ACM multimedia





DFEW: A Large-Scale Database for Recognizing Dynamic Facial Expressions in the Wild

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Southeast University

- Introduction
- **DFEW Database**
- EC-STFL
- **Experiments**
- Conclusion

Introduction

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Introduction - Background

Expression of Emotion = 7% Text + 38% Voice + 55% Facial Expression & Action



Facial Expression Recognition is an important topic!







Introduction - In-the-lab V.S. In-the-wild



(b) in-the-wild

Introduction - In-the-wild Database

Database	Modality	#Sample	Expression Distribution	Annotation Time	From
EmotioNet	image	1,000,000	23 emotions	Automatically based on AU	Web
AffectNet	image	450,000 (labeled)	8 basic expressions & Valence-Arousal	1	Web
RAF-DB	image	29,672	7 basic expressions	About 40 Times	Web
CAER-S	image	70,000	7 basic expressions	3	79 TV shows
Aff-Wild	clip	298	Valence-Arousal	8	Web
AFEW 7.0	clip	1,809	7 basic expressions	2	54 Movies
AFEW-VA	clip	600	Valence-Arousal	2	AFEW database
CAER	clip	13,201	7 basic expressions	3	79 TV shows

Lack Large-scale well-annotated Dynamic Facial Expression Database!

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Competition	Rank	Accuracy	Database	# Sample	Train/Val/Test
EmotiW 2019	Champion	62.78%	AFEW 7.0	1,809	773/383/653
EmotiW 2018	Champion	61.87%	AFEW 7.0	1,809	773/383/653
EmotiW 2017	Champion	60.34%	AFEW 7.0	1,809	773/383/653
EmotiW 2016	Champion	59.02%	AFEW 6.0	1,749	773/383/593

Champion of the 7th Audio-Video based FERW, EmotiW2019 : accuracy only 62.78%!

Lack Large-scale well-annotated Dynamic Facial Expression Database!

Introduction - DFEW Database

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DFEW	clip	16,372	7 basic expressions	10	1500 movies
		Largest!		Largest!	Largest!

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Introduction - EC-STFL Loss



EC-STFL : Expression Clustered - SpatioTemporal Feature Learning

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DFEW Database - Collection



- Clips from movies to mimic our real life
- 1500+ high-definition movies
- Extract clips manually for accurate samples
- Extract at most 20 clips each movie
- Pre-annotation: Check clips whether containing one of the seven typical emotions.
- Additional reward for rare expression samples, i.e., disgust and fear.



DFEW Database - Annotation



- Expert crowdsourcing annotation, high-quality • and time-saved.
- Annotators both from greater China, the same cultural background.
- Ten independent annotators for intensive and reliable annotations.
- Release both multi-label annotation (emotion distribution) and single-label annotation

Database	#Sample	Source	Expression Distribution	#Annotation Times	Available?
Aff-Wild	298	Web	Valence-arousal	8	Yes
AFEW 7.0	1,809	54 Movies	7 basic expressions	2	Yes
AFEW-VA	600	AFEW database	Valence-arousal	2	Yes
CAER	13,201	79 TVshows	7 basic expressions	3	Yes
DFEW	16,372	1500 movies	7 basic expressions	10	Yes
l	_argest !	Largest !		Largest !	

Largest !

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DFEW Database - Annotation

Emotions		C	lips		Porcont	
Emotions	0-2s	0-2s 2-5s 5s+		Total	reicent	
Нарру	852	1252	384	2488	20.63	
Sad	440	915	653	2008	16.65	
Neutral	832	1335	542	2709	22.46	
Angry	762	1091	376	2229	18.48	
Surprise	691	648	159	1498	12.42	
Disgust	71	58	17	146	1.22	
Fear	408	435	138	981	8.14	
Total	4056	5734	2269	12059	100.00	



Single-labeled:

- Selected from the multi-labeled, i.e., all 16,372 clips with 7-dim emotion ground truth.
- At least 6 annotators (10 totally) believe this clip belong to one specific emotion.

DFEW Database – Agreement Test

Fleiss's Kappa test:

to discuss the annotation's quality

Order	Нарру	j-th Emotion	•••	Fear
1	10	0	•••	0
2	8	1	•••	0
i-th	0	0	•••	3
•••	•••		•••	•••

$$\begin{cases} p_j = \frac{1}{N \times n} \sum_{i=1}^N n_{ij} \\ \sum_{j=1}^K p_j = 1 \end{cases}$$

$$P_i = \frac{1}{n \times (n-1)} \left[(\sum_{j=1}^K n_{ij}^2) - n \right]$$
$$\bar{P} = \frac{1}{N} \sum_{i=1}^N P_i \qquad \bar{P_e} = \sum_{j=1}^k p_j^2$$
$$\kappa = \frac{\bar{P} - \bar{P_e}}{1 - \bar{P_e}}$$

- n_{ij} : the number of annotators who assigned the i-th clip and the j-th emotion.
- p_j : the proportion of all assignments which were to the j-th emotion.

DFEW Database – Agreement Test

Fleiss's Kappa test:

to discuss the annotation's quality

Order	Нарру	j-th Emotion	• • •	Fear
1	10	0	•••	0
2	8	1	•••	0
i-th	0	0	•••	3
•••	•••		•••	•••

κ	Interpretation
<0	Poor agreement
0.01-0.20	Slight agreement
0.21-0.40	Fair agreement
0.41-0.60	Moderate agreement
0.61-0.80	Substantial agreement
0.81-1.00	Almost perfect agreement

- n_{ij} : the number of annotators who assigned the i-th clip and the j-th emotion.
- p_j : the proportion of all assignments which were to the j-th emotion.

Types	κ
Single-labeled DFEW	0.63
Multi-labeled DFEW	0.70

DFEW Database - Demo Samples



Нарру



Sad



Neutral



Angry

Surprise

Disgust



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Experiments - EC-STFL Experiments

EC-STFL Loss: Expression Clustered SpatioTemporal Feature Learning Target: Loss: $L = L_s + \lambda L_{EC-STFL}$ $\min_{W} \sum_{i,j} \frac{P_{ij}\phi(x_i, x_j)}{Q_{ij}\phi(x_i, x_j)}$ $L_{EC-STFL} = \frac{\sum_{1 \le i, j \le n, x_j \in \mathcal{N}\{x_i\}} \frac{\|x_i - x_j\|}{N_{x_i}}}{\sum_{1 \le i, j \le n, x_j \notin \mathcal{N}\{x_i\}} \frac{\|x_i - x_j\|}{N_{x_j}}}$ $\phi(x_i, x_j) = \|x_i - x_j\|$ $P_{ij} = \begin{cases} 0, & \text{if } x_i \text{ and } x_j \text{ has the same label} \\ 1, & \text{otherwise} \end{cases}$ *L*_s : Softmax loss $Q_{ij} = \begin{cases} 0, & \text{if } x_i \text{ and } x_j \text{ has the different label} \\ 1, & \text{otherwise} \end{cases}$ $N{x_i}$: the same labeled set of sample x_i in mini-batch. N_{x_i} : the set size of N{ x_i }. $x \in \mathbb{R}^d$: extracted from the final : mini-batch size. hidden fully connected layers

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Data Protocol:

- For single-labeled DFEW database with 12,059 video clips.
- Using 5-fold cross-validation protocol for single-labeled benchmarks.

Evaluation:

- UAR: Unweighted Average Recall, i.e., the accuracy per class divided by the number of classes without considerations of instances per class.
- WAR: Weighted Average Recall, i.e., accuracy

Preprocessing:

- Acquire Face region and landmarks: face++ API
- Remove the non-face/undetected frames: manually
- Remove the clips which useful frames less than 50%: remove 362 clips
- Face affine transformation: Seetaface toolbox and face++ landmarks
- Fixed temporal length: Time Interpolation method

Model		Metric							
Widdei	Happy	Sad	Neutral	Angey	Surprise	Disgust	Fear	UAR	WAR
C3D	75.17	39.49	55.11	62.49	45.00	1.38	20.51	42.74	53.54
P3D	74.85	43.40	54.18	60.42	50.99	0.69	23.28	43.97	54.47
R3D18	79.67	39.07	57.66	50.39	48.26	3.45	21.06	42.79	53.22
3D Resnet18	73.13	48.26	50.51	64.75	50.10	0.00	26.39	44.73	54.98
I3D-RGB	78.61	44.19	56.69	55.87	45.88	2.07	20.51	43.40	54.27
VGG11+LSTM	76.89	37.65	58.04	60.70	43.70	0.00	19.73	42.39	53.70
Resnet18+LSTM	78.00	40.65	53.77	56.83	45.00	4.14	21.62	42.86	53.08

Get start easily:

some baseline for the evaluation of methods

Experiments - EC-STFL Experiments

		_	Han	75.17	7.40	8.34	7.61	0.74	0.20	0.53	Han	75.87	8.10	8.02	6.79	0.74	0.12	0.37
Model	Me	tric	nap	16 47	39.49	25 55	8 55	5 70	0 32	3 91	nap	11 89	49 26	21.07	8.08	4 86	0 1 1	4 75
	UAR	WAR	Sad	10.47	55.45	25.55	3.55	5.70	0.52	3.91	Sad ·	11.00	49.20	21.07	0.00	4.00	5.11	4.73
C3D	42.74	53.54	Neu	6.97	11.28	55.11	13.11	10.72	0.49	2.32	Neu ·	6.59	11.80	54.81	13.49	9.82	0.34	3.15
C3D,EC-STFL	45.10	55.50	Ang	6.81	5.43	17.63	62.49	4.37	0.23	3.04	Ang ·	6.53	7.36	16.84	61.53	4.00	0.09	3.64
P3D	43.97	54.47	Sur	2.59	7.08	30.97	8.24	45.00	0.07	6.06	Sur	2.79	7.83	25.80	8.58	45.95	0.34	8.71
P3D,EC-STFL	45.22	56.48	Dis	8.97	7.59	46.21	16.55	13.10	1.38	6.21	Dis ·	15.86	13.10	40.69	15.17	6.90	3.45	4.83
R3D18	42.79	53.22	Fea	4.88	14.97	20.40	16.96	22.06	0.22	20.51	Fea	4.55	15.63	17.63	16.85	20.40	0.11	24.83
R3D18,EC-STFL	45.05	56.19		Нар	Sad	Neu	Ang a)C3	Sur	Dis	Fea		Нар	Sad	Neu b)C3		Sur	Dis	Fea
3D Resnet18	44.73	54.98					.,						,		2,20			
3D Resnet18,EC-STFL	45.35	56.51	Нар	73.13	9.57	7.81	7.81	0.78	0.00	0.90	Hap ·	79.18	6.38	7.48	5.07	0.70	0.20	0.98
I3D-RGB	43.40	54.27	Sad	10.35	48.26	20.96	8.71	6.12	0.00	5.60	Sad ·	12.78	49.05	23.02	6.60	3.96	0.48	4.12
I3D-RGB,EC-STFL	45.05	56.19	Neu	5.17	13.30	50.51	17.01	10.45	0.07	3.48	Neu ·	6.89	11.65	57.85	12.85	8.06	0.49	2.21
VGG11+LSTM	42.39	53.70	Ang	5.89	6.76	14.59	64.75	4.00	0.00	4.00	Ang ·	6.99	5.61	18.45	60.98	4.33	0.46	3.18
VGG11+LSTM,EC-STFL	44.78	56.25	Sur	2.38	6.06	22.33	10.21	50.10	0.00	8.92	Sur	2.79	6.54	28.18	8.44	46.15	0.54	7.35
Resnet18+LSTM	42.86	53.08	Dis	10.34	13.10	39.31	17.24	11.72	0.00	8.28	Dis ·	11.03	14.48	45.52	14.48	6.90	2.76	4.83
Resnet18+LSTM,EC-STFL	43.60	54.72	Fea	4.43	14.41	13.97	17.41	23.39	0.00	26.39	Fea ·	5.10	12.75	18.29	18.18	23.61	0.55	21.51
	Rot	torl		Нар	Sad	Neu	Ang	Sur	Dis	Fea		Нар	Sad	Neu	Ang	Sur	Dis	Fea
					(c)3D	Res	net1	8			(d)3D) Res	net1	8,EC	-STF	L

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Experiments - EC-STFL Experiments

Madal		Metric							
Model	Нарру	Sad	Neutral	Angry	Surprise	Disgust	Fear	UAR	WAR
C3D	75.17	39.49	55.11	62.49	45.00	1.38	20.51	42.74	53.54
C3D, center loss	75.62	44.67	54.18	63.14	42.21	2.07	22.17	43.44	54.17
C3D,EC-STFL	75.87	49.26	54.81	61.53	45.95	3.45	24.83	45.10	55.50
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3D Resnet18, center loss	78.49	44.30	54.89	58.40	52.35	0.69	25.28	44.91	55.48
3D Resnet18,EC-STFL	79.18	49.05	57.85	60.98	46.15	2.76	21.51	45.35	56.51



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Experiments - Transfer Learning Task

Pretrained	Finetuned models				
	C3D	C3D, EC-STFL	3D Resnet18	3D Resnet18, EC-STFL	
Sports 1M	41.78	44.91	-	-	
UCF101	41.25	42.34	-	-	
Kinect700	-	-	49.35	49.61	
Kinect700+Moments In Time	-	-	49.35	49.35	
DFEW, fd2	44.91	45.56	53.00	53.26	Detterl
DFEW, fd5	49.8 7	49.87	49.61	49.66	Better!

DFEW, fd2: used the pre-trained models trained on the second data split.

Transfer Learning:

- From action database / DFEW database to AFEW database
- Initializing models with weights trained from source database, then go on training and test models on AFEW

Purpose:

• Verify the **necessity** of DFEW database for developing excellent emotion prediction models in real-life applications.

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Conclusion

- We present a new large-scale unconstrained Dynamic Facial Expression database in-the-Wild, DFEW.
 - 16372 video clips from over 1500 different movies.
 - Reliable distribution information of 7 basic expression annotated by 10 annotators, release both multi-labeled and single-labeled annotation.
- We propose a new EC-STFL loss to improve the performance of FERW.
- We conduct extensive experiments on DFEW
 - Extensive baseline experiments as well as EC-STFL to get DFEW database started easily.
 - Transfer tasks to verify the necessity of DFEW database.

Thanks for Listening !



DFEW project page



Email: jiangxingxun@seu.edu.cn Homepage: <u>https://jiangxingxun.github.io/</u>

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