

$$\max_{\beta} \ -\nu^{\top}\beta - \gamma \sum_{i} \mu_{i} \ln \left(\sum_{j} \exp \left(\frac{-C_{ij} - \beta_{j}}{\gamma} \right) \right),$$

then the optimal plan P^* for (1) can be recovered as follows

$$p_{ij}^* = \mu_i \frac{\exp\left(\frac{-C_{ij} - \beta_j^*}{\gamma}\right)}{\sum_k \exp\left(\frac{-C_{ik} - \beta_k^*}{\gamma}\right)}.$$

Lagrangian:

$$L(P, \Lambda, \alpha, \beta) = \langle C, P \rangle + \gamma H(P) - \langle \Lambda, P \rangle + \alpha^{\top} (P \mathbf{1} - \mu) + \beta^{\top} (P^{\top} \mathbf{1} - \nu)$$
$$= -\mu^{\top} \alpha - \nu^{\top} \beta - \langle \Lambda - C - \alpha \mathbf{1}^{\top} - \mathbf{1}^{\top} \beta, P \rangle + \gamma H(P)$$

Dual function:

$$\hat{g}(\Lambda, \alpha, \beta) = -\mu^{\top} \alpha - \nu^{\top} \beta - \underbrace{\sup_{P} \left[\langle \Lambda - C - \alpha \mathbf{1}^{\top} - \mathbf{1}^{\top} \beta, P \rangle - \gamma H(P) \right]}_{=:h(\Lambda, \alpha, \beta)}.$$

Note that

$$h(\Lambda, \alpha, \beta) = \gamma \sum_{i,j} \sup_{p \in \mathbb{R}} \left[\frac{\Lambda_{ij} - C_{ij} - \alpha_i - \beta_j}{\gamma} p - p \ln p \right].$$

The supremum is attained at a point

$$p_{ij}^* = \exp\left(\frac{\Lambda_{ij} - C_{ij} - \alpha_i - \beta_j}{\gamma} - 1\right),\tag{2}$$

and it holds

$$h(\Lambda, \alpha, \beta) = \gamma \sum_{i,j} \exp\left(\frac{\Lambda_{ij} - C_{ij} - \alpha_i - \beta_j}{\gamma} - 1\right).$$

Thus, we have

$$\hat{g}(\Lambda, \alpha, \beta) = -\mu^{\top} \alpha - \nu^{\top} \beta - \gamma \sum_{i,j} \exp\left(\frac{\Lambda_{ij} - C_{ij} - \alpha_i - \beta_j}{\gamma} - 1\right).$$

The dual problem reads as

$$\max_{\Lambda \geq 0, \alpha, \beta} \hat{g}(\Lambda, \alpha, \beta).$$

Observe that for any pair α, β , the function $\hat{g}(\cdot, \alpha, \beta)$ achieves its maximum at $\Lambda = 0$. We therefore consider the function $\tilde{g}(\alpha, \beta) := \hat{g}(0, \alpha, \beta)$. It holds

$$\frac{\partial \tilde{g}}{\partial \alpha_i} = -\mu_i + e^{-\alpha_i/\gamma} \sum_j \exp\left(\frac{-C_{ij} - \beta_j}{\gamma} - 1\right),\,$$

$$\frac{\partial \tilde{g}}{\partial \alpha_{i}} = 0 \iff e^{-\alpha_{i}^{*}(\beta)/\gamma} = \frac{\mu_{i}}{\sum_{j} \exp\left(\frac{-C_{ij} - \beta_{j}}{\gamma} - 1\right)}
\iff \alpha_{i}^{*}(\beta) = \gamma \ln\left(\sum_{j} \exp\left(\frac{-C_{ij} - \beta_{j}}{\gamma} - 1\right)\right) - \gamma \ln \mu_{i}
= \gamma \ln\left(\sum_{j} \exp\left(\frac{-C_{ij} - \beta_{j}}{\gamma}\right)\right) - \gamma - \gamma \ln \mu_{i}.$$
(3)

Thus, we reduce the problem to maximization of the function

$$g(\beta) := \tilde{g}(\alpha^*(\beta), \beta) = -\gamma \sum_{i} \mu_i \ln \left(\sum_{j} \exp \left(\frac{-C_{ij} - \beta_j}{\gamma} \right) \right) + \gamma + \gamma H(\mu) - \nu^\top \beta$$
$$-\gamma \sum_{i} e^{-\alpha_i^*(\beta)/\gamma} \sum_{j} \exp \left(\frac{-C_{ij} - \beta_j}{\gamma} - 1 \right)$$
$$= -\gamma \sum_{i} \mu_i \ln \left(\sum_{j} \exp \left(\frac{-C_{ij} - \beta_j}{\gamma} \right) \right) + \gamma H(\mu) - \nu^\top \beta.$$

Thus, we arrive at the dual problem

$$\max_{\beta} \ -\nu^{\top}\beta - \gamma \sum_{i} \mu_{i} \ln \left(\sum_{j} \exp \left(\frac{-C_{ij} - \beta_{j}}{\gamma} \right) \right).$$

If β^* is the optimal dual variable, we can recover the optimal plan using (2) and (3):

$$p_{ij}^* = e^{-\alpha_i^*(\beta)/\gamma} \exp\left(\frac{-C_{ij} - \beta_j}{\gamma} - 1\right) = \mu_i \frac{\exp\left(\frac{-C_{ij} - \beta_j}{\gamma}\right)}{\sum_k \exp\left(\frac{-C_{ik} - \beta_k}{\gamma}\right)}.$$