



Generalized Category Discovery with Decoupled Prototypical Network

Wenbin An^{1,3}, Feng Tian^{2,3}, Qinghua Zheng^{2,3}, Wei Ding⁴, Qian Ying Wang⁵, Ping Chen⁶

¹School of Automation Science and Engineering, Xi'an Jiaotong University

²School of Computer Science and Technology, Xi'an Jiaotong University

³National Engineering Laboratory for Big Data Analytics

⁴Department of Computer Science, University of Massachusetts Boston

⁵Lenovo Research

⁶Department of Engineering, University of Massachusetts Boston
wenbinan@stu.xjtu.edu.cn, {fengtian,qhzheng}@mail.xjtu.edu.cn
{wei.ding,Ping.Chen}@umb.edu, wangqya@lenovo.com

Code: <https://github.com/Lackel/DPN>

2023. 2. 16 • ChongQing

— AAAI 2023



gesis
Leibniz-Institut
für Sozialwissenschaften



Reported by Renhui Luo

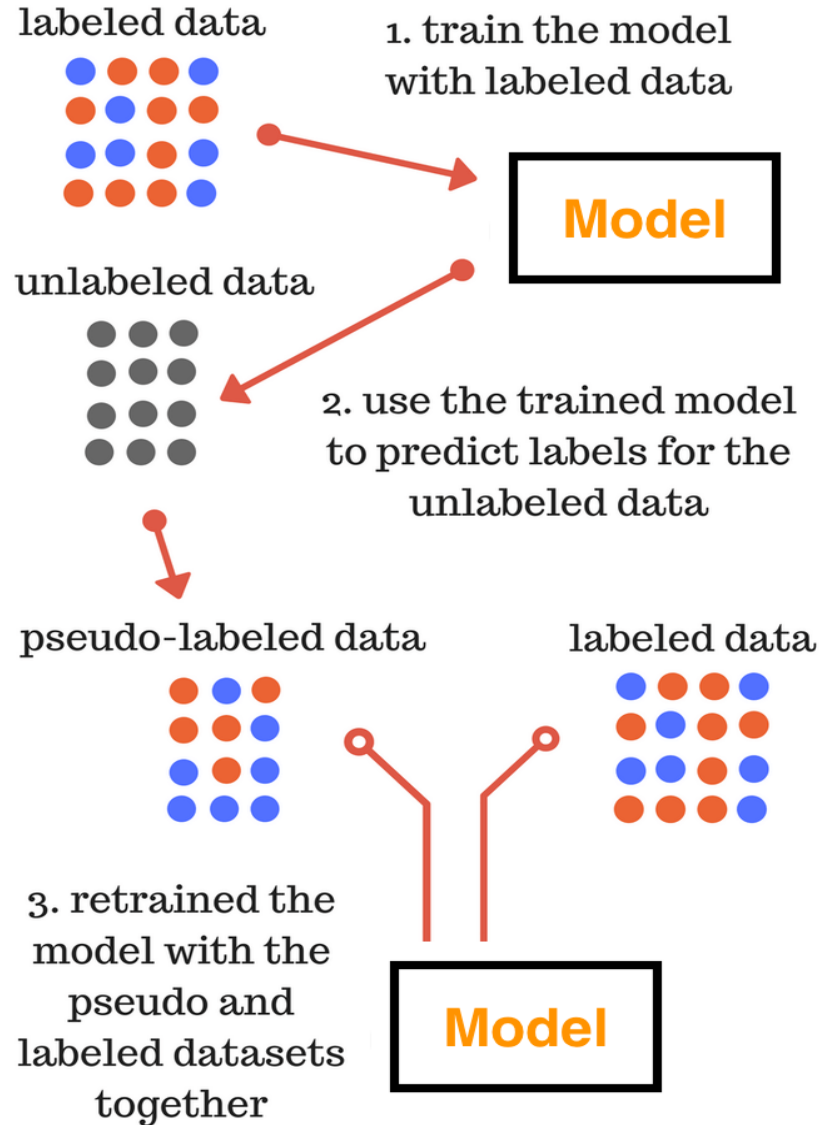


- 1.Introduction**
- 2.Overview**
- 3.Methods**
- 4.Experiments**





Introduction



Overview

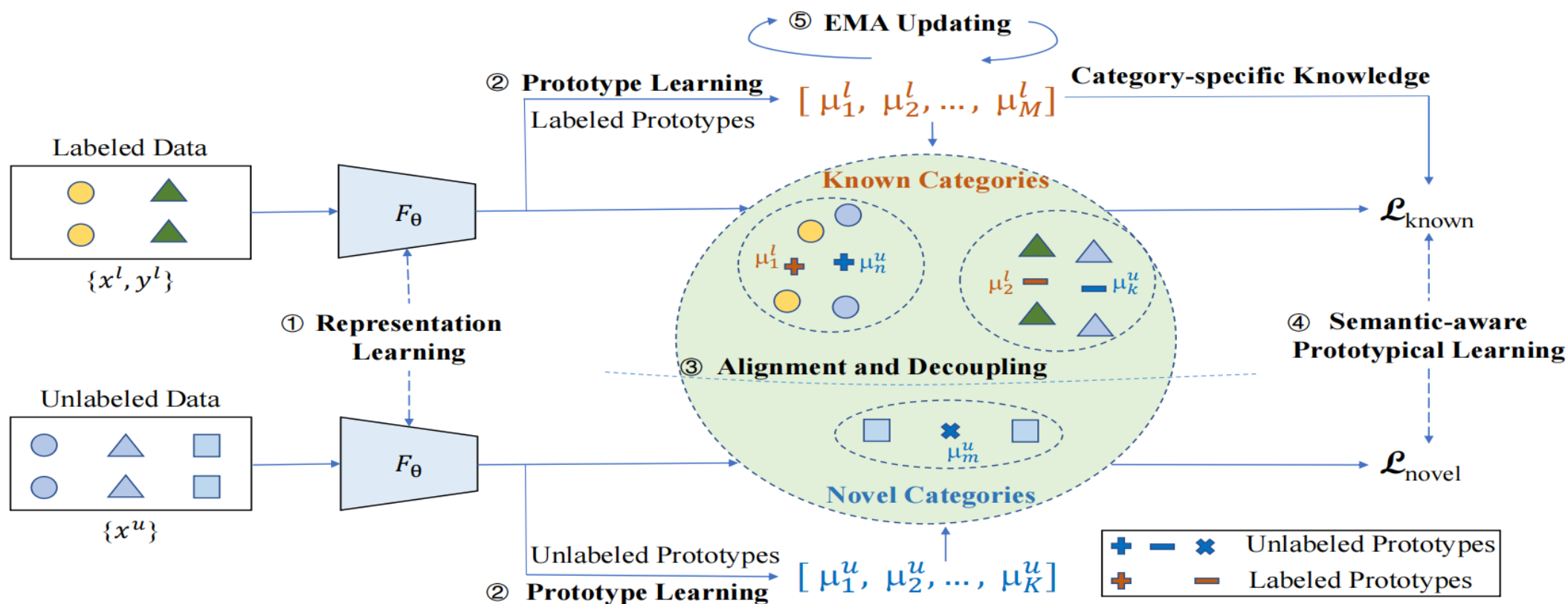
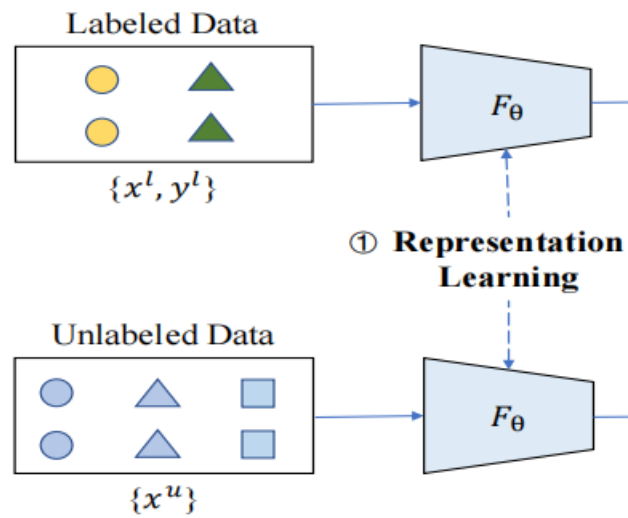


Figure 1: An overview of our model.

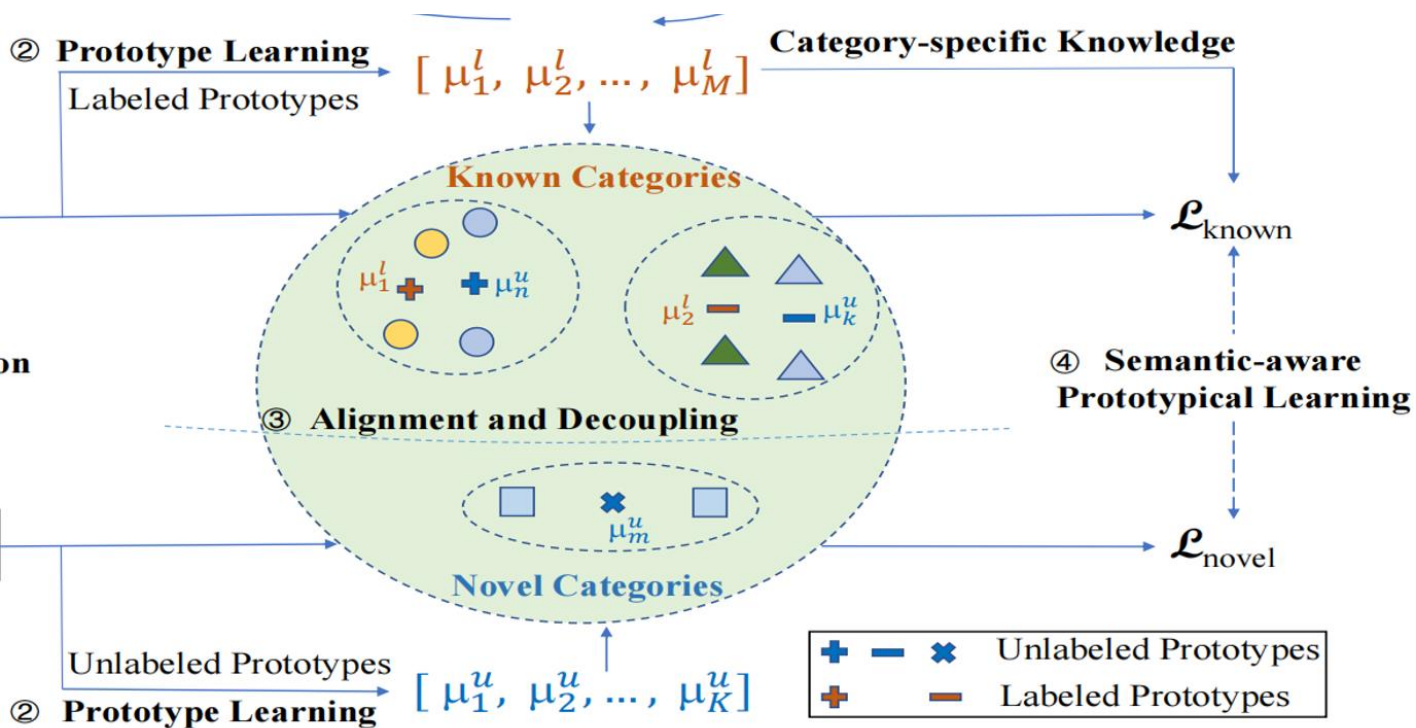


Method



$$\mathcal{L}_{pre} = \mathcal{L}_{ce}(\mathcal{D}^l) + \mathcal{L}_{mlm}(\mathcal{D}^l, \mathcal{D}^u) \quad (1)$$

Method



$$\hat{\mathcal{P}} = \arg \min_{\mathcal{P} \in \mathcal{P}_{\text{all}}} \sum_{i=1}^M \mathcal{L}_{\text{match}}(\mu_i^l, \mu_{\mathcal{P}(i)}^u) \quad (2)$$

$$\mathcal{L}_{\text{match}}(\mu_i^l, \mu_{\mathcal{P}(i)}^u) = \|\mu_i^l - \mu_{\mathcal{P}(i)}^u\|_2 \quad (3)$$

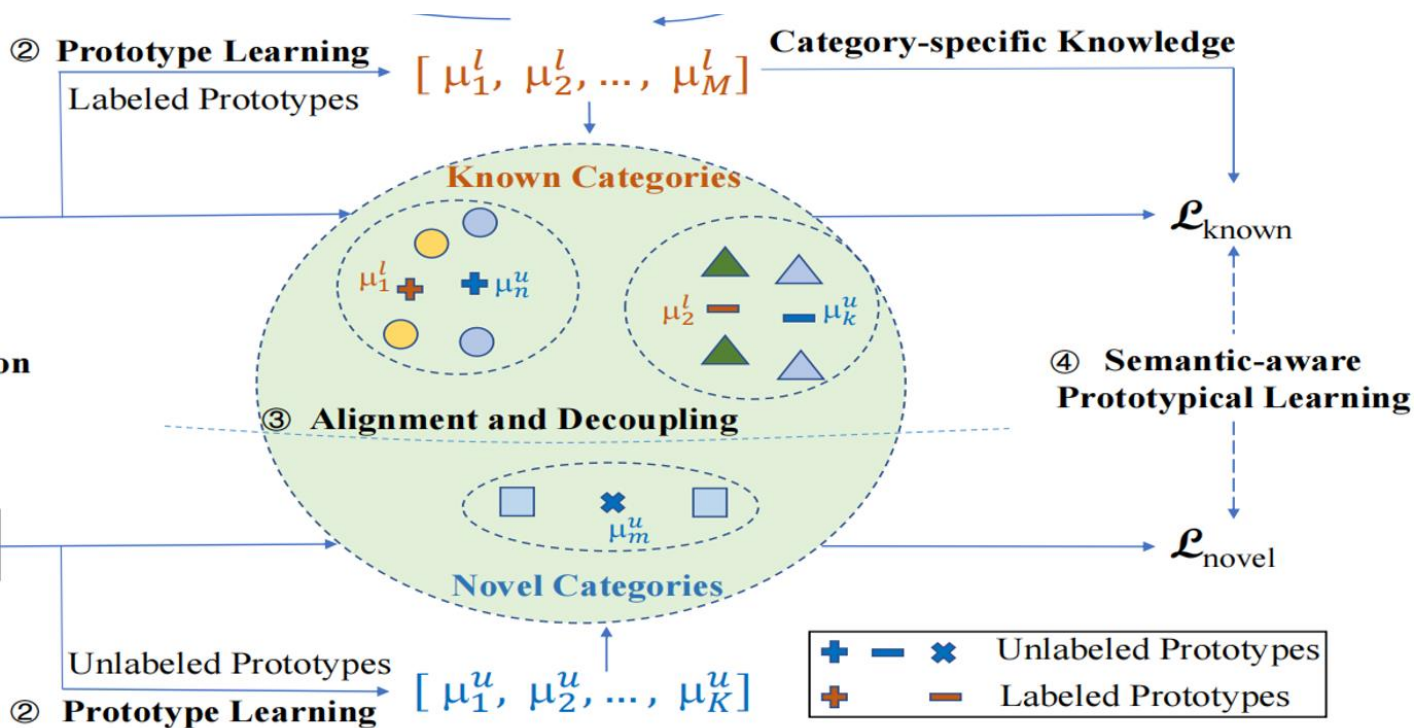
$$\mathcal{L}_{\text{pl}} = -\frac{1}{n} \sum_{i=1}^n \log \frac{e^{-d(F_{\theta}(x_i), \mu_g^u)}}{\sum_{j=1}^K e^{-d(F_{\theta}(x_i), \mu_j^u)}} \quad (4)$$

$$\mathcal{L}_{\text{spl}} = \frac{1}{n} \sum_{i=1}^n \sum_{k=1}^K d(F_{\theta}(x_i), \mu_k^u) \frac{e^{s(F_{\theta}(x_i), \mu_k^u)}}{\sum_{j=1}^K e^{s(F_{\theta}(x_i), \mu_j^u)}} \quad (5)$$

$$\mathcal{L}_{\text{spl}} = \frac{1}{n} \sum_{i=1}^n \sum_{k=1}^K \|F_{\theta}(x_i) - \mu_k^u\|_2 \frac{e^{\cos(F_{\theta}(x_i), \mu_k^u)/\tau}}{\sum_{j=1}^K e^{\cos(F_{\theta}(x_i), \mu_j^u)/\tau}} \quad (6)$$

$$\mathcal{L}_{\text{novel}} = \mathcal{L}_{\text{spl}}(\mathcal{D}^{\text{un}}, \mathcal{P}^{\text{un}}) \quad (7)$$

Method

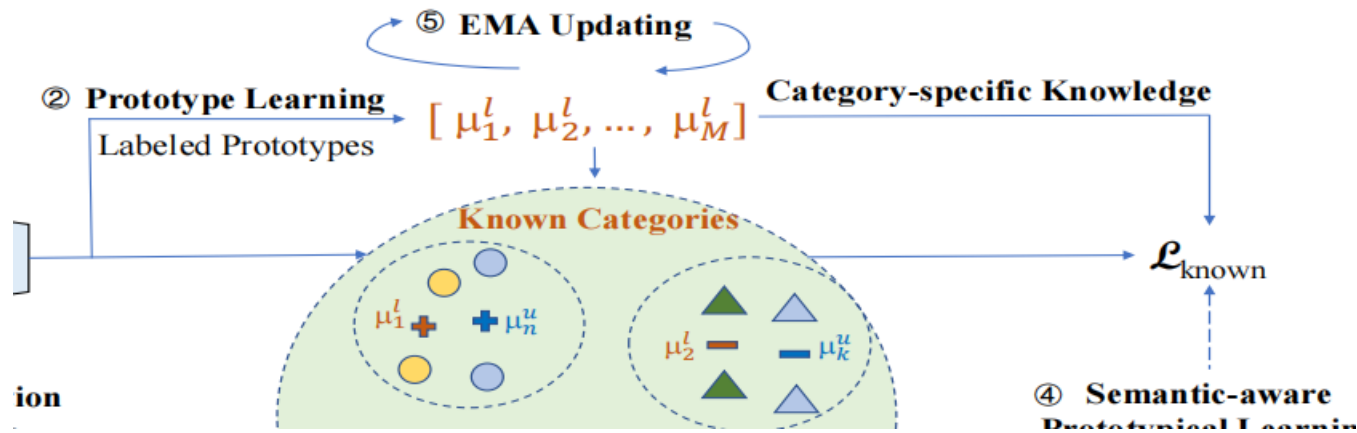


$$\mathcal{L}_{reg} = \frac{1}{r} \sum_{i=1}^r \sum_{k=1}^M (1 - \cos(F_{\theta}(x_i), \mu_k^l)) \frac{e^{\cos(F_{\theta}(x_i), \mu_k^l)/\tau}}{\sum_{j=1}^M e^{\cos(F_{\theta}(x_i), \mu_j^l)/\tau}} \quad (8)$$

$$\mathcal{L}_{known} = \mathcal{L}_{spl}(\mathcal{D}^{uk}, P^{uk}) + \mathcal{L}_{ce}(\mathcal{D}^l) + \lambda \cdot \mathcal{L}_{reg}(\mathcal{D}^{uk}, P^l) \quad (9)$$

$$\mathcal{L}_{dpn} = \mathcal{L}_{novel} + \mathcal{L}_{known} \quad (10)$$

Method



$$\mathcal{P}_{t+1}^l \leftarrow \alpha \cdot \mathcal{P}_t^l + (1 - \alpha) \cdot \mathcal{P}_{t+1}^l \quad (11)$$



Experiments

Table 1: Statistics of datasets. $|\mathcal{Y}_k|$, $|\mathcal{Y}_n|$, $|\mathcal{D}^l|$, $|\mathcal{D}^u|$ and $|\mathcal{D}^t|$ represent the number of known categories, novel categories, labeled data, unlabeled data and testing data, respectively.

Dataset	$ \mathcal{Y}_k $	$ \mathcal{Y}_n $	$ \mathcal{D}^l $	$ \mathcal{D}^u $	$ \mathcal{D}^t $
BANKING	58	19	673	8,330	3,080
StackOverflow	15	5	1,350	16,650	1,000
CLINC	113	37	1,344	16,656	2,250

Experiments

Table 2: Model comparison results (%) on testing sets. Average results over 3 runs are reported.

Method	BANKING			StackOverflow			CLINC		
	All	Known	Novel	All	Known	Novel	All	Known	Novel
DeepCluster	13.95	13.94	13.99	17.37	18.22	14.80	26.92	27.34	25.67
DCN	17.85	18.94	14.35	29.10	28.94	29.51	29.64	30.00	28.45
DEC	19.30	20.36	15.84	19.30	20.36	15.84	19.99	20.18	19.40
BERT	21.29	21.48	20.70	16.80	16.67	17.20	34.52	34.98	33.16
KM-GloVe	29.18	29.11	29.39	28.40	28.60	28.05	51.64	51.74	51.50
AG-GloVe	30.09	29.69	31.29	29.23	28.49	31.56	44.70	45.17	43.20
SAE	38.05	38.29	37.27	60.33	57.36	69.02	46.59	47.35	44.24
Semi-DC	50.73	53.37	42.63	64.90	66.13	61.20	74.52	75.60	71.34
CDAC+	53.09	55.42	46.01	76.67	77.51	74.13	69.75	70.08	68.77
Self-Labeling	56.19	61.64	39.56	71.03	78.53	48.53	72.69	80.06	49.65
DTC	56.56	59.98	46.10	70.50	80.93	51.87	76.42	82.34	58.95
DAC	63.63	69.60	45.44	70.77	76.13	54.67	84.42	89.10	70.59
Semi-KM	66.23	73.62	43.68	73.13	81.02	49.47	81.42	89.03	59.01
LASKM	67.55	75.16	44.34	74.83	82.00	53.33	79.26	89.64	48.66
DPN (Ours)	73.61	80.04	53.99	83.20	84.18	80.27	87.62	91.29	76.79
Improvement	+6.06	+4.88	+7.89	+6.53	+2.18	+6.14	+3.20	+1.65	+5.45



Experiments

Table 3: Results (%) of different model variants.

Model	All	Known	Novel
Ours	83.20	84.18	80.27
w/o Cross Entropy	82.53	83.47	79.73
w/o EMA	82.00	81.87	74.00
w/o Decoupling	81.10	82.93	75.60
w/o Soft Assignment	80.80	83.87	71.60
w/o Semantic Weights	70.50	72.53	64.40

Experiments

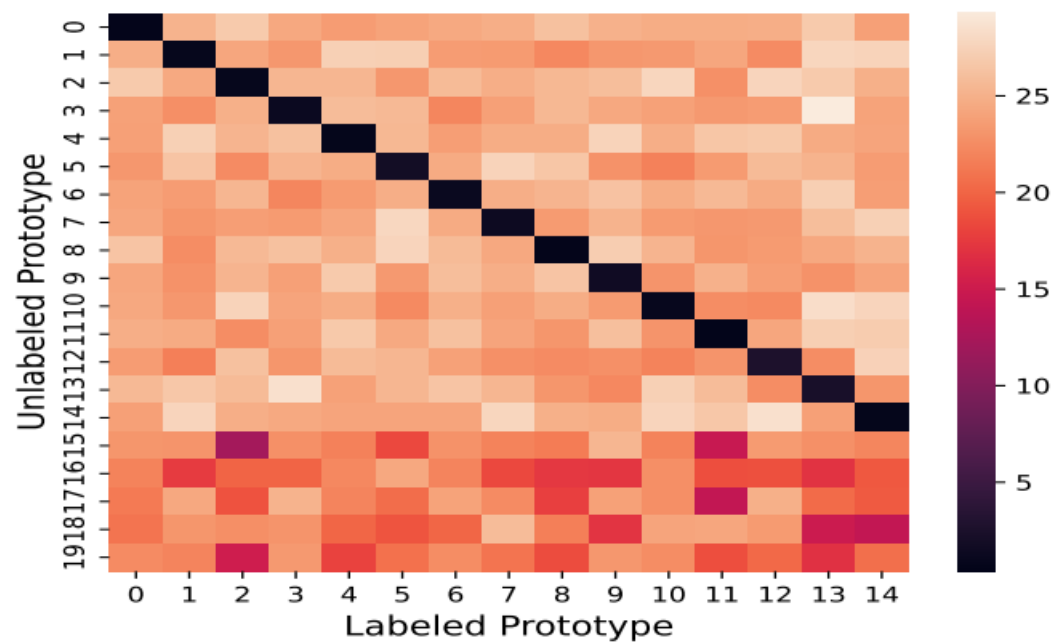


Figure 2: Distances between 15 labeled prototypes and 20 aligned unlabeled prototypes. Darker colors represent closer distances.

Experiments

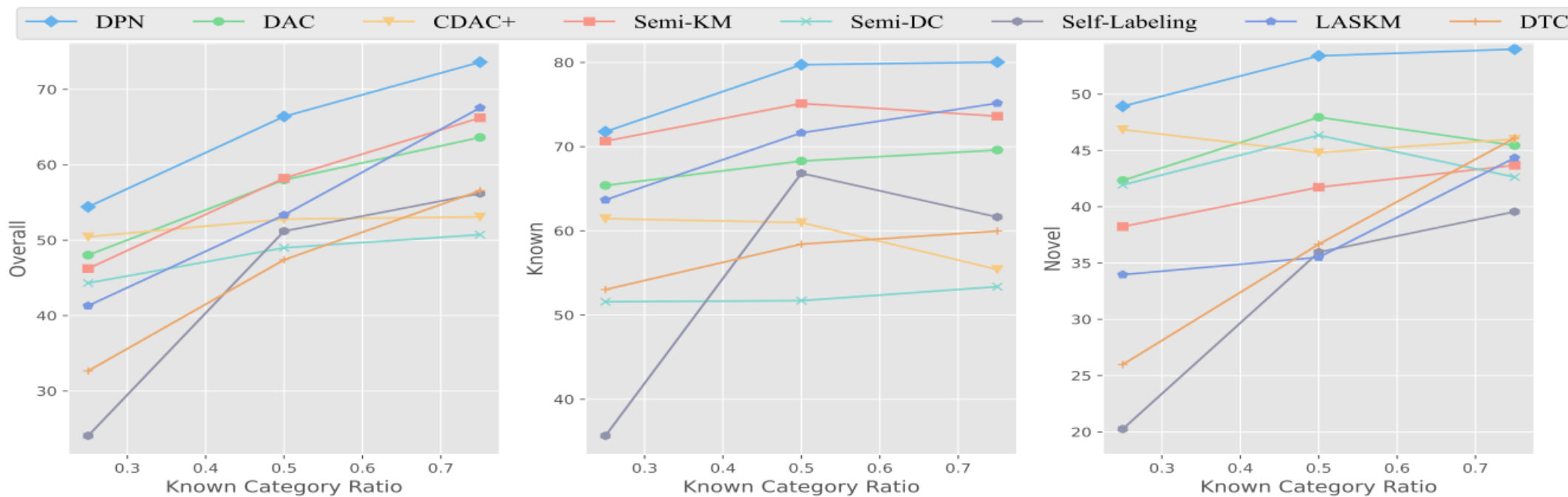


Figure 3: Effect of known category ratio on the BANKING dataset.

Experiments

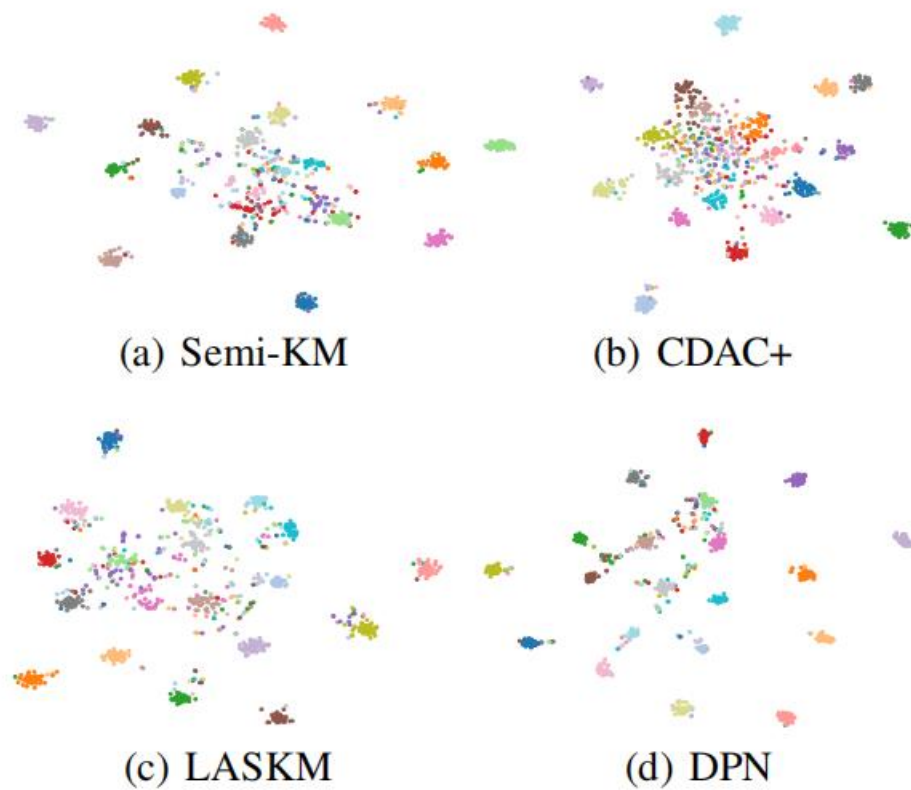


Figure 4: The t-SNE visualizations of embeddings.



Experiments

Table 4: Estimation of the number of categories.

	CLINC	BANKING	StackOverflow
Ground Truth	150	77	20
DAC	130	66	15
Ours	137	67	18
Error	8.7%	13.0%	10.0%



Thanks!