# **Baifeng Shi**

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# EDUCATION BACKGROUND

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University of California, Berkeley	09/2021 – 06/2026 (expected)
Ph.D. student, Computer Science	
Advisor: Trevor Darrell	
Awards and Honors:	
BAIR Ignition Reward, UC Berkeley	09/2021
Peking University	09/2017 - 06/2021
B.S., Computer Science	
Advisor: Yadong Mu	
• Overall GPA: 3.75 / 4	
Awards and Honors:	
Gold Medal (3 / 360), Chinese Physics Olympiad final contest	10/2016
EECS Dean Scholarship, Peking University	09/2017
Merit Student, Peking University	09/2018 & 09/2020
RESEARCH APPOINTMENTS	
University of California, Berkeley, Research Assistant	09/2021 - now
Advisor: Prof. Trevor Darrell	
• University of California, Berkeley, Research Intern	03/2020 - 11/2020
Advisor: Dr. Huijuan Xu & Prof. Trevor Darrell	
Microsoft Research Asia, Research Intern	09/2019 - 03/2021
Advisor: Dr. Qi Dai & Dr. Jingdong Wang	
Peking University, Research Intern	09/2018 - 09/2019
Advisor: Prof. Yadong Mu	

### **PUBLICATIONS**

- Baifeng Shi, Trevor Darrell, Xin Wang, Top-down Visual Attention from Analysis by Synthesis
- Baifeng Shi, Yale Song, Neel Joshi, Trevor Darrell, Xin Wang, Visual Attention Emerges from Recurrent Sparse Reconstruction, ICML 2022
- Baifeng Shi, Qi Dai, Judy Hoffman, Kate Saenko, Trevor Darrell, Huijuan Xu, *Temporal Action Detection with Multi-level Supervision*, ICCV 2021
- Baifeng Shi, Judy Hoffman, Kate Saenko, Trevor Darrell, Huijuan Xu, Auxiliary Task Reweighting for Minimum-data Learning, NeurIPS 2020
- Zhekun Luo, Devin Guillory, **Baifeng Shi**, Wei Ke, Fang Wan, Trevor Darrell, Huijuan Xu, *Weakly-Supervised Action Localization with Expectation-Maximization Multi-Instance Learning*, **ECCV 2020**
- Baifeng Shi\*, Dinghuai Zhang\*, Qi Dai, Zhanxing Zhu, Yadong Mu, Jingdong Wang, Informative Dropout for Robust Representation Learning: A Shape-bias Perspective, ICML 2020
- Baifeng Shi, Qi Dai, Jingdong Wang, Yadong Mu, Weakly-Supervised Action Localization by Generative Attention Modeling, CVPR 2020

#### RESEARCH EXPERIENCE

Independent Research, advised by Dr. Xin Wang and Prof. Trevor Darrell

*UC Berkeley* 03/2022 – now

# **Top-down Visual Attention from Analysis by Synthesis**

- Current self-attention is bottom-up, highlighting all the objects in an image and mix the representations of all
  objects together. In contrast, top-down attention enables us to only attend to the objects that are related to the
  current task or goal.
- Showed that by formulating the visual recognition process as Bayesian Inference (Analysis by Synthesis), top-down attention will naturally emerge, where the high-level goal acts as a prior to modulate the low-level attention.
- Proposed AbSViT, a ViT model with a feedback loop which is trained to variationally approximate Analysis by Synthesis. Showed that AbSViT learns to control the top-down attention given language or class guidance.
- Applied AbSViT in various real-world scenarios, such as vision-language tasks of VQA and zero-shot recognition where a language prompt guides the attention, as well as image classification and robustness where no specific guidance is available.

#### **Visual Attention Emerges from Recurrent Sparse Reconstruction**

08/2021 - 02/2022

- Proposed a formulation of visual attention by taking inspiration from how human visual attention works, especially
  focusing on two key ingredients in human visual attention, recurrent connection and sparse representation
- Showed that adding recurrent connections into feedforward network is equivalent to adding sparse reconstruction blocks
- Showed that attention can naturally emerges from sparse reconstruction, where recurrent connections group features into separate objects and the sparsity constraint selects the most salient objects.
- Pointed out that self-attention is a special case of our attention formulation, providing a new perspective on how self-attention works.
- Observed higher robustness and more consistency with human eye fixation map.
- The work is summarized in a paper and accepted to ICML 2022.

#### Independent research, advised by Dr. Huijuan Xu and Prof. Trevor Darrell

*UC Berkeley* 06/2020 – 11/2020

#### **Unsupervised Foreground Mining for Omni-supervised Action Localization**

- Built the first baselines for semi-supervised and omni-supervised action localization.
- Designed error analysis to find the main sources of error in the baseline models.
- Proposed to solve the action incompleteness problem in the semi-supervised baseline by learning object-centric
  representations. Built a structural causal model of the foreground/background action in neighboring frames, and
  proposed to detect foreground objects by minimizing the conditional mutual information between foreground and
  background motion.
- Proposed to solve the action-context confusion problem in the omni-supervised baseline by designing an information bottleneck to discard scene information while preserve action information.
- The work is summarized in a paper and accepted to ICCV 2021.

Independent research, advised by Dr. Huijuan Xu and Prof. Trevor Darrell

UC Berkeley

## Auxiliary Task Reweighting for Minimum-data Learning

03/2020 - 06/2020

- Addressed the problem of automatically reweighting multiple auxiliary tasks to learn the main task with minimum information (supervision).
- Exploited the key insight that *information required for inference can be reduced by a good prior*, and formulated the problem as optimizing the KL divergence between the true prior and the surrogate prior given by the weighted likelihood of auxiliary tasks.
- Utilized tools and concepts including Fisher score and Langevin dynamics, and further simplified the optimization of the KL divergence into minimizing the 12 distance between main/auxiliary task gradients, which gives a light-weight algorithm to reweight auxiliary tasks on-the-fly.
- Derived theoretical guarantees that our algorithm finds the optimal task weights up to a small error.
- Experimentally observed that our algorithm finds the optimal task weights and minimizes the data requirement under various settings, *e.g.*, semi-supervised learning, domain generalization, and multi-label classification.
- The work is summarized in a paper and accepted to NeurIPS 2020.

#### Independent research, mentored by Dr. Qi Dai, Dr. Jingdong Wang, and Prof. Yadong Mu Human Vision Inspired Shape-bias for Model Robustness

*Microsoft Research Asia* 11/2019 – 02/2020

- Analyzed the relationship between the texture-bias of CNN and its multiple kinds of vulnerability.
- Proposed to discriminate texture from shape by the self-information in an image, resembling the mechanism of saliency detection and eye movement in the human visual system.
- Proposed a Dropout-like algorithm to de-correlate the model output with the texture information in the input, thus enhancing the shape-bias of the model.
- Conducted experiments under different scenarios (domain generalization, few-shot classification, image corruption, and adversarial perturbation) and observed a universal improvement in model robustness.
- The work is summarized in a paper and accepted to ICML 2020.

# Independent research, mentored by Dr. Qi Dai, Dr. Jingdong Wang, and Prof. Yadong Mu

Microsoft Research Asia 08/2019 – 11/2019

Weakly Supervised Action Localization
 Tackled with a common challenge in weakly supervised action localization, namely action-context confusion.

- Built a probabilistic graphical model and formulated the problem as modeling the frame-wise class-agnostic likelihood and optimizing the maximum a posteriori (MAP) estimation.
- Proposed to separate foreground and context by modeling the appearance-level frame likelihood using a generative model, *viz.* conditional variational auto-encoder (CVAE).
- Improved the results on two common datasets by a large margin (10% relative improvement).
- The work is summarized in a paper and accepted to CVPR 2020.

# Fast Video Understanding with Reinforcement Learning

- Proposed an algorithm to use as few frames as possible to classify a video while preserving an expected accuracy.
- Formulated the reward function as the Lagrangian of the constrained optimization problem, and optimized it using reinforcement learning algorithm, *viz.* soft actor-critic (SAC).
- Boost the original algorithm by 400% while preserving the accuracy.