



# Families in the Wild (FIW)

# Large-Scale Kinship Image Database and Benchmarks Joseph P. Robinson, Ming Shao, Yue Wu, Yun Fu

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#### Problem Formulation

- Automatic kinship recognition is a challenging feat
- Pre-existing datasets do not properly represent true data distributions
- Many factors are still undiscovered by the machine vision community
- · Research has not yet reached reality, i.e., technology has not matured enough to address real world problems and data

## Goal: Build and Benchmark a Large-scale Kinship Dataset to best support the task of kinship recognition

Dataset No.

CornellKin

UB KinFace-I

Family

150

People

180

Faces

300

Vary

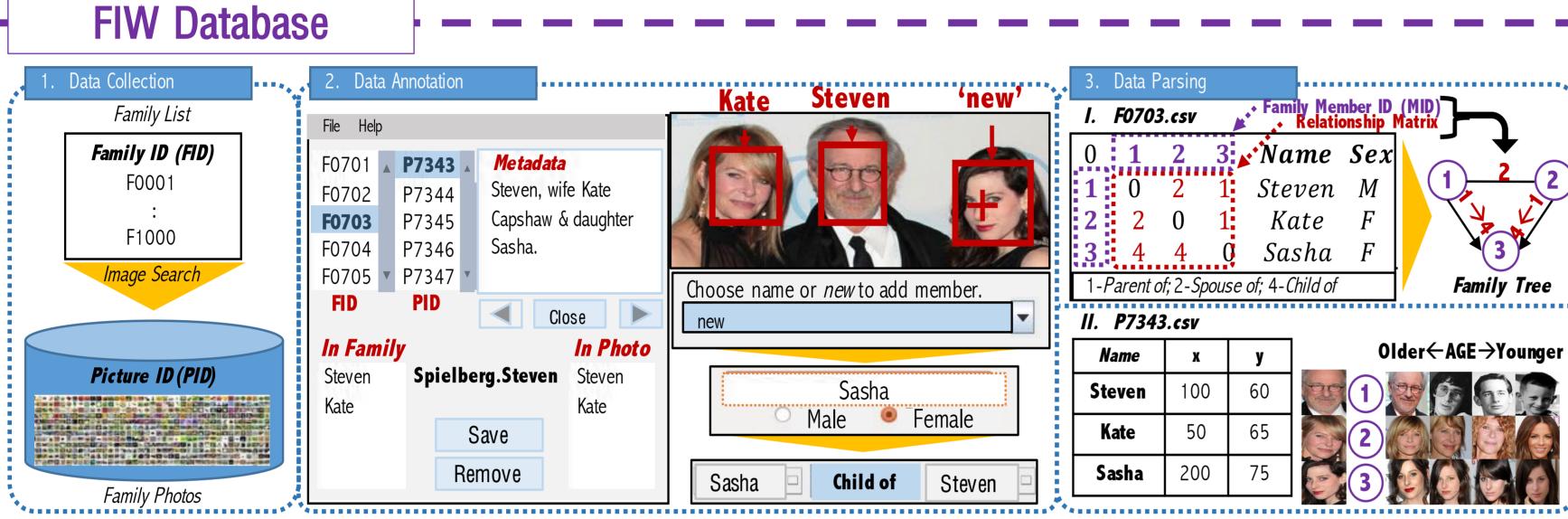
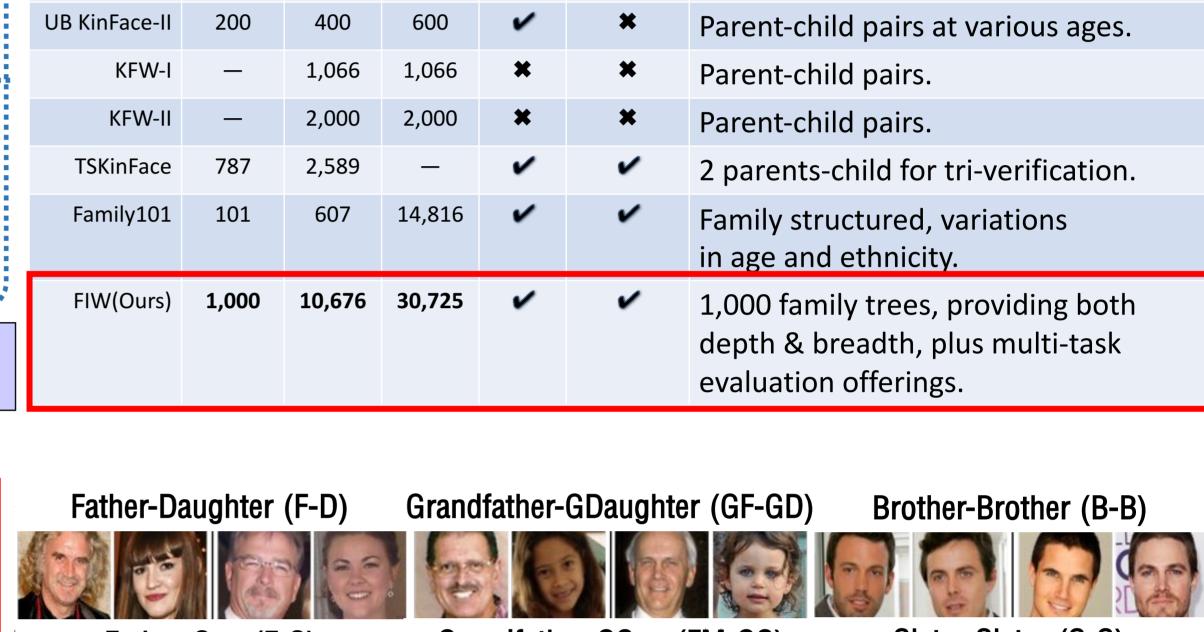


Fig 1 Process to build FIW (1) Data Collection: list of candidate families & photos were collected (2) Data Annotation: label tool to mark complex relationships for 1,000 families (3) *Post-Processing:* parsed two label-types generated by tool for verification & recognition.

- Much larger, spanning with depth and breadth (i.e., multiple generations & samples per subject)
- Quality images taken in the wild
- Abundance of full family trees; many more pair-wise samples than pre-existing datasets (i.e., far outdoes our predecessors)
- Serves multi-task purposes supported by laboratory style evaluations & benchmarked results

Table 2 Image Pair Count Comparison							
Type	KFW-II	Siblin g Face	Group Face	Family 101	FIW (Ours)		
B-B		232	40		86,000		
S-S		211	32		86,000		
SIBS		277	53		75,000		
F-D	250		69	147	45,000		
F-S	250		69	213	43,000		
M-D	787		_	148	44,000		
M-S	101		70	184	37,000		
GF-GD					410		
GF-GS					350		
GM-GD					550		
GM-GS					770		
Total	1,000	720	395	607	418,060		



Comparison of FIW with related datasets.

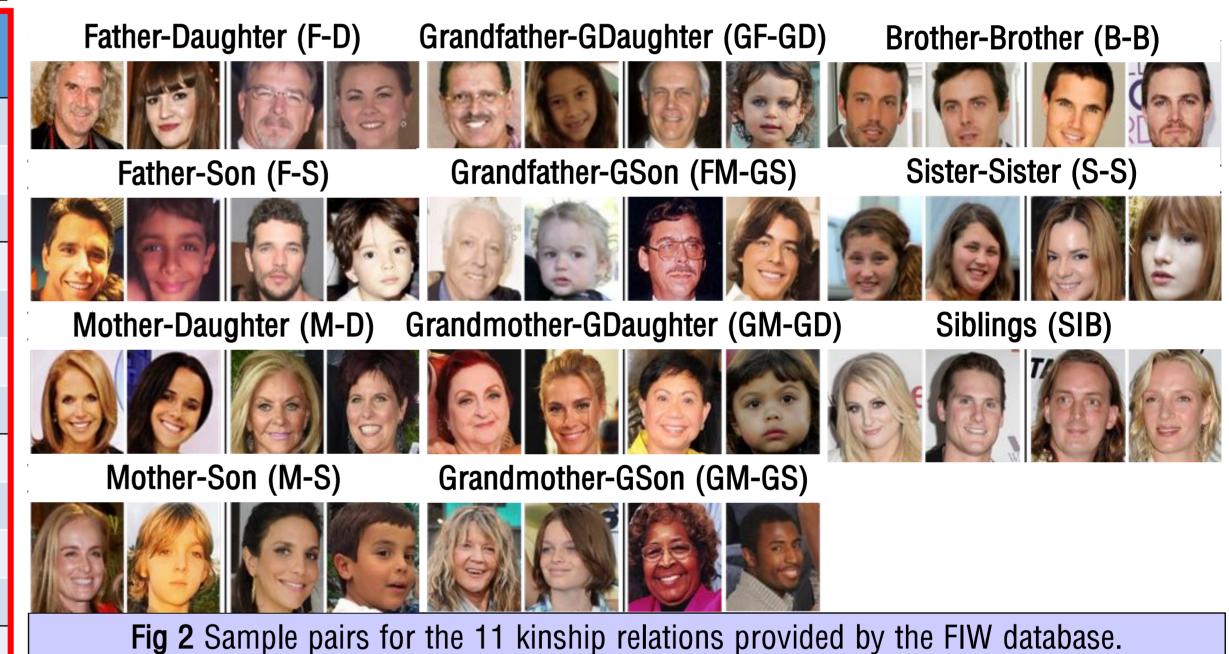
Family

Structure

Highlights

Parent-child pairs.

Parent-child pairs at various ages.



## Benchmarks

Top accuracies for each task resulted from fine-tuning the VGG-Face model.

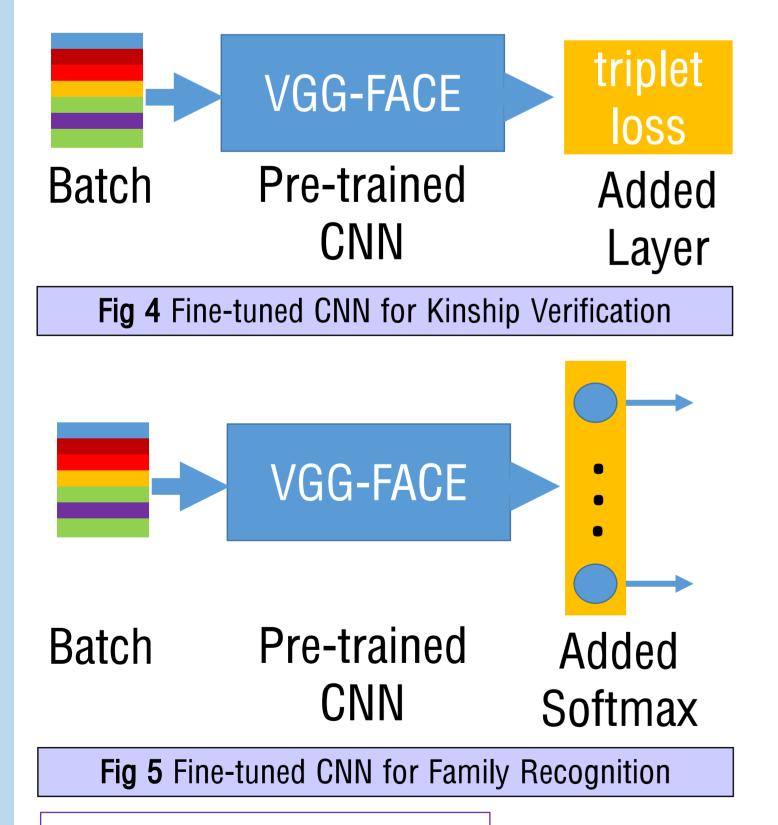
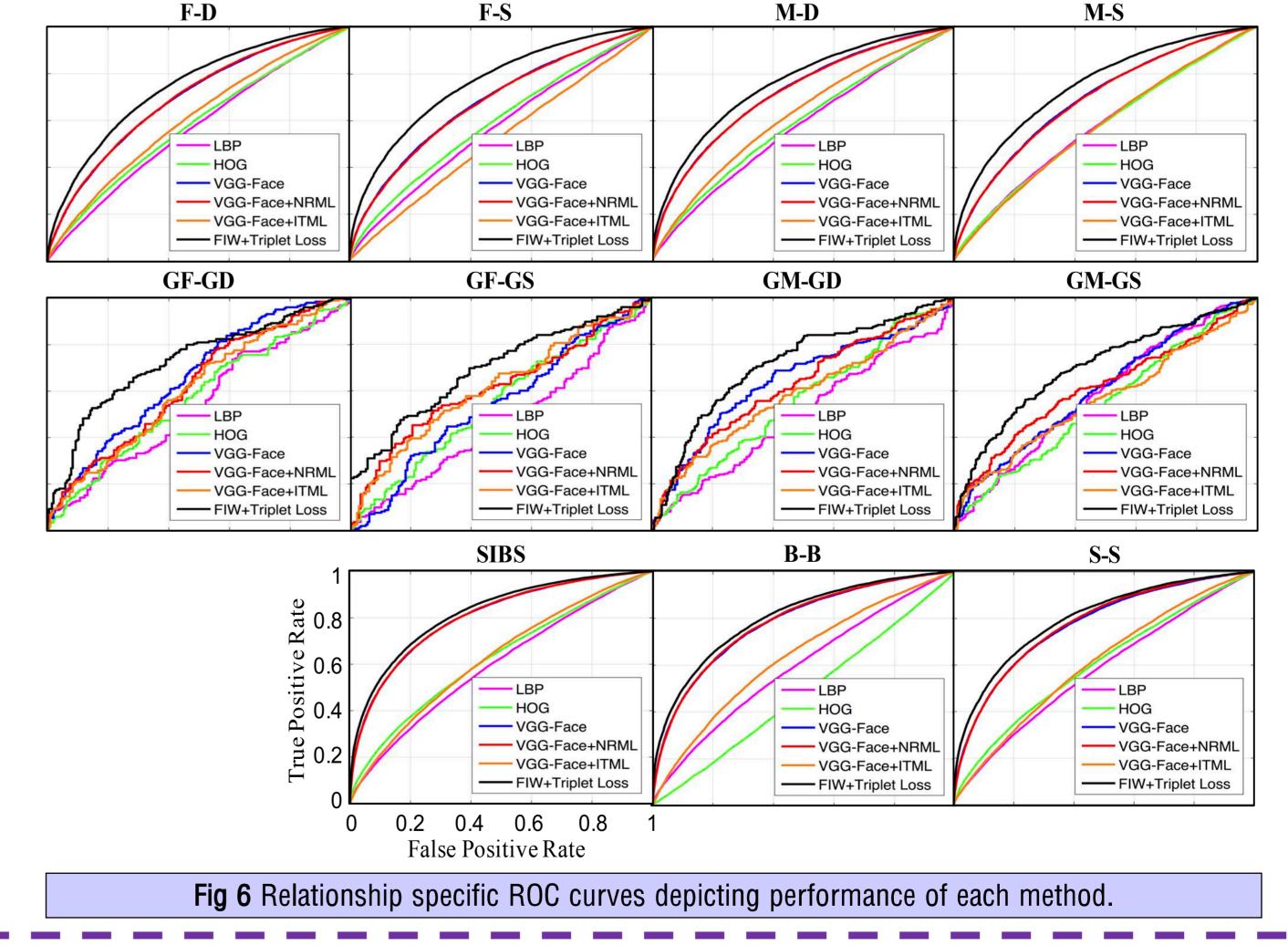


Table 3 Verification scores for 5-fold experiment.							
	HOG L	.BP \	/GG-Face	Fine-Tuned			
F-D	56.1	55	64.4	69.4			
F-S	56.5	55.3	63.4	68.2			
M-D	56.4	55.4	66.2	68.4			
M-S	55.3	55.9	64	69.4			
SIBS	58.7	57.1	73.2	74.4			
B-B	50.3	56.8	71.5	73			
S-S	57.4	55.8	70.8	72.5			
GF-GD	59.3	58.5	64.4	72.9			
GF-GS	66.9	59.1	68.6	72.3			
GM-GD	60.4	55.6	66.2	72.4			
GM-GS	56.9	60.1	63.5	68.3			
Avg.	57.7	56.8	66.9	71			
Table 4 Family recognition results, 5-fold experiment.							
Fold	VGG-Face		Fine-Tuned				
1	9.6		10.9				
2	14.5		14.8				
3	11.6		12.5				
4	12.7		14.8				
5	13.1		13.5				
Avg.	12.3		13.3				



## **Dataset**

• Finish project page with data, labels, features, source code, & CNN models

### **Evaluations**

Release additional benchmarks

**Discussion** 

Search & retrieval (missing child); fine-grain categorization (build family trees), & more

### Better results

• Further investigation of deep learning techniques for these problems

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