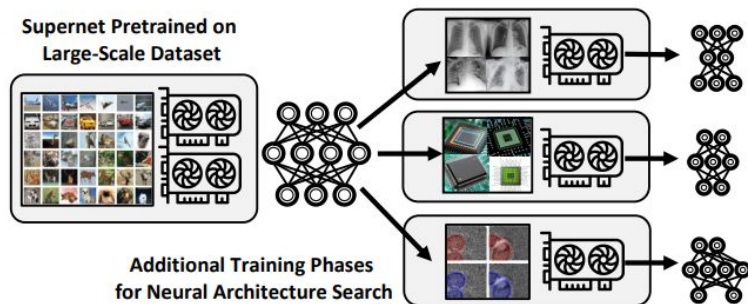


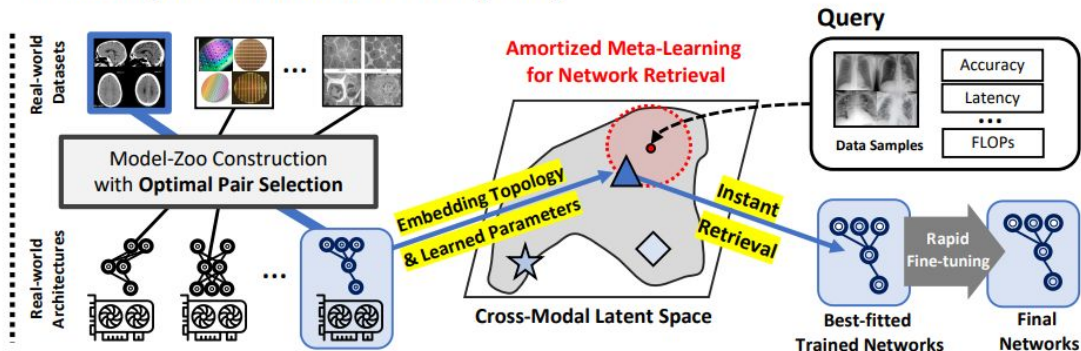
# Task-Adaptive Neural Network Search with Meta-Contrastive Learning

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## Conventional Neural Architecture Search



## Task-Adaptive Network Search (Ours)



**Figure 1: Comparison between conventional NAS and our method:** Conventional supernet-based NAS approaches (Left) sample subnets from a fixed supernet trained on a single dataset. TANS (Right) can dynamically select the best-fitted neural networks that are trained on diverse datasets, adaptively for each query dataset.

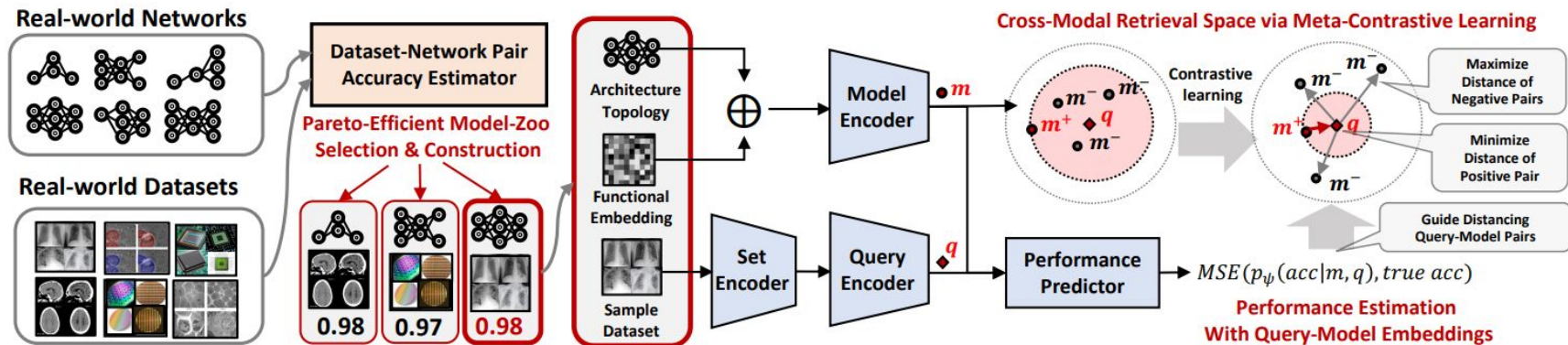


Figure 3: **Illustration for overall framework of our proposed method (TANS):** We first construct our model-zoo with pareto-optimal pairs of dataset and network, rather than exhaustively train all possible pairs. We then embed a model and a dataset with a graph-functional model and a set encoder. After that, we meta-learn the cross-modal retrieval network over multiple model-query pairs, guided by our performance predictor.

# Problem statement

$$\tau = \{D^\tau, M^\tau, s^\tau\}$$

$$\mathbf{q} = E_Q(D^\tau; \boldsymbol{\theta}) \quad \text{and} \quad \mathbf{m} = E_M(M^\tau; \boldsymbol{\phi}),$$

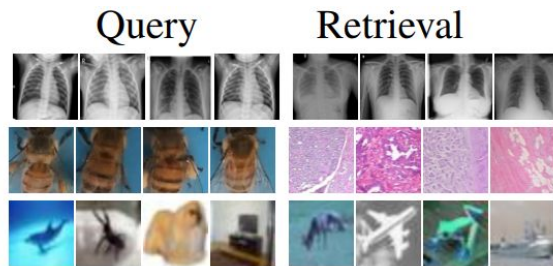
$$\mathcal{L}_m(\tau; \boldsymbol{\theta}, \boldsymbol{\phi}) = \max \left( 0, \alpha - \log \frac{\exp(f_{sim}(\mathbf{q}, \mathbf{m}^+))}{\exp \left( \sum_{\gamma \in p(\tau), \gamma \neq \tau} f_{sim}(\mathbf{q}, \mathbf{m}^-) \right)} \right)$$

$$\min_{\phi, \boldsymbol{\theta}, \boldsymbol{\psi}} \sum_{\tau \in p(\tau)} \mathcal{L}_m(\tau; \boldsymbol{\theta}, \phi) + \mathcal{L}_q(\tau; \boldsymbol{\theta}, \phi) + \lambda \cdot \mathcal{L}_s(\tau; \boldsymbol{\psi}),$$

$$\mathcal{L}_s(\tau; \boldsymbol{\psi}) = (s_{\text{acc}}^\tau - S(\tau; \boldsymbol{\psi}))^2$$

# Inference

$$M^* = \max_{M^\tau} \{f_{sim}(\tilde{\mathbf{q}}, \mathbf{m}^\tau) \mid \tau \in p(\tau)\},$$





# Results

Target Dataset	Method	# Epochs	FLOPs (M)	Params (M)	Search Time (GPU sec)	Training Time (GPU sec)	Speed Up	Accuracy (%)
Averaged Performance	MobileNetV3 [26]	50	132.94	4.00	-	257.78 $\pm$ 09.77	1.00 $\times$	94.20 $\pm$ 0.70
	PC-DARTS [65]	500	566.55	<b>3.54</b>	1100.37 $\pm$ 22.20	5721.13 $\pm$ 793.71	0.04 $\times$	79.22 $\pm$ 1.69
	DrNAS [10]	500	623.43	4.12	1501.75 $\pm$ 43.92	5659.77 $\pm$ 403.62	0.04 $\times$	84.06 $\pm$ 0.97
	FBNet-A [60]	50	246.69	4.3	-	293.42 $\pm$ 57.45	0.88 $\times$	93.00 $\pm$ 1.95
	OFA [8]	50	148.76	6.74	121.90 $\pm$ 0.00	226.58 $\pm$ 03.13	0.74 $\times$	93.89 $\pm$ 0.84
	MetaD2A [31]	50	512.67	6.56	2.59 $\pm$ 0.13	345.39 $\pm$ 28.36	0.74 $\times$	95.24 $\pm$ 1.14
	<b>TANS (Ours)</b>	10	181.74	5.51	0.002 $\pm$ 0.00	40.19 $\pm$ 03.06	-	95.17 $\pm$ 2.20
	<b>TANS (Ours)</b>	50	181.74	5.51	<b>0.002<math>\pm</math>0.00</b>	<b>200.93<math>\pm</math>11.01</b>	<b>1.28<math>\times</math></b>	<b>96.28<math>\pm</math>0.30</b>
Colorectal Histology Dataset (Easy)	MobileNetV3 [26]	50	132.94	<b>4.00</b>	-	577.18 $\pm$ 04.15	1.00 $\times$	96.23 $\pm$ 0.07
	PC-DARTS [65]	500	534.64	4.02	2062.42 $\pm$ 49.14	12124.18 $\pm$ 1051.16	0.04 $\times$	96.17 $\pm$ 0.68
	DrNAS [10]	500	614.23	4.12	4183.20 $\pm$ 188.60	11355.18 $\pm$ 1352.62	0.04 $\times$	97.51 $\pm$ 0.13
	FBNet-A [60]	50	215.45	4.3	-	696.00 $\pm$ 295.19	0.83 $\times$	95.43 $\pm$ 0.57
	OFA [8]	50	134.85	6.74	121.90 $\pm$ 0.00	537.61 $\pm$ 03.52	0.88 $\times$	96.40 $\pm$ 0.52
	MetaD2A [31]	50	506.88	5.93	2.58 $\pm$ 0.12	784.45 $\pm$ 79.32	0.73 $\times$	96.57 $\pm$ 0.56
	<b>TANS (Ours)</b>	10	171.74	4.95	0.001 $\pm$ 0.00	98.56 $\pm$ 04.24	-	96.87 $\pm$ 0.21
	<b>TANS (Ours)</b>	50	171.74	4.95	<b>0.001<math>\pm</math>0.00</b>	<b>492.81<math>\pm</math>21.19</b>	<b>1.17<math>\times</math></b>	<b>97.67<math>\pm</math>0.05</b>
Food Classification Dataset (Hard)	MobileNetV3 [26]	50	132.94	4.00	-	235.57 $\pm$ 07.57	1.00 $\times$	87.52 $\pm$ 0.78
	PC-DARTS [65]	500	567.85	<b>3.62</b>	1018.49 $\pm$ 6.31	6323.40 $\pm$ 938.83	0.03 $\times$	55.42 $\pm$ 2.46
	DrNAS [10]	500	632.67	4.12	1276.38 $\pm$ 0.00	5079.89 $\pm$ 161.05	0.04 $\times$	61.45 $\pm$ 0.68
	FBNet-A [60]	50	251.29	4.3	-	251.24 $\pm$ 3.31	0.94 $\times$	84.33 $\pm$ 1.41
	OFA [8]	50	152.34	6.74	121.90 $\pm$ 0.00	<b>190.86<math>\pm</math>03.48</b>	0.75 $\times$	87.43 $\pm$ 0.59
	MetaD2A [31]	50	521.11	8.23	2.60 $\pm$ 0.23	324.62 $\pm$ 34.97	0.72 $\times$	89.72 $\pm$ 1.53
	<b>TANS (Ours)</b>	10	179.83	5.07	0.002 $\pm$ 0.00	40.59 $\pm$ 04.84	-	93.11 $\pm$ 0.24
	<b>TANS (Ours)</b>	50	179.83	5.07	<b>0.002<math>\pm</math>0.00</b>	202.93 $\pm$ 24.21	<b>1.16<math>\times</math></b>	<b>93.71<math>\pm</math>0.24</b>