



IBM Research

Alleviating Noisy Data in Image Captioning with Cooperative Distillation

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Motivation: Leverage clean and noisy datasets



Background

- ✓ Multimodal datasets with noisy labels are ubiquitous, cheap to collect and available abundantly at scale
- ✗ Clean multimodal data is expensive to collect and available at a smaller scale



Goal

- Leverage noisy and clean datasets to **alleviate shortcomings** of each while **benefiting from** their respective **strengths**:
- Overcome the noise barrier
 - Improve the accuracy of the trained models

Setup: Teacher and student models each learning on their respective datasets

- Student model \mathcal{S} learning from noisy dataset \mathcal{S} (we used **Google Conceptual Captioning**¹, GCC, for the student dataset)
- Teacher model \mathcal{T} learning from a clean dataset \mathcal{T} (we use **MS COCO**² for the teacher dataset)
- Both teacher and student are transformer models

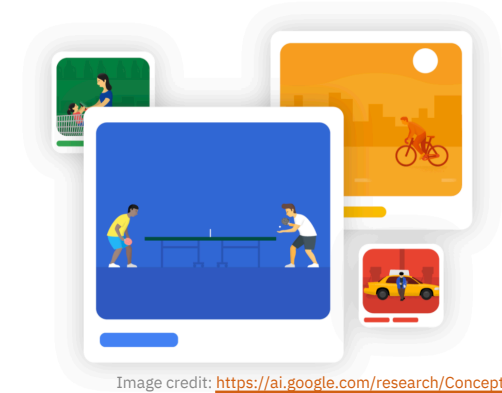


Image credit: <https://ai.google.com/research/ConceptualCaptions>

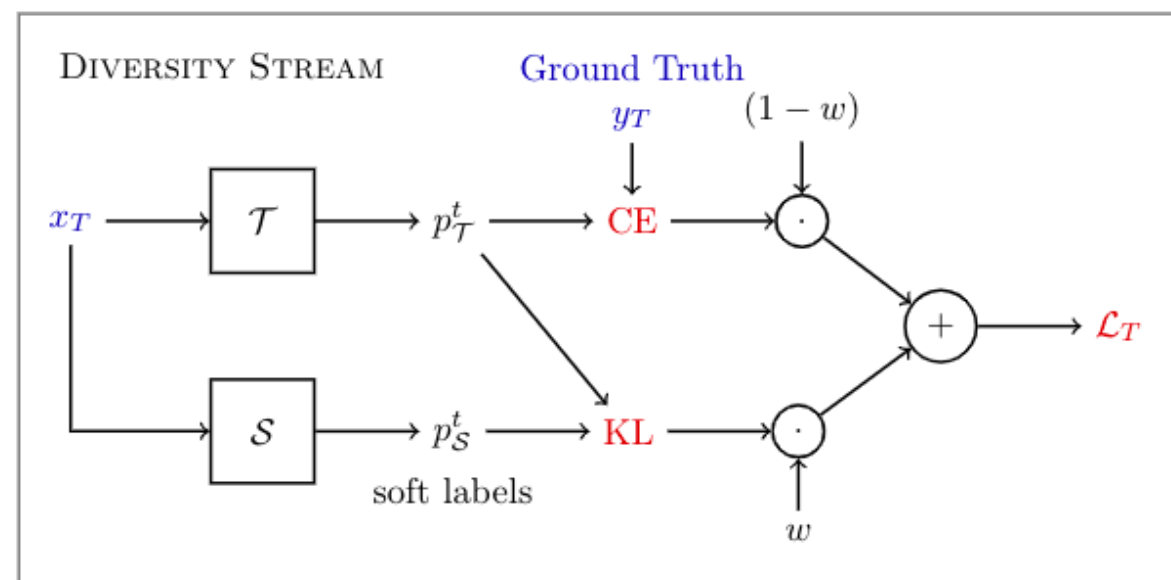
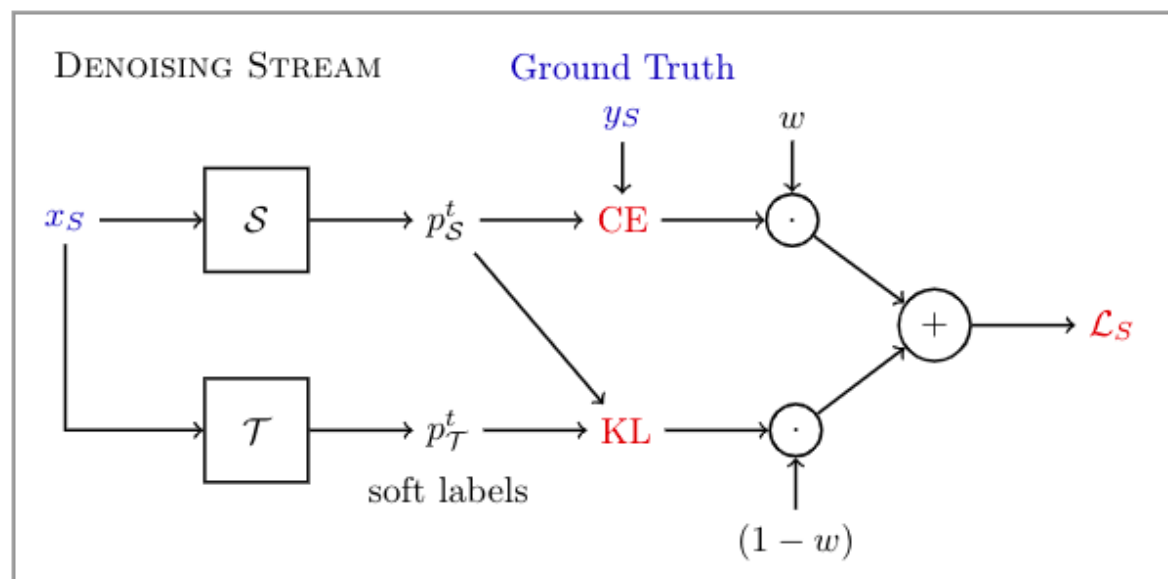


Image credit: <http://cocodataset.org/#home>

1. Sharma, P., Ding, N., Goodman, S., Soricut, R.: Conceptual captions: A cleaned hypernymed, image alt-text dataset for automatic image captioning. In: Proceedings of ACL (2018)
2. Lin, T., Maire, M., Belongie, S.J., Bourdev, L.D., Girshick, R.B., Hays, J., Perona, P., Ramanan, D., Dollár, P., Zitnick, C.L.: Microsoft COCO: common objects in context. EECV (2014)

CoDistill: Alternating between *Denoising* and *Diversity* streams

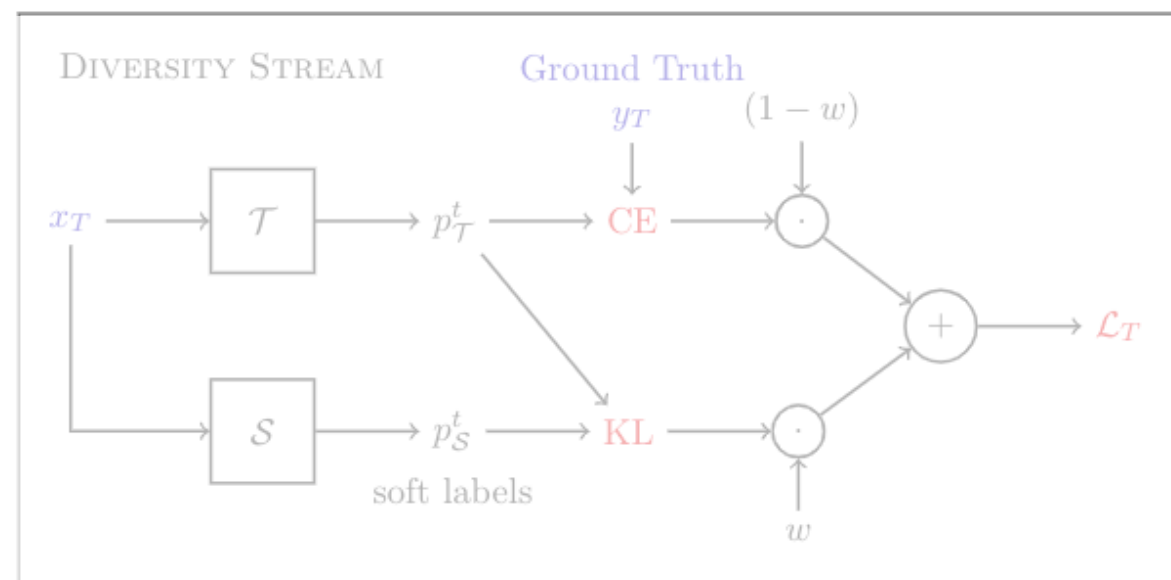
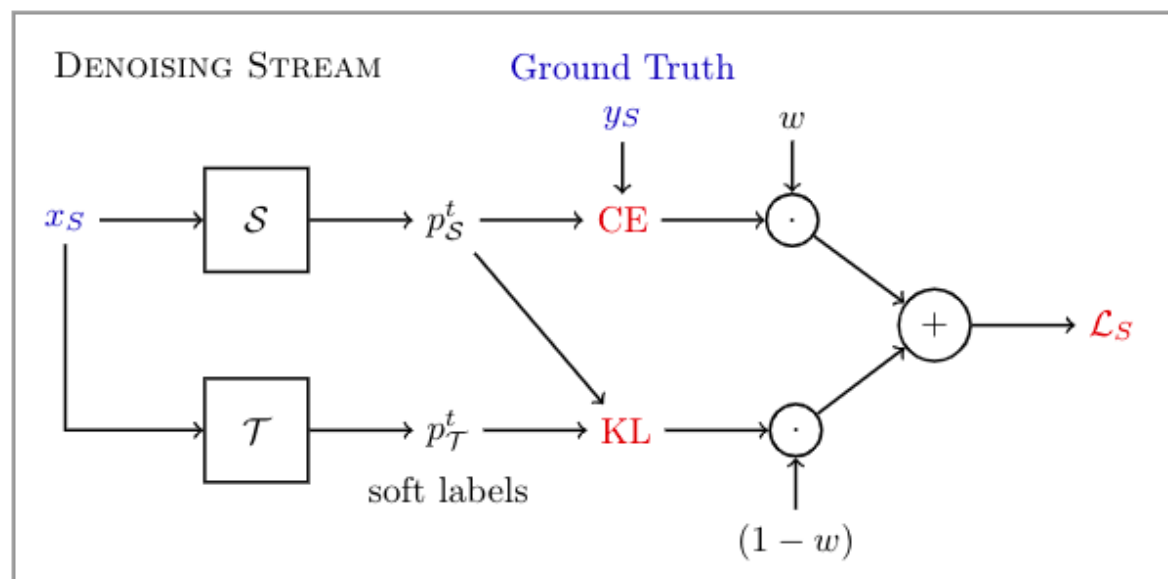
Student learns, teacher is fixed



Teacher learns, student is fixed

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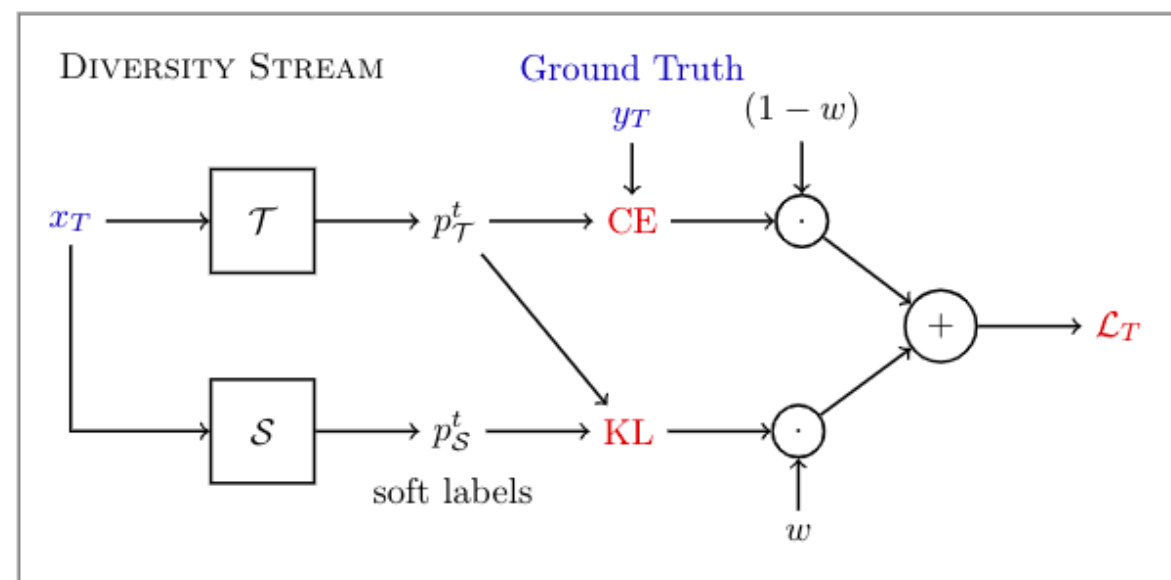
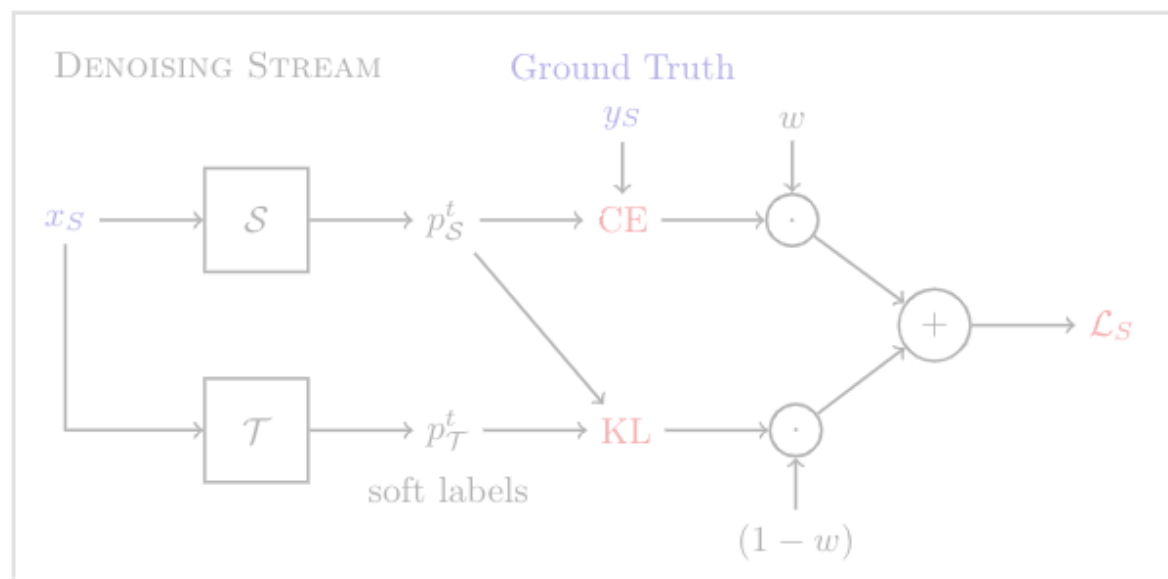
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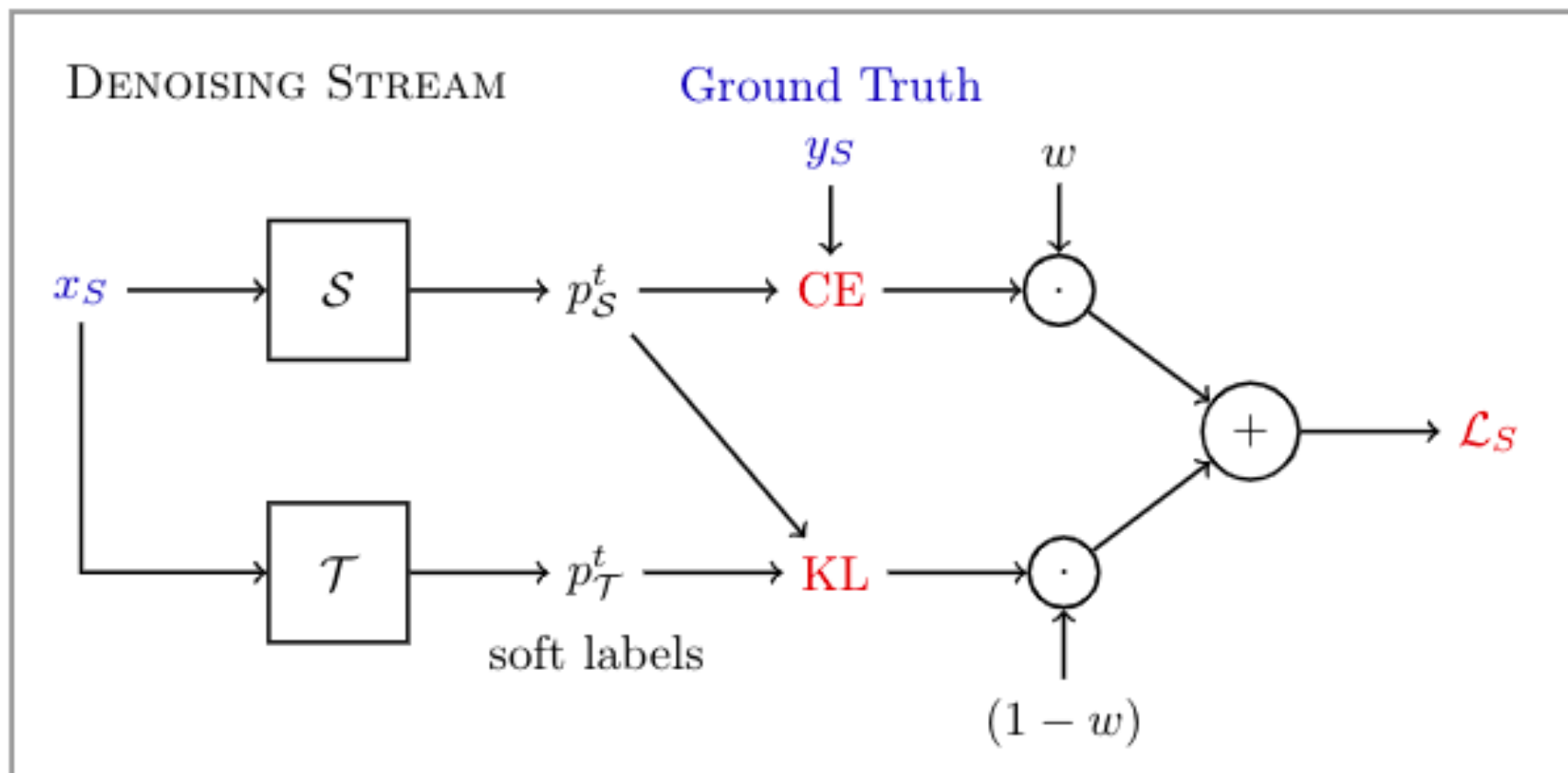
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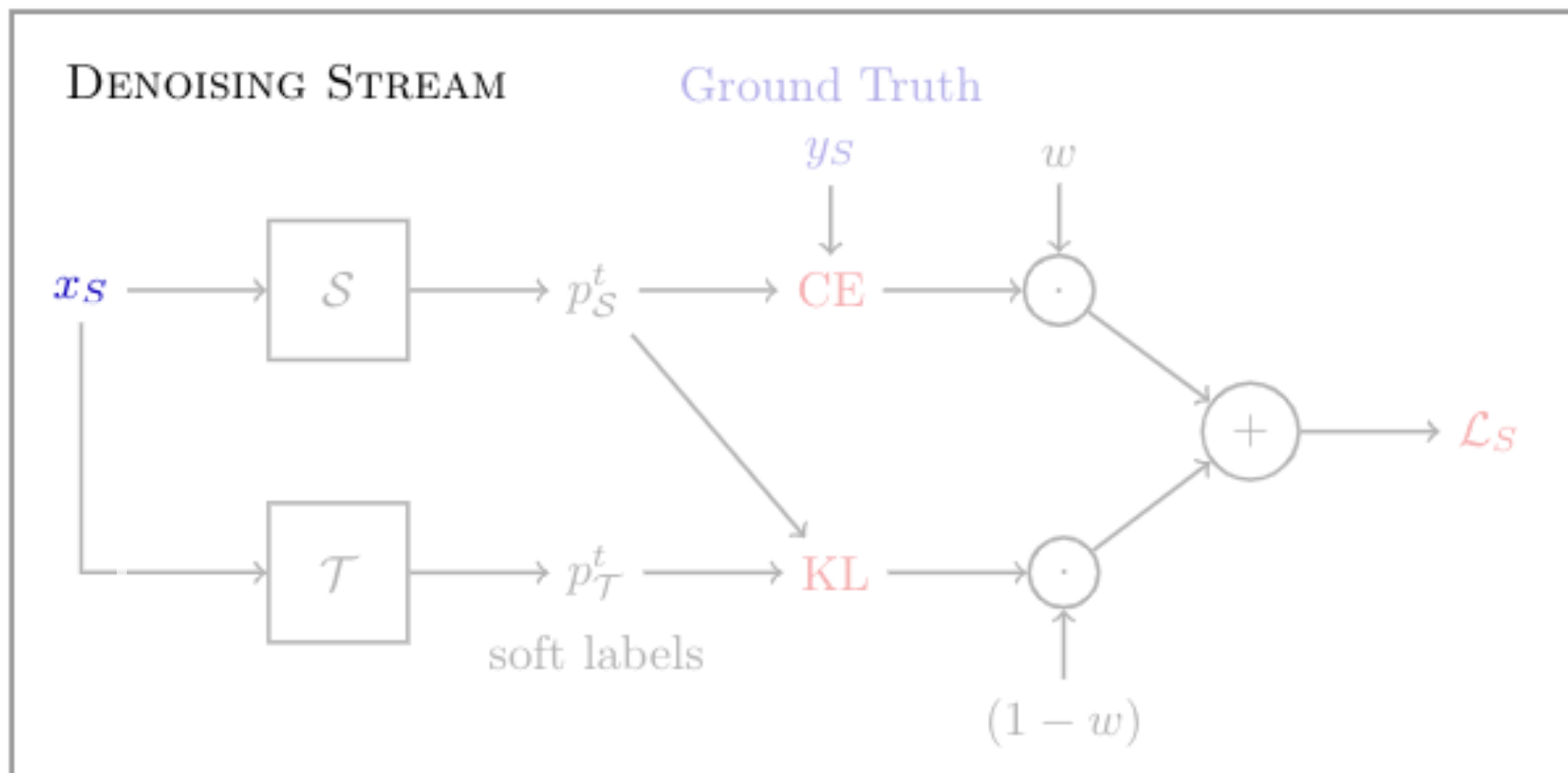
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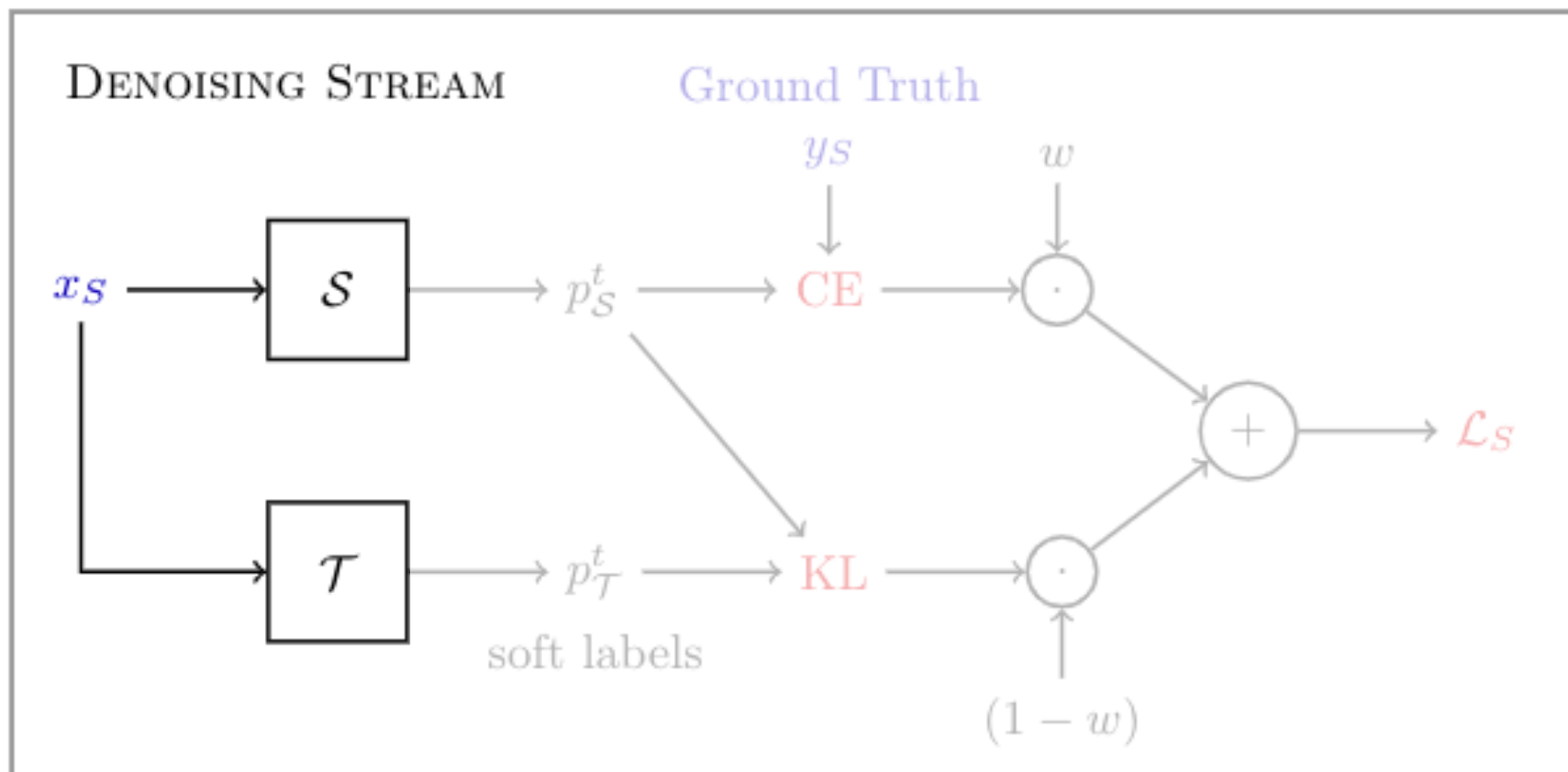
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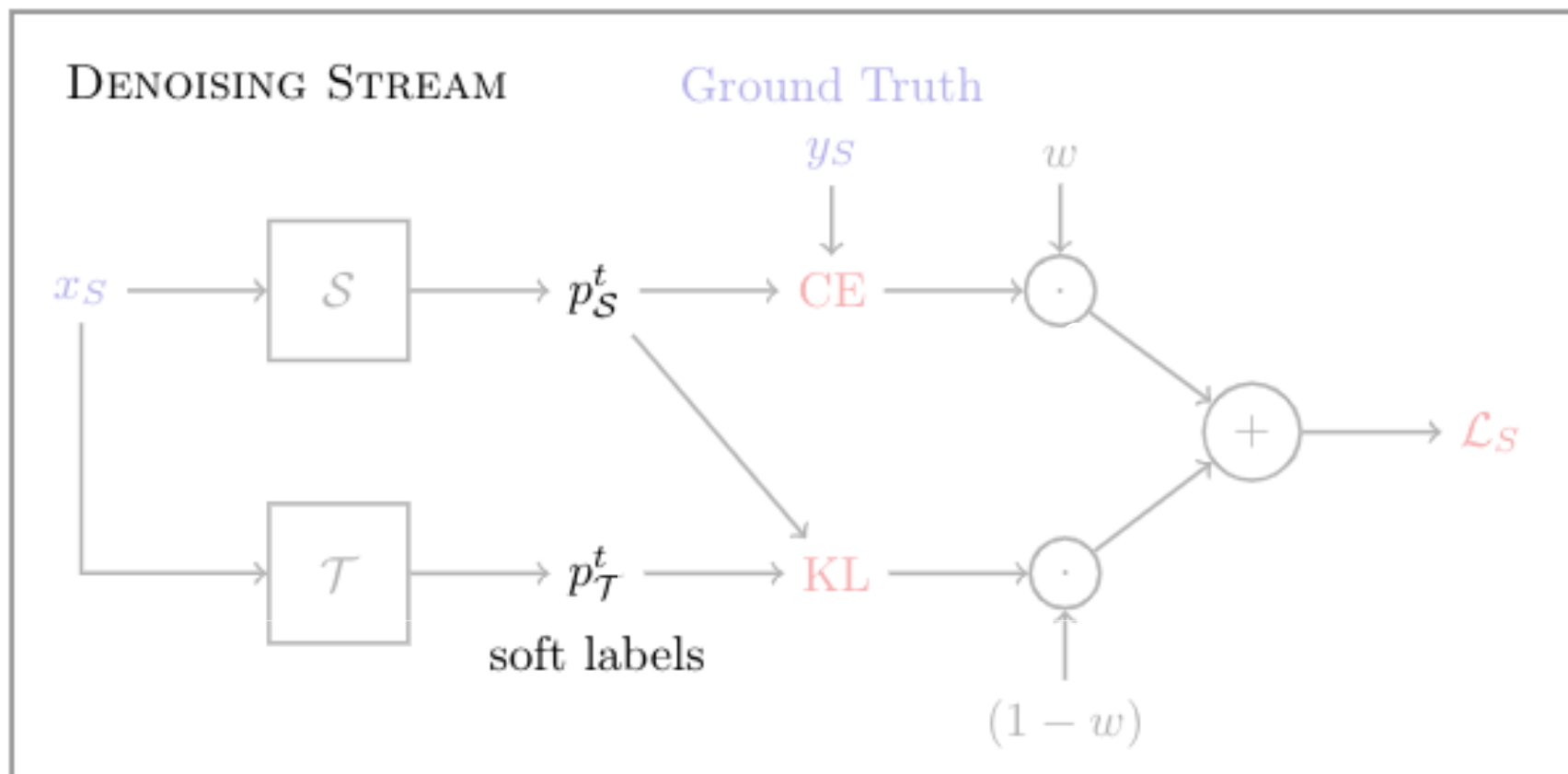


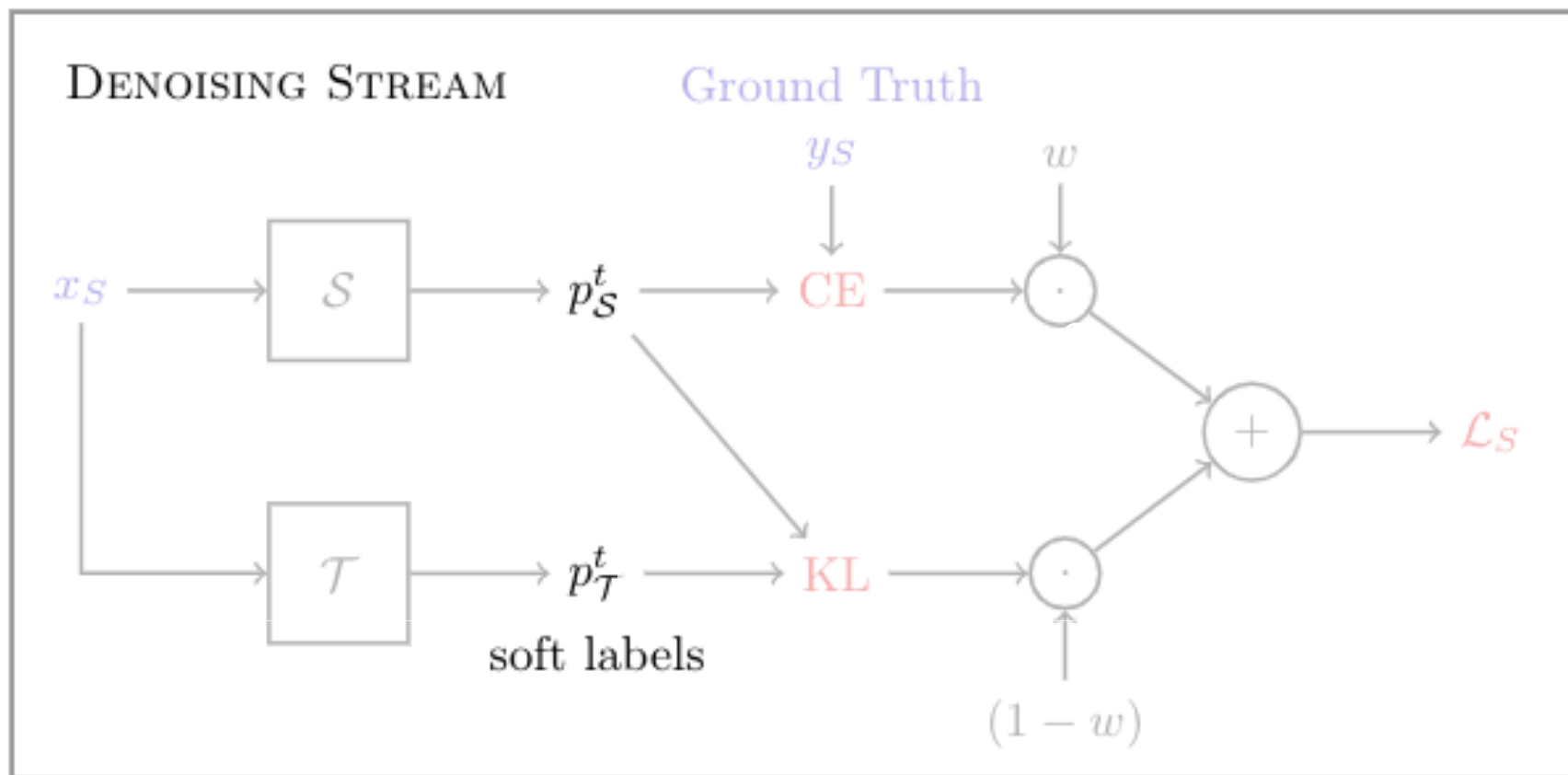
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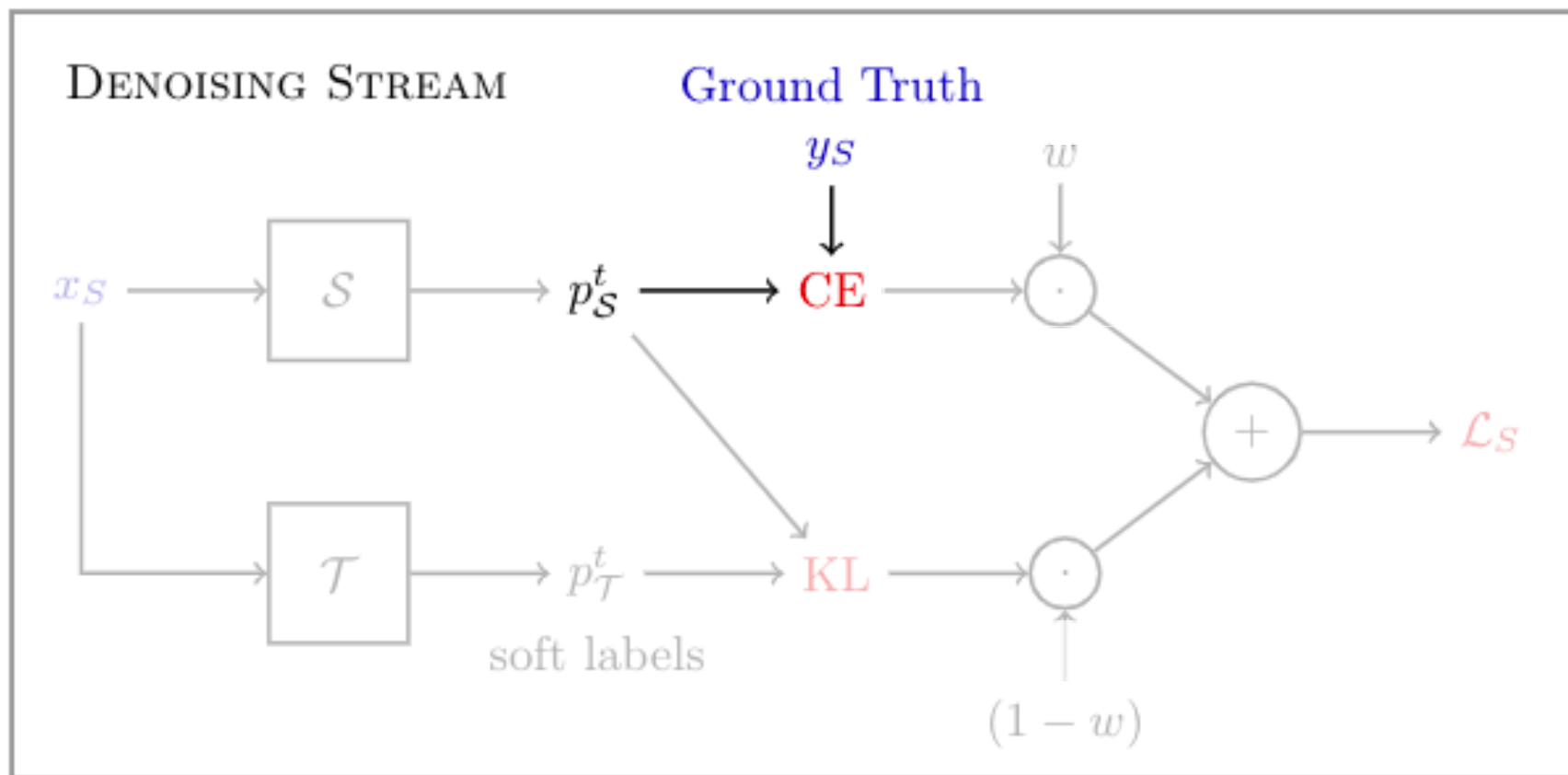




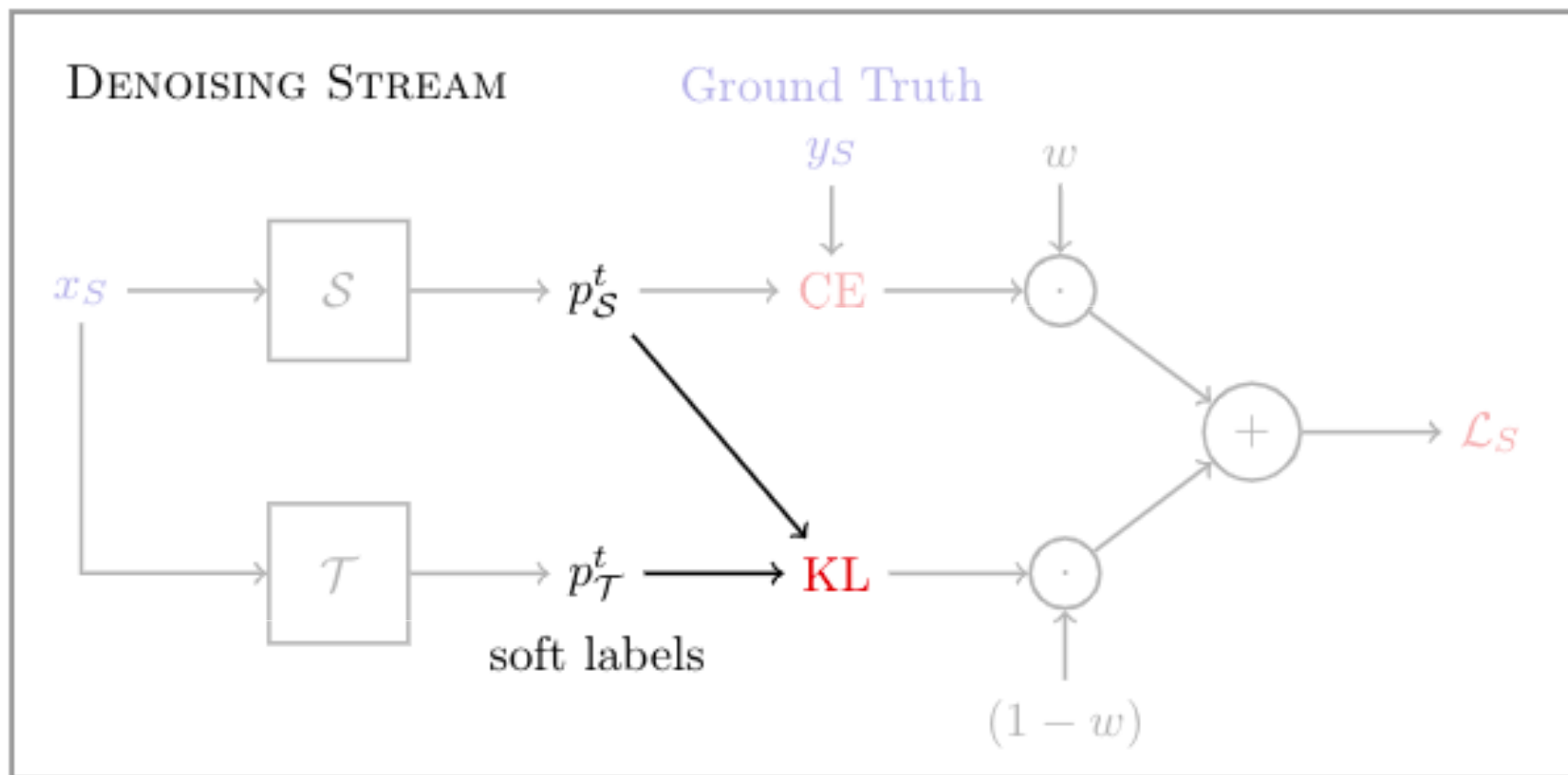




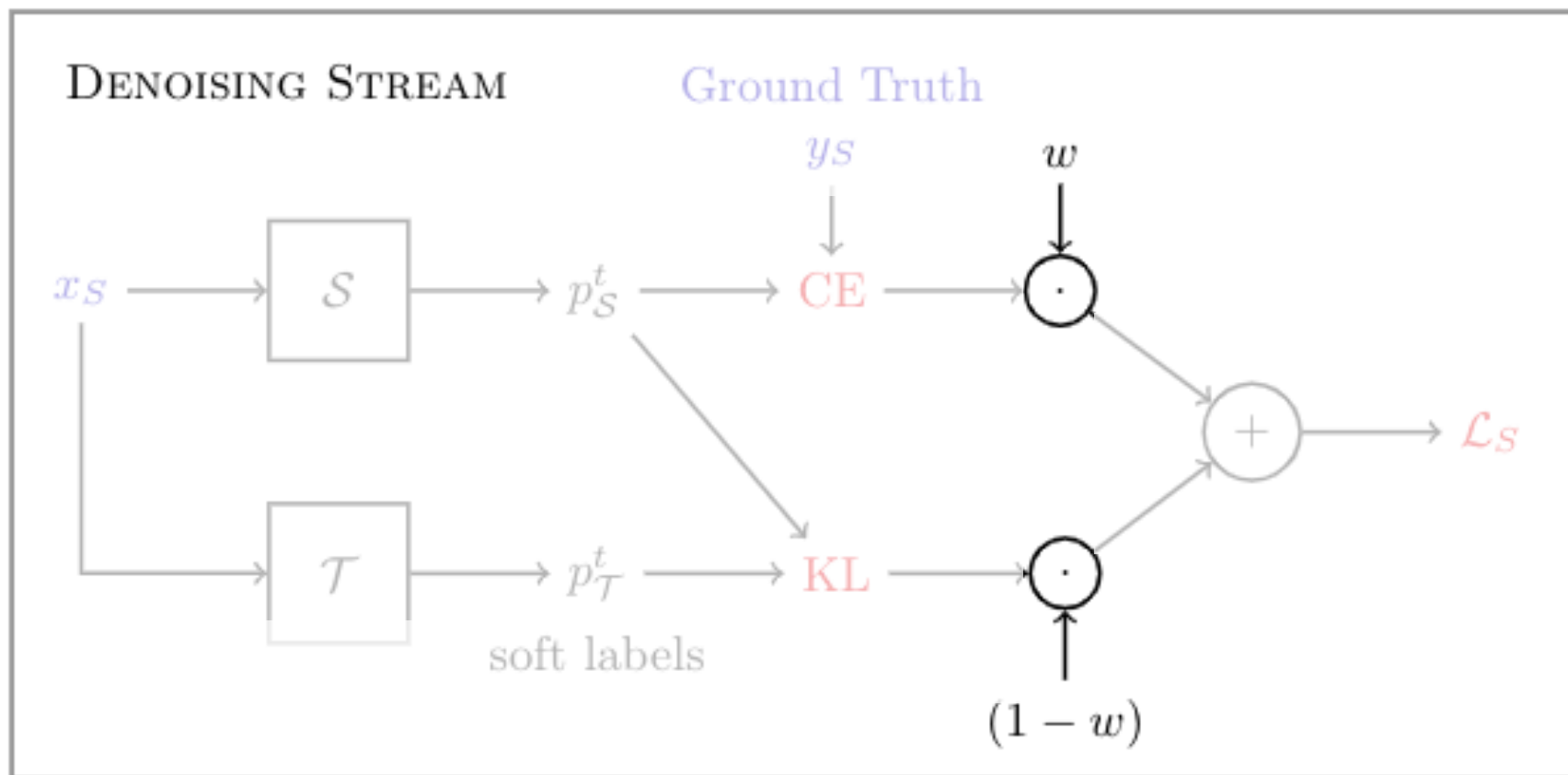
$$w(x_S, y_S) \left(- \sum_{t=1}^{L_S} \langle y_S^t, \log p_S^t(x_S) \rangle \right) + (1 - w(x_S, y_S)) \sum_{t=1}^{L_T} \text{KL}(p_T^t, p_S^t)$$



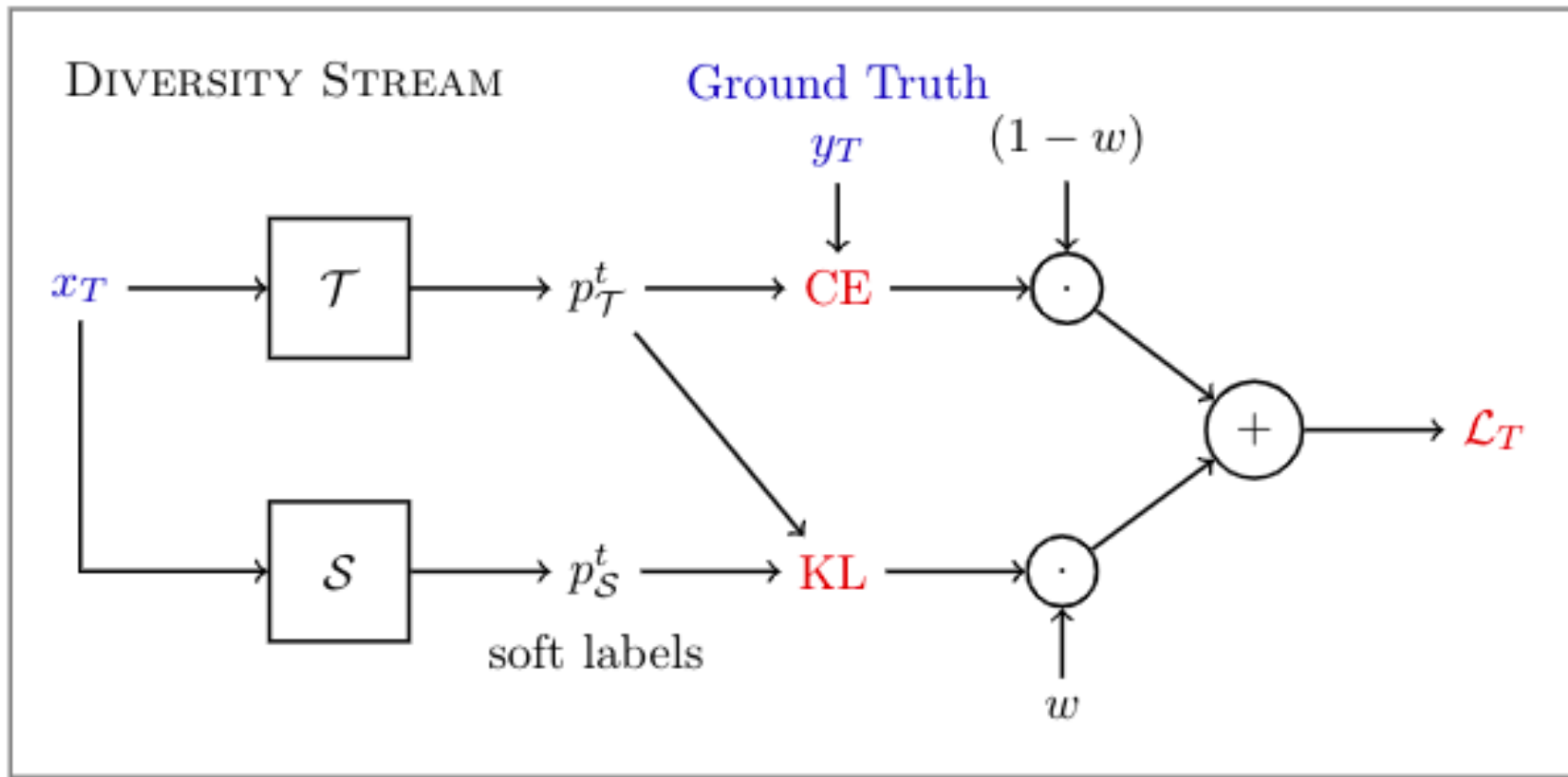
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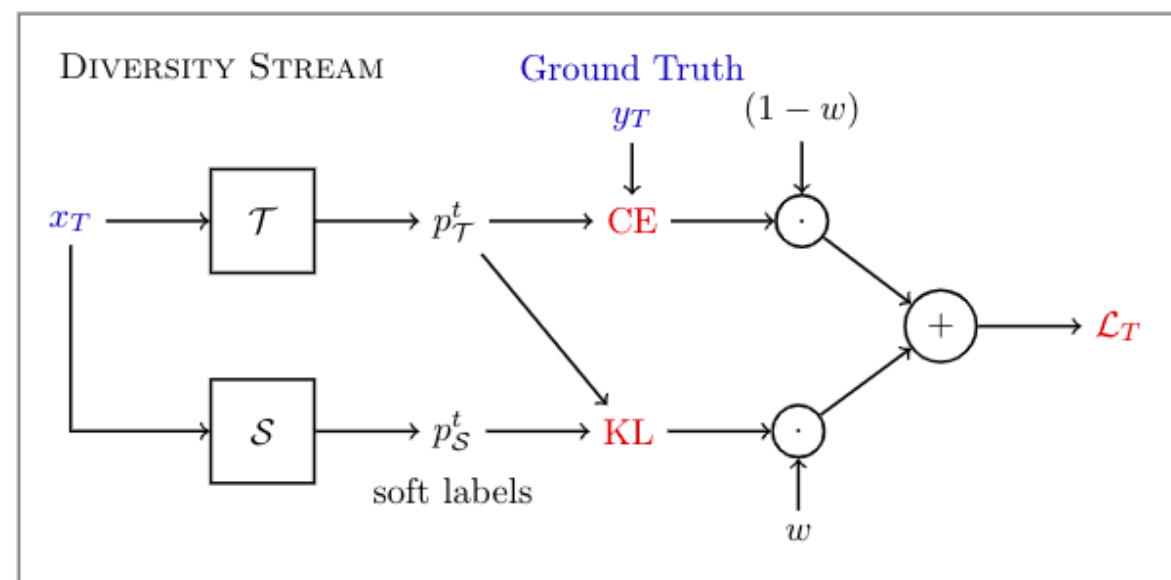
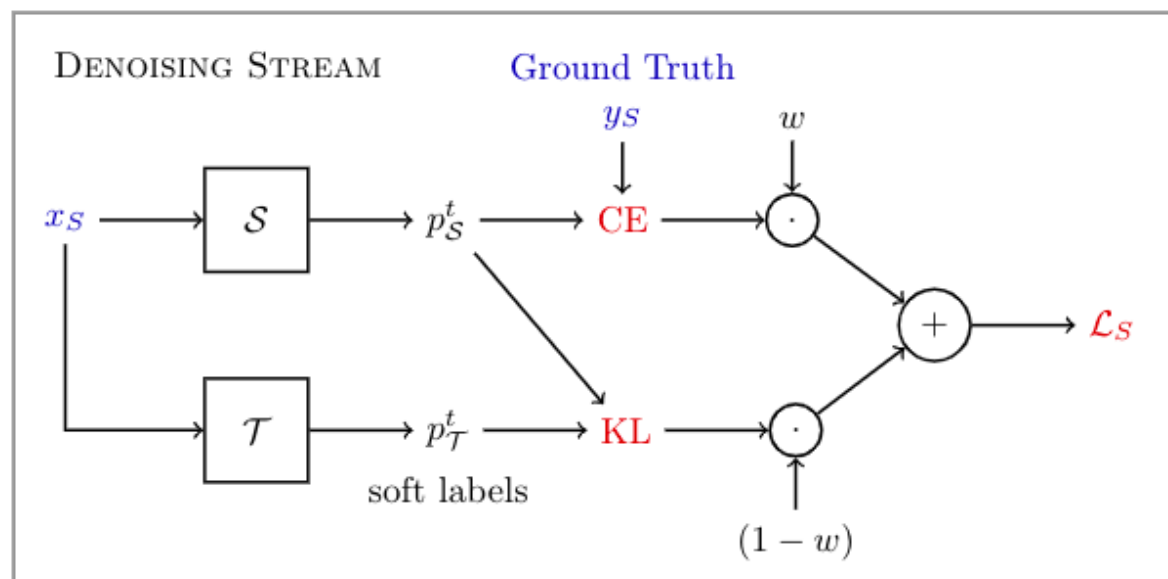
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$$(1 - w(x_T, y_T)) \left(- \sum_{t=1}^{L_T} \langle y_T^t, \log p_{\mathcal{T}}^t(x_T) \rangle \right) + w(x_T, y_T) \sum_{t=1}^{L_S} \text{KL}(p_{\mathcal{S}}^t, p_{\mathcal{T}}^t)$$

CoDistill: Alternating between *Denoising* and *Diversity* streams

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Denoising in action



GT: things to do before you move into a new house
S: a fire in a fireplace surrounded by logs



GT: the most stylish dog on the internet
S: a dog wearing a blue hat and glasses



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**Thank
you!**