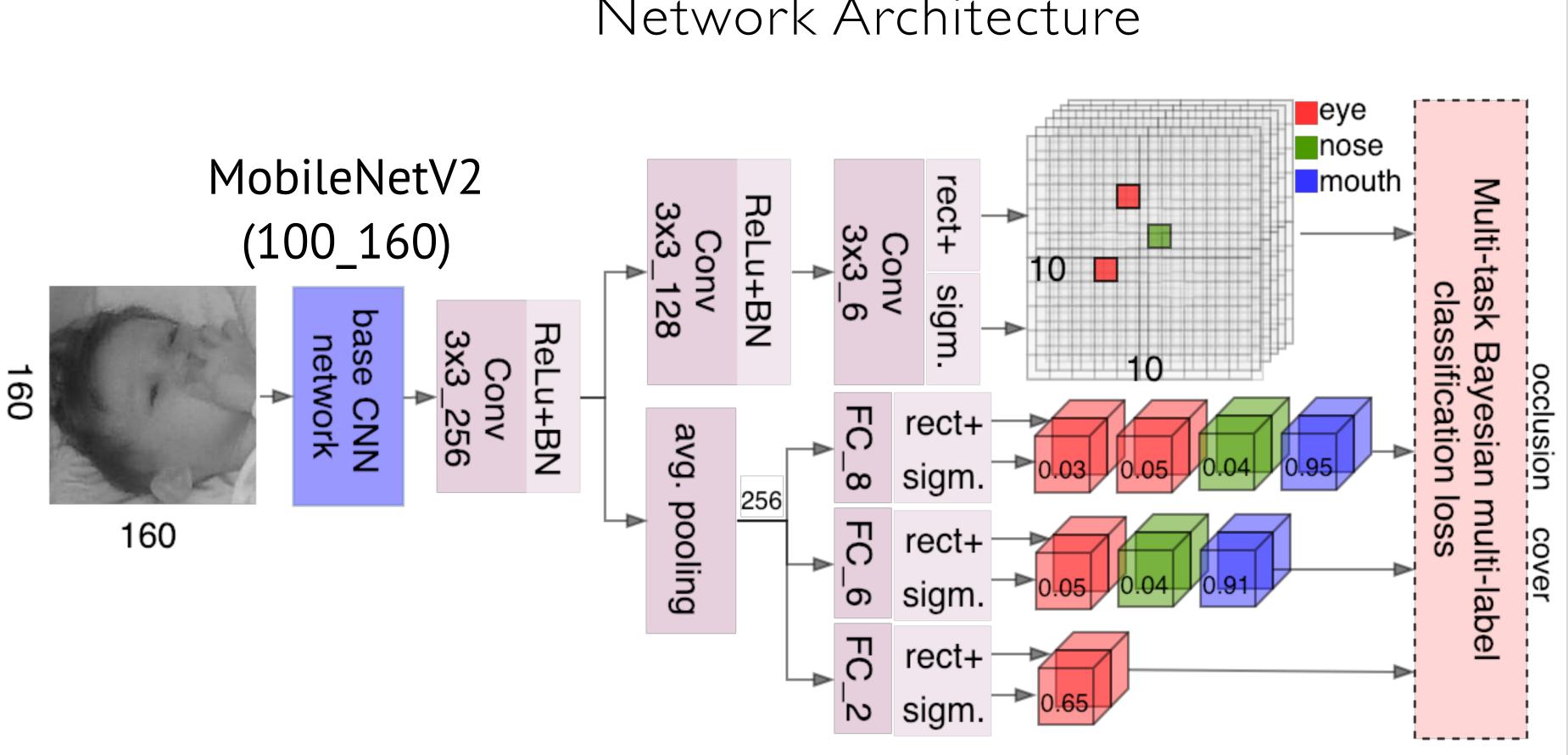


The Problem and Contributions

- **Analyze visual signs** that links to **discomforts** in infants \* **Detect**
- \* Visible facial landmarks
- \* Occlusions / Covering on facial parts
- \* Eye openness
- \* Main Contributions
- \* Presented a novel computer vision application for neonatal care
- \* Proposed a multi-task multi-label loss that accounts for (aleatoric) data uncertainty
- \* Showed that the model which accounts for data uncertainty outperforms other baseline models that do not

Network Architecture



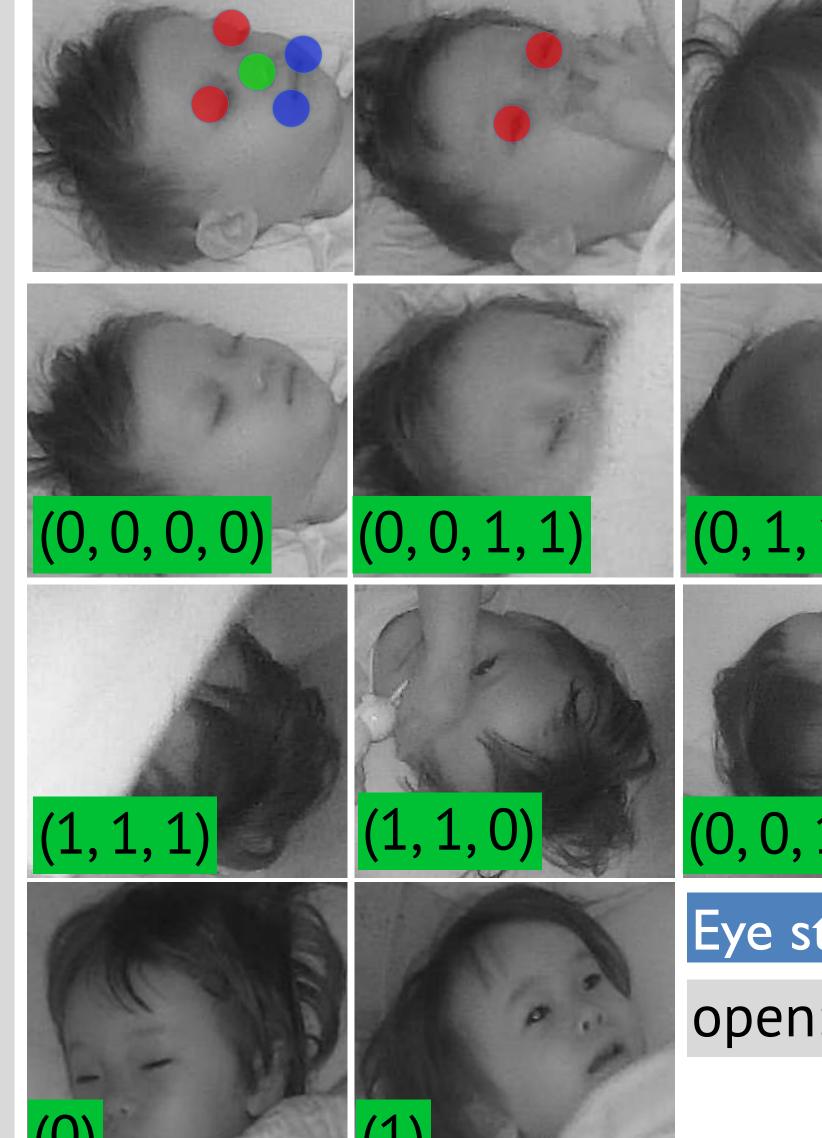
YunInfants Dataset and Labeling Protocols

\* We introduced YunInfants dataset, 19716 images captured by Cubo

\* Contain daytime and night-vision images, varying head positions

\* Age group from 0 to 2

# A Multi-Task Bayesian Deep Neural Net for Detecting Life-Threatening Infant Incidents From Head Images Julius Wang, Jorma Laaksonen 🔶, Yi-Ping Liao 🗖 , Bo-Zong Wu, and Shih-Yun Shen 🔷 🔶 Aalto University, Espoo, Finland 🔎 Independent Researcher 🔶 Yun yun Al Baby camera Co., Ltd., Taiwan (R.O.C)



# Number of images in train /

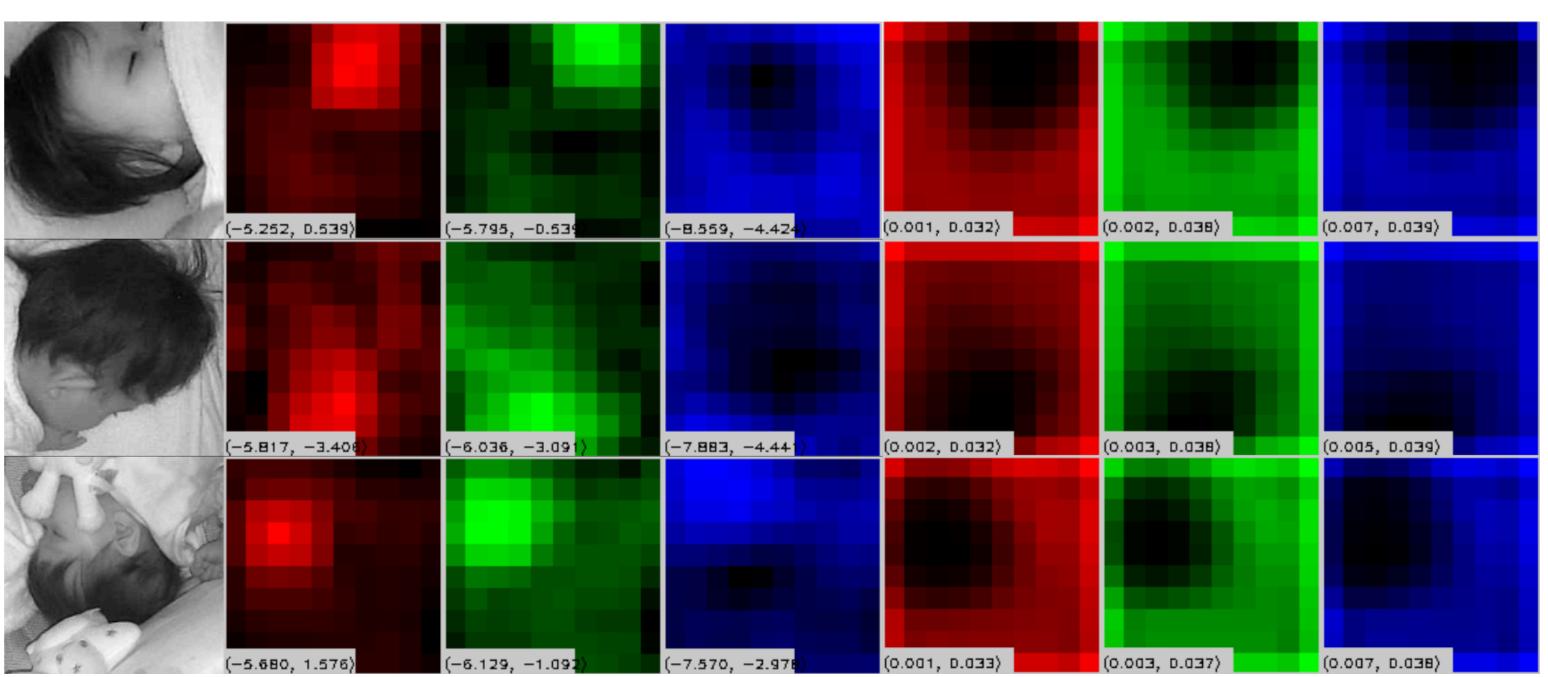
	train / val / test							
% of occl.	eye	eyes	nose	mouth				
	37 / 38 / 38	21/21/15	38 / 38 / 32	34/35/39				
% of cover.	eye	nose	mouth					
	12/13/13	21/21/24	34 / 35 / 39					
% of eye open	22/19/20							

	occlusion				cover			eye	all tasks		
F1-score (%)	eye	eyes	nose	mouth	avg.	eye	nose	mouth	avg.	openness	avg.
$\{0,c,e\}$ +baux	80.71	75.64	83.24	89.97	82.39	69.52	78.01	81.11	76.21	82.56	80.39
$\{0,c,e\}$ +baux+bayes	80.01	72.06	84.09	89.63	81.45	70.78	77.12	80.91	76.27	83.20	80.31
o+c+e+bayes	79.96	76.76	83.93	90.14	82.70	72.71	78.72	81.50	77.64	83.29	81.21
o+c+e+baux	80.26	75.02	83.56	90.00	82.23	71.17	79.00	81.76	77.31	84.51	81.35
o+c+e+baux+bayes	82.19	77.16	85.64	91.09	84.02	74.82	79.57	83.11	79.17	83.92	82.37

Landmarks * centers of eyes * tip of the noise * two mouth corners	Γ
(1, 1)Occlusions (1/0)(1, 0, 0, 1)* eye, eyes* nose* nose* mouth cornersCovering (1/0)	$c_*: \ \hat{p}^t_{*,c}$
* eye * nose * mouth status	
n: 1, not open: 0	
val / test: 12850 / 3211 / 3655	
train / val / test	2

# Multi-task Loss Functions with Aleatoric Uncertainty





Landmark detection provides critical visual cues for other tasks and can be used as model's diagnosis tool Much higher uncertainty revealed in the background pixels \* Uncertainty map acts as a face detector

## Experimental Results

**Total loss:**  $L = L_{occ} + L_{cov} + L_{eye} + \alpha_{lm}L_{lm}$ 

**Per-task loss:**  $L_* = -\frac{1}{T} \sum_{i=1}^T \frac{1}{c_*} \sum_{i=1}^{c_*} \log(\hat{p}_{*,c,i}^t), i: \text{ target class}$ 

number of attributes to predict for task  $* = \{occ, cov, eye, lm\}$  $c_i$ : prob. of  $c^{\text{th}}$  attribute estimated from the  $t^{\text{th}}$  sample drawn from  $\mathcal{N}(f_{*,c,i}, \sigma_{*,c,i}^2)$ 

## Prediction and Uncertainty Visualizations



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