

Ensembles with shared representations

achieve human-level performance on facial expression recognition with a **low computational cost**.



Efficient Facial Feature Learning with Wide Ensemble-based Convolutional Neural Networks

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Ensemble Methods:

- Have proven to be efficient for reducing remaining residual generalization error.
- Are suitable for real-world applications due to their robustness and accuracy.

Problems with Ensembles of Deep Neural Networks:

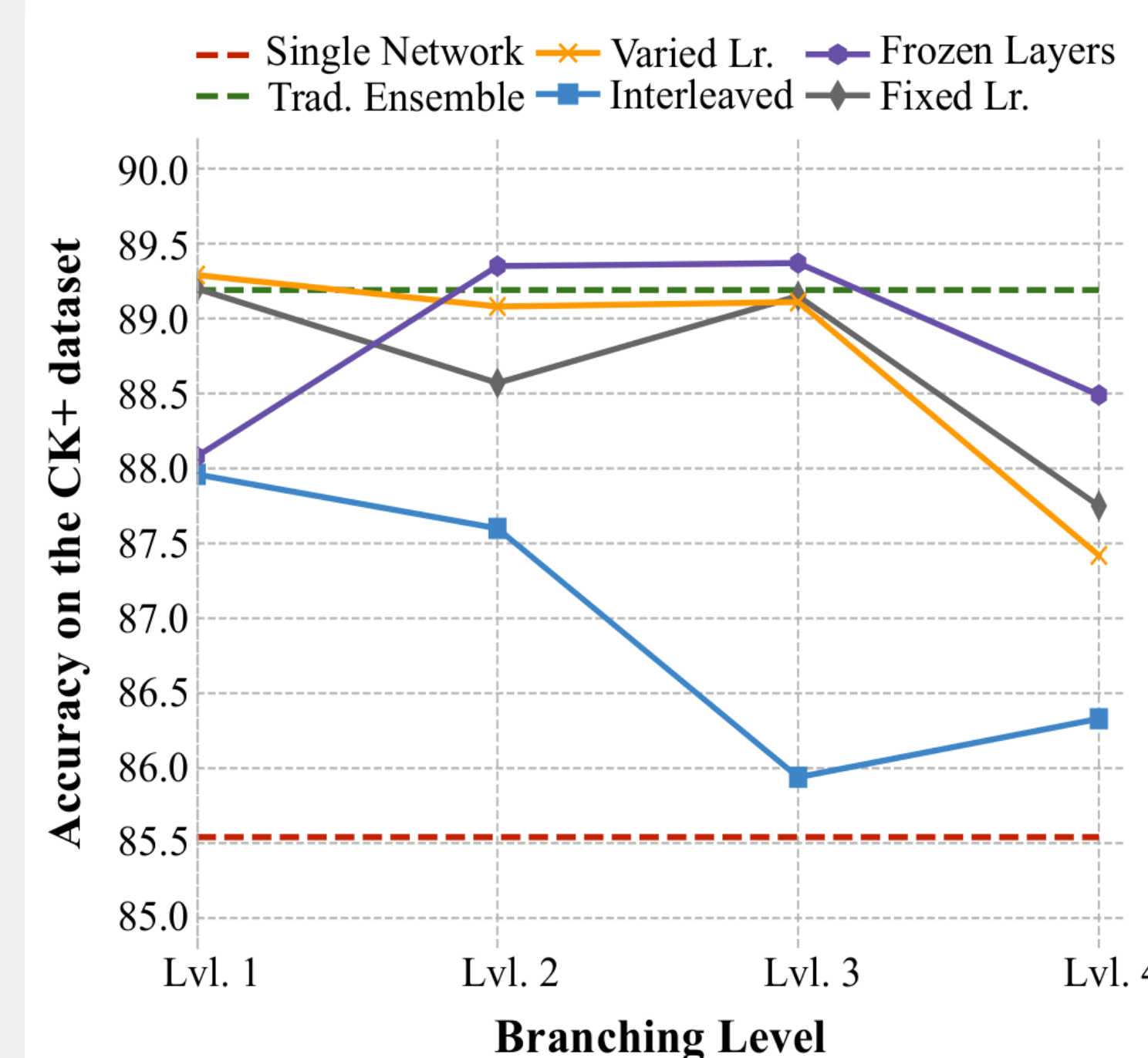
- Costly training.
- High redundancy.
- Inefficient data processing.

Ensembles with Shared Representations:

- Shared layers are responsible for the reduction of redundancy, training and inference time.
- Convolutional branches learn complementary visual representations from data.
- Inductive transfer learning from a combined loss under a multitask learning setup.

Results:

- Efficient transfer learning and data processing.
- Robust against unbalanced distributions.
- Outperformed state-of-the-art methods on in-the-wild benchmarks of facial expressions.



Approach	#	Accuracy
Single Network	131.208	85.5 ± 3.5%
Traditional Ensemble	524.832	89.2 ± 1.2%
ESR-4 Lvl. 3	355.104	89.4 ± 2.2%
ESR-4 Lvl. 4	243.936	88.5 ± 3.8%

	TE	Lvl. 3	Lvl. 4
Single Network	0.004 ✓	0.005 ✓	0.043 ✓
Trad. Ensemble (TE)	—	0.956 ✗	0.614 ✗
Lvl. 3	—	—	0.514 ✗

Approach	#	Acc ↓
ESR-9 (Our network)	8	59.3%
AlexNet-WL (Mollahosseini et al. 2019)	8	58.0%
VGGNet (Hewitt and Gunes 2018)	8	58.0%

Approach	RMSE	
	Aro ↓	Val
ESR-9 (Our network)	0.33	0.36
VGG16-PLD (Barsoum et al. 2016)	0.37	0.41
MobileNet (Hewitt and Gunes 2018)	0.38	0.42

Approach	Acc ↓
ESR-9 (Our network)	87.15 ± 0.1%
VGG16-PLD (Barsoum et al. 2016)	84.99 ± 0.37%
VGG16-CEL (Barsoum et al. 2016)	84.72 ± 0.24%

FER+										#
Target	Ne	89.8	2.3	6.0	0.9	0.1	0.0	0.9	0.0	Ne 10,996
	Ha	2.5	95.0	0.7	1.2	0.0	0.0	0.7	0.0	Ha 9,038
	Sa	20.7	2.6	72.5	0.0	0.5	0.3	3.4	0.0	Sa 3,752
	Su	5.8	3.3	11.6	89.1	1.0	0.0	0.3	0.0	Su 3,941
	Fe	7.0	2.3	0.0	31.4	45.3	0.0	2.3	0.0	Fe 682
	Di	0.0	6.2	1.9	12.5	0.0	56.2	25.0	0.0	Di 157
	An	7.1	3.3	0.0	1.1	0.0	0.4	86.2	0.0	An 2,656
	Co	40.0	6.7	20.0	0.0	0.0	0.0	13.3	20.0	Co 151
Ensemble Prediction										

