# Bayesian SegNet:

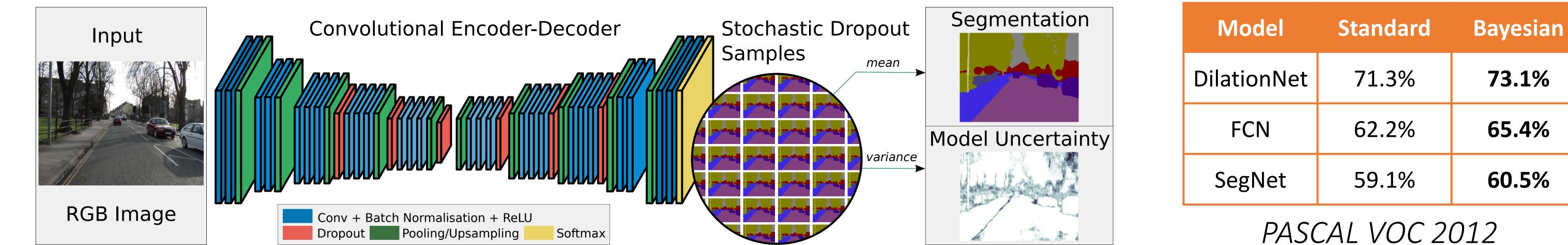
Model Uncertainty in Deep Convolutional Encoder-Decoder Architectures for Scene Understanding

Alex Kendall, Vijay Badrinarayanan and Roberto Cipolla http://mi.eng.cam.ac.uk/projects/segnet/ 

BMVC LONDON

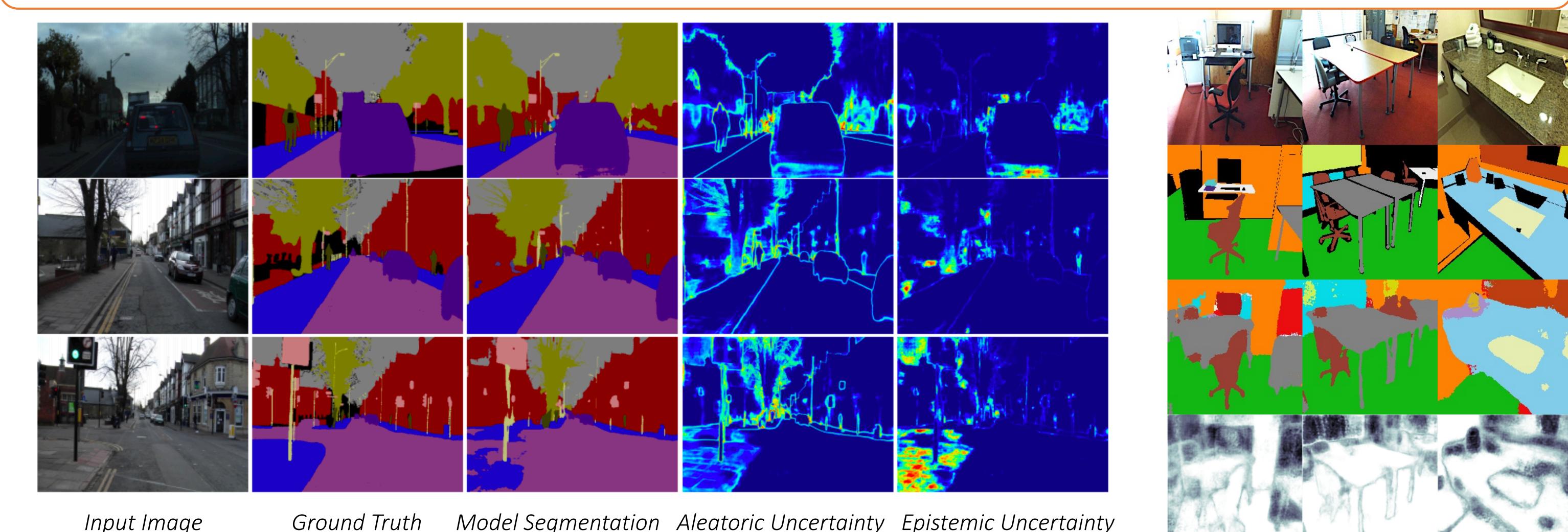
## **VERSITY OF** CAMBRIDGE

We use Monte Carlo dropout sampling at test time to generate a posterior distribution of pixel class labels.



*Test Server Performance* 

Bayesian SegNet architecture

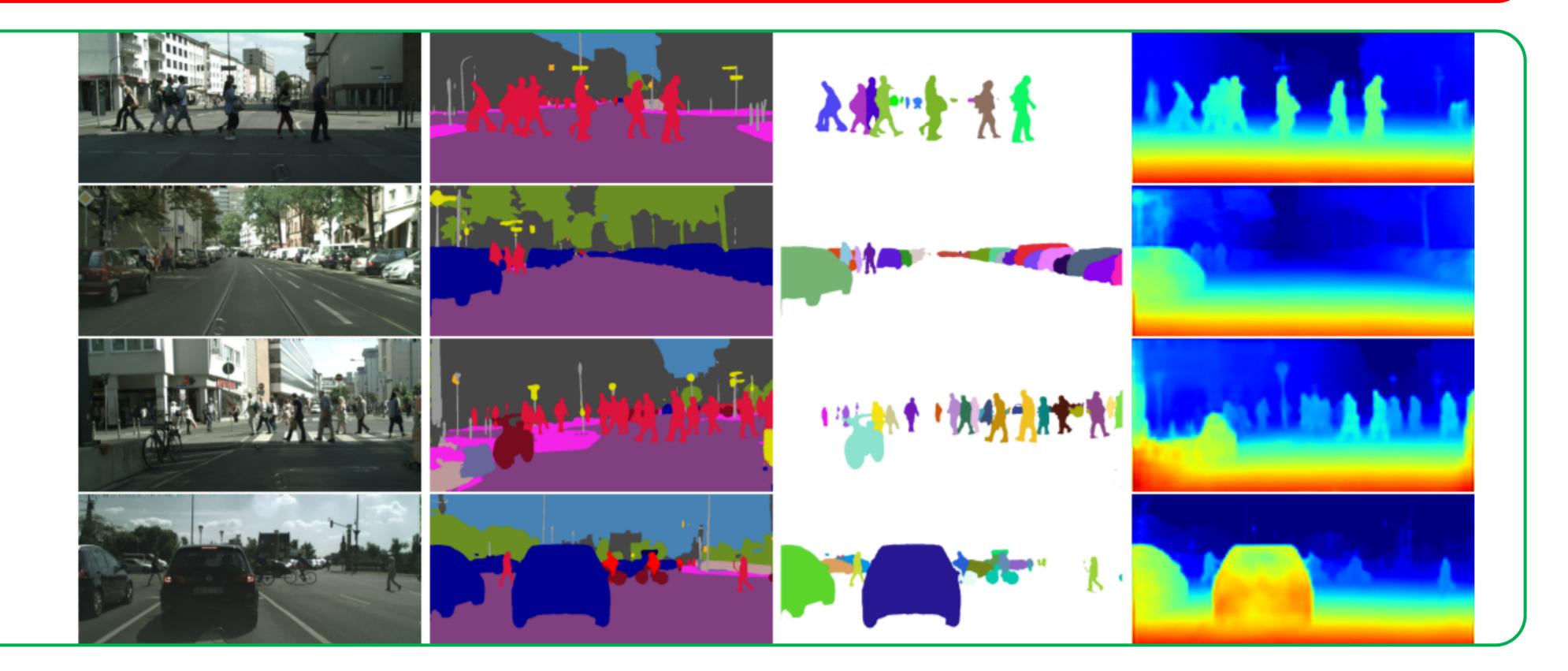


#### Insights

- We can obtain per-class model uncertainty estimates for scene understanding models
- Bayesian inference more important in late encoder and early decoder layers
- **Improves segmentation performance by 2-3%** across popular models
- MC dropout outperforms weight averaging after 6 samples and converges after 40 samples
- Especially effective for small datasets (e.g. CamVid)
- Model uncertainty increases for rare and difficult classes
- Model uncertainty is useful for safe autonomous decision making, active learning and label propagation

### **Further Applications**

Distinguish **Aleatoric** (sensor) uncertainty and **Epistemic** (model) uncertainty [3]



- Use uncertainty to improve **multi-task** learning [4]
- Semantic segmentation, instance segmentation and **depth** regression from a single input image [4]

#### References

- Vijay Badrinarayanan, Alex Kendall, Roberto Cipolla. SegNet: A Deep Convolutional Encoder-Decoder Architecture for Image Segmentation. PAMI, 2017.
- Alex Kendall, Vijay Badrinarayanan and Roberto Cipolla. Bayesian SegNet: Model Uncertainty in Deep Convolutional Encoder-Decoder Architectures for Scene Understanding. BMVC, 2017.
- Alex Kendall and Yarin Gal. What Uncertainties Do We Need in Bayesian Deep Learning for Computer Vision? arXiv preprint arXiv:1703.04977, 2017. 3.
- Alex Kendall, Yarin Gal and Roberto Cipolla. Multi-Task Learning Using Uncertainty to Weigh Losses for Scene Geometry and Semantics. arXiv preprint 4. arXiv:1705.07115, 2017.