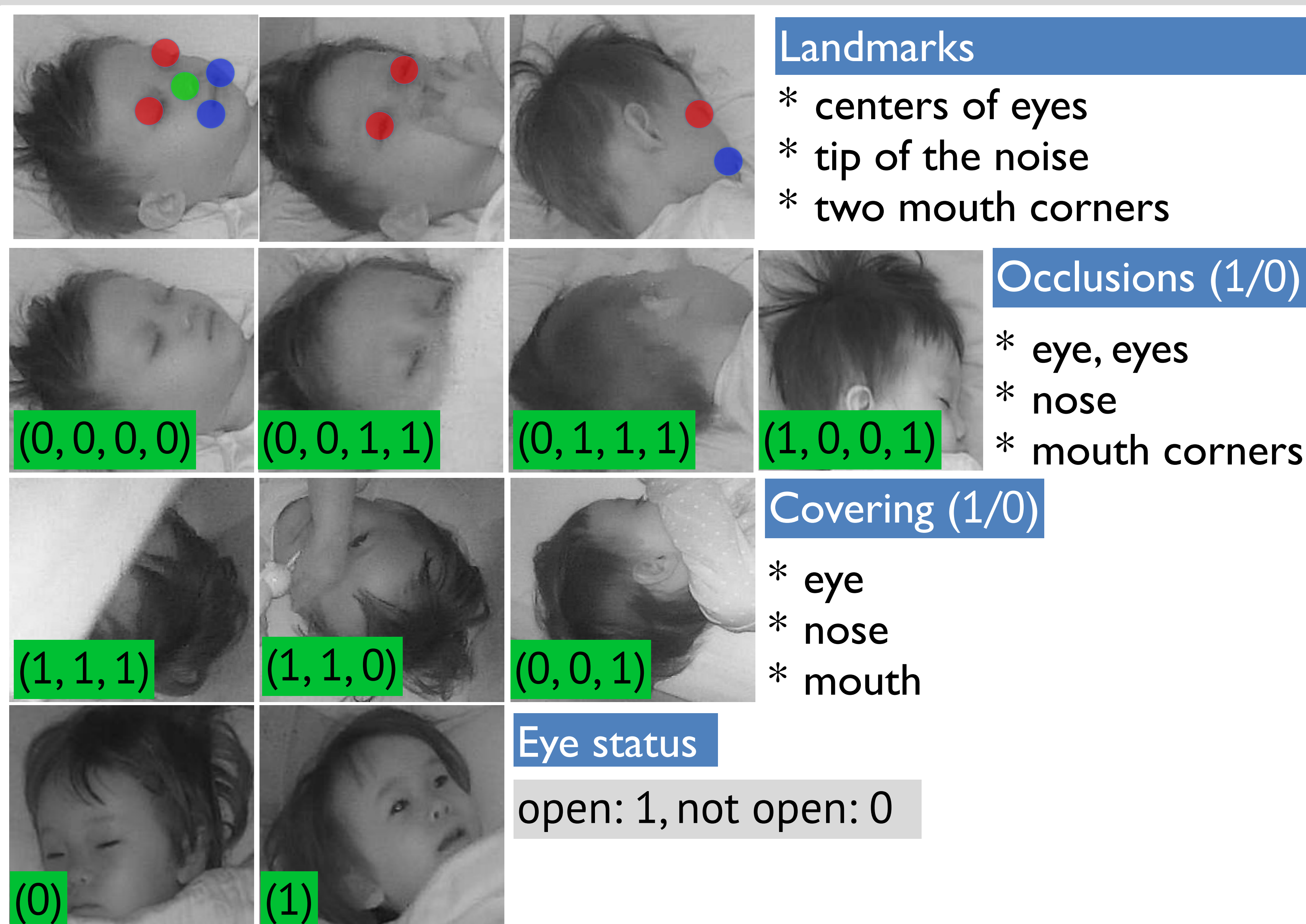
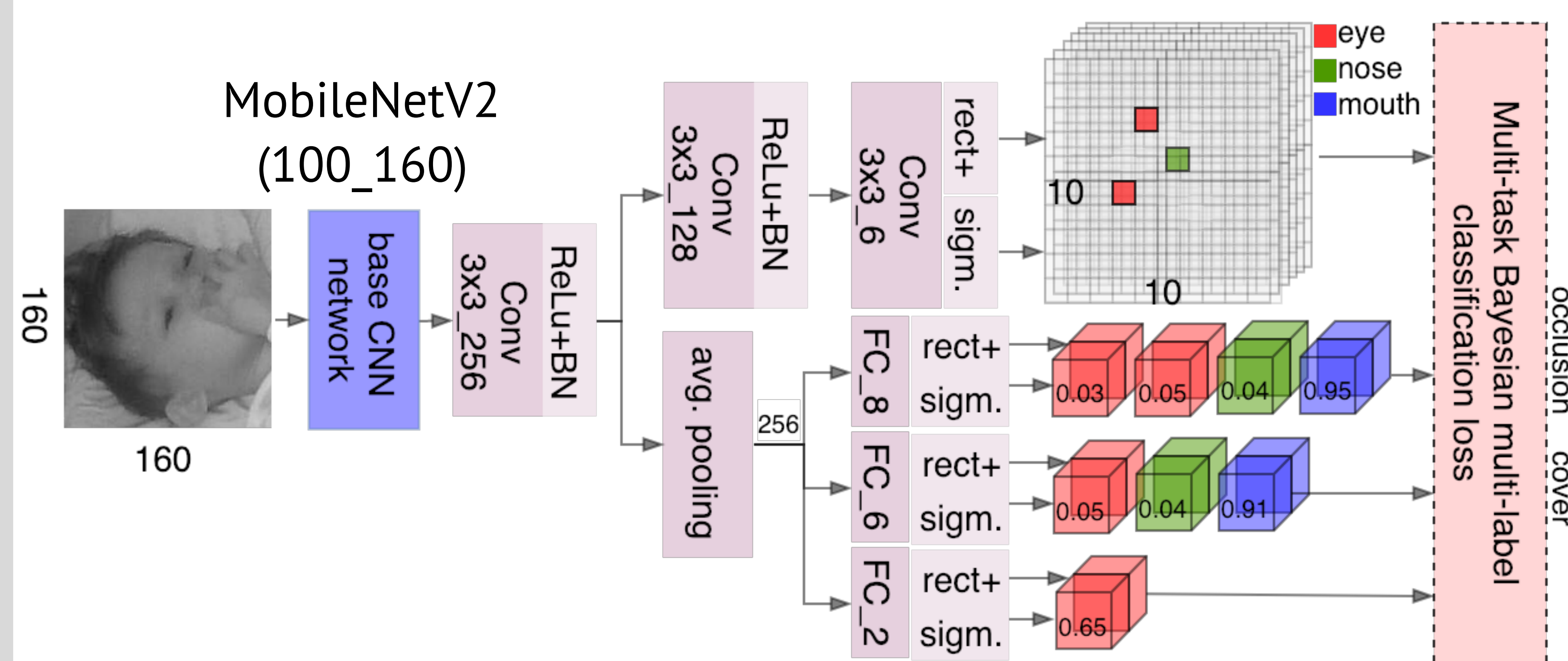


## The Problem and Contributions

- \* **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



Number of images in train / val / test: 12850 / 3211 / 3655

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

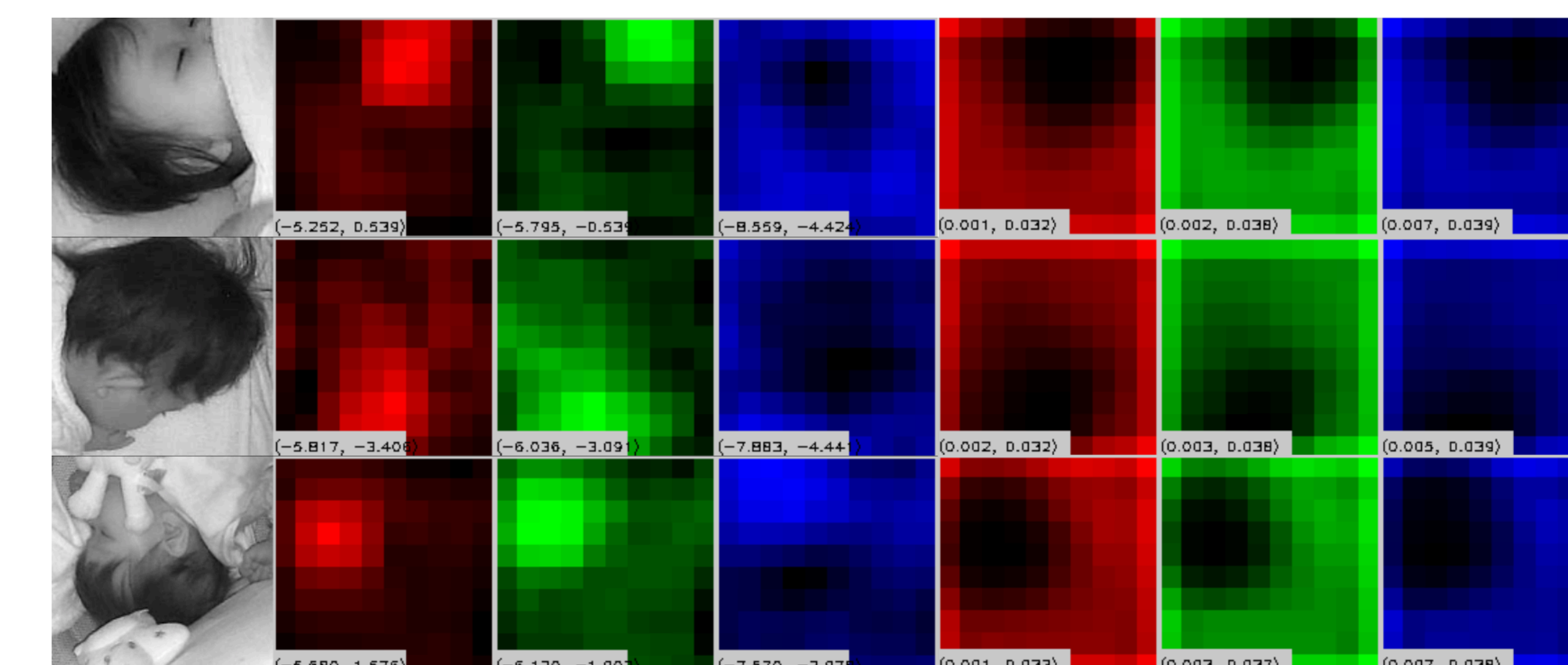
## Multi-task Loss Functions with Aleatoric Uncertainty

$$\text{Total loss: } L = L_{occl} + L_{cov} + L_{eye} + \alpha_{lm} L_{lm}$$

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

$c_*$ : number of attributes to predict for task  $*$  = {occl, cov, eye, lm}  
 $\hat{p}_{*,c,i}^t$ : 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



- \* 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

## 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

## Experimental Results

F1-score (%)	occlusion					cover				eye	all tasks
	eye	eyes	nose	mouth	avg.	eye	nose	mouth	avg.	openness	avg.
{o,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
{o,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	<b>84.51</b>	81.35
o+c+e+baux+bayes	<b>82.19</b>	<b>77.16</b>	<b>85.64</b>	<b>91.09</b>	<b>84.02</b>	<b>74.82</b>	<b>79.57</b>	<b>83.11</b>	<b>79.17</b>	83.92	<b>82.37</b>

