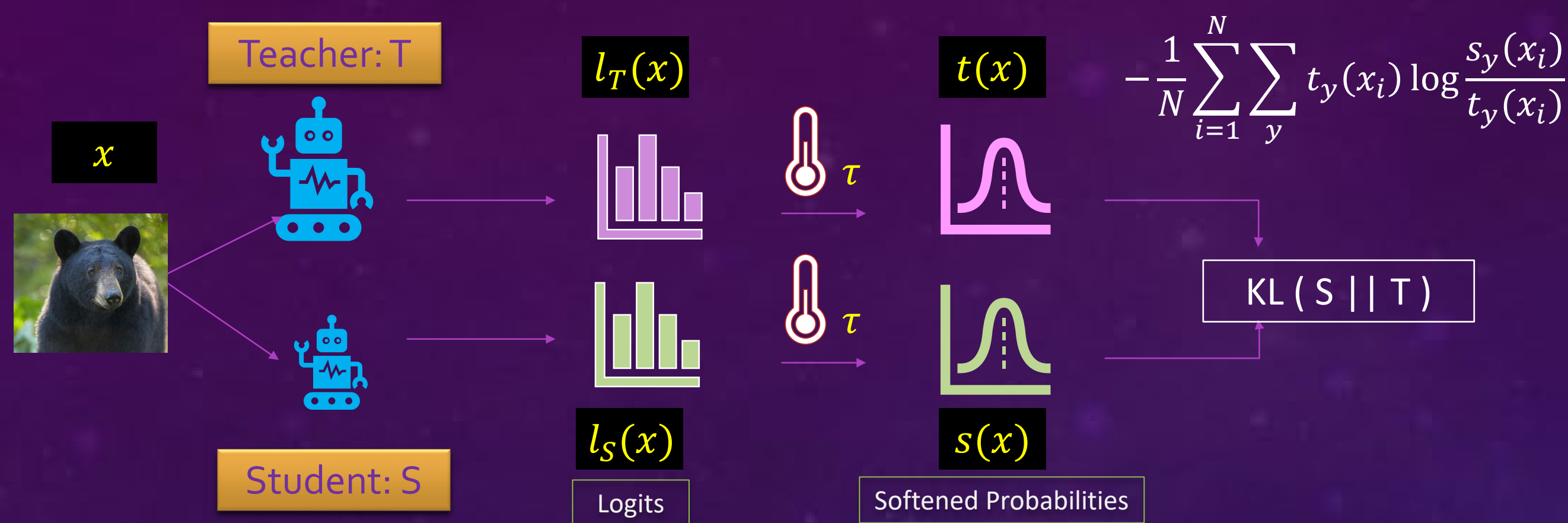
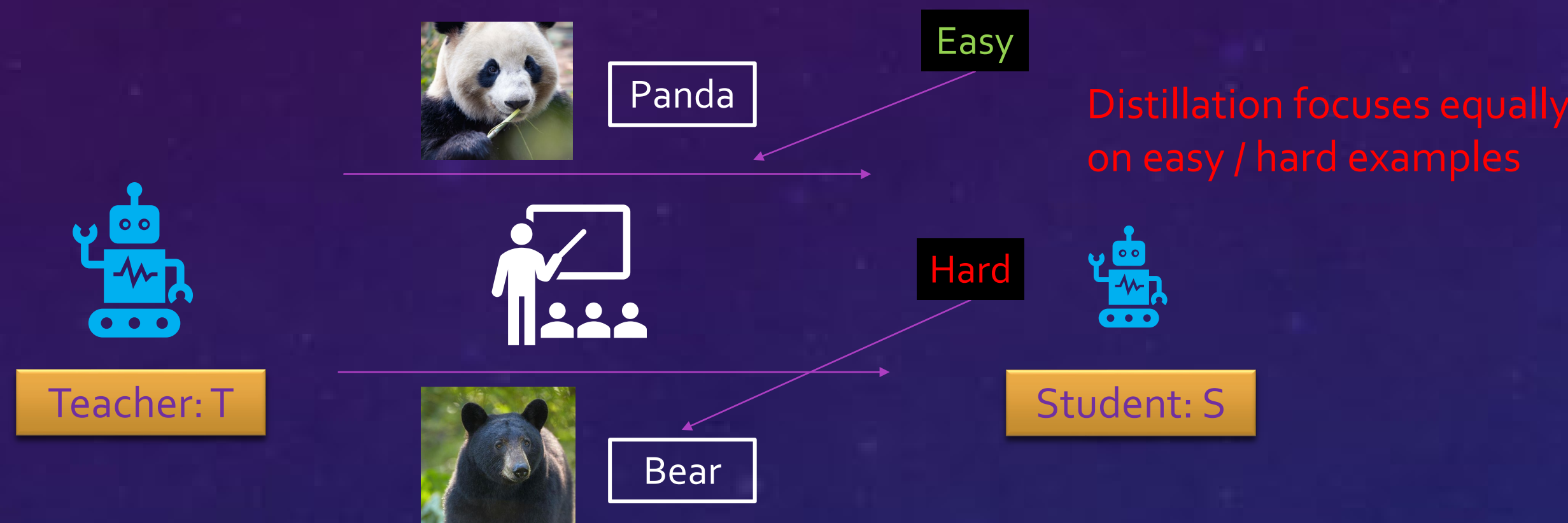


## Problem: Teaching a Tiny Model

### Knowledge Distillation Setup



### Tiny Models have limited capacity

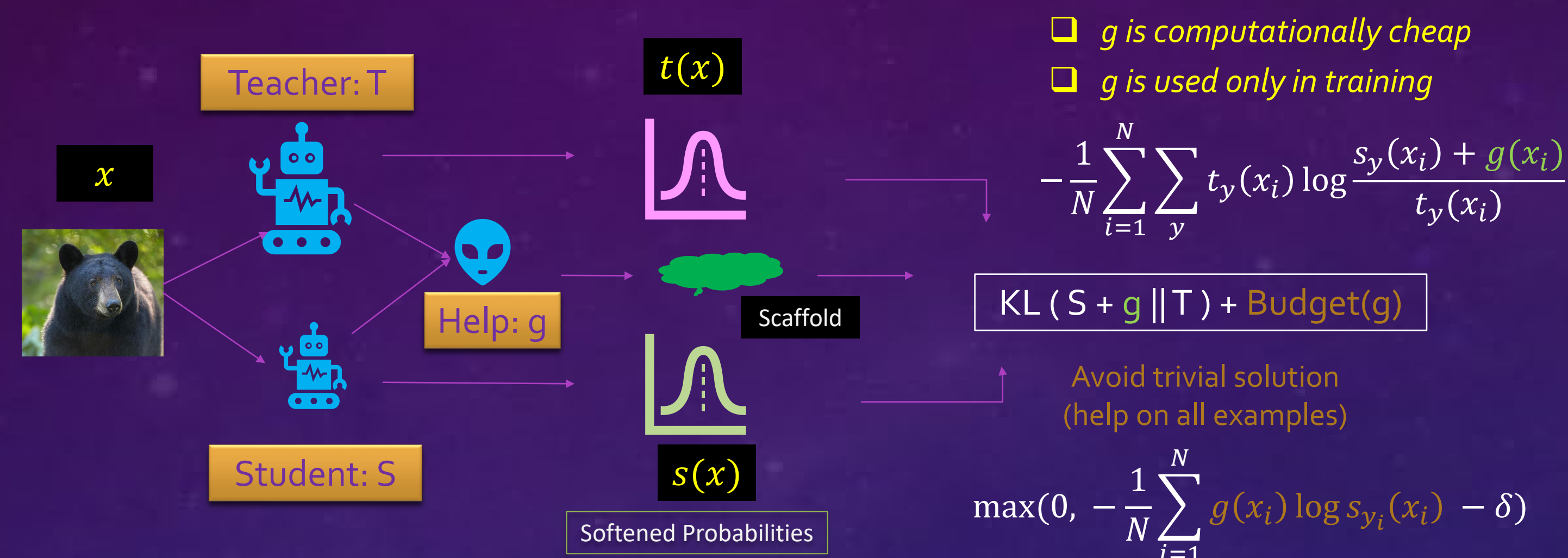


### Our Proposal : Scaffold hard-to-learn inputs for student by exploiting teacher

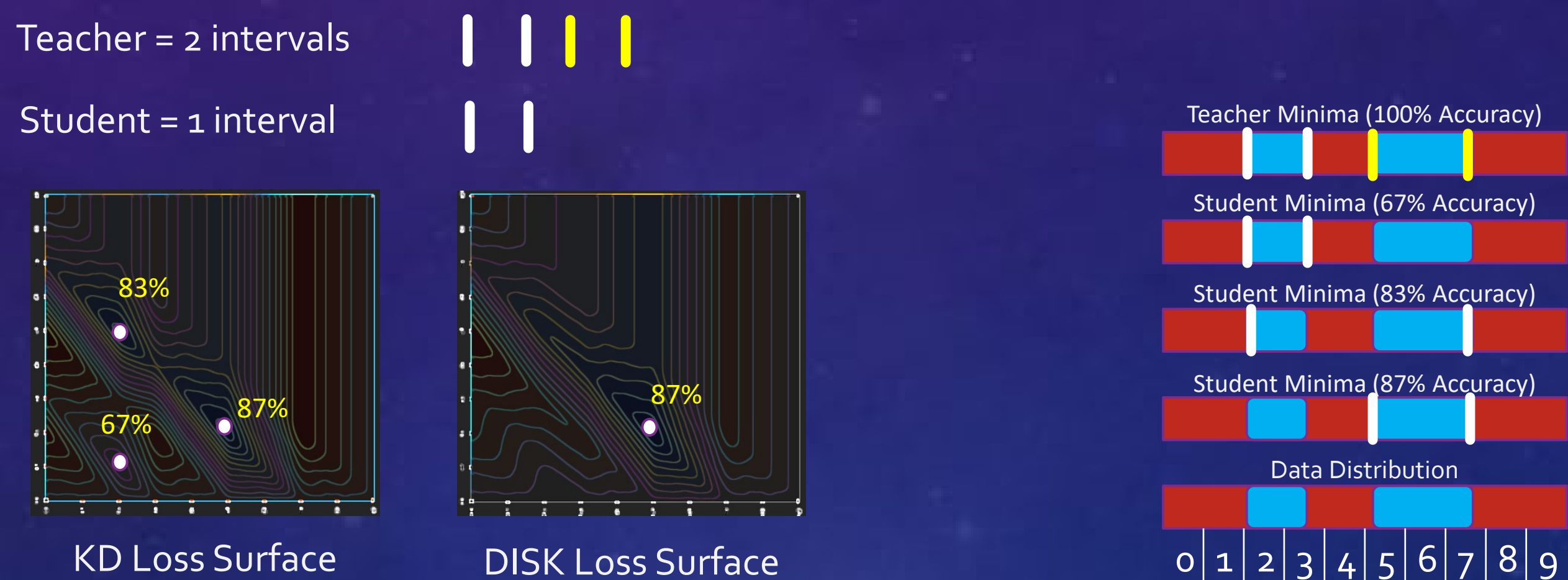


## DiSK: Distilling Scaffolded Knowledge

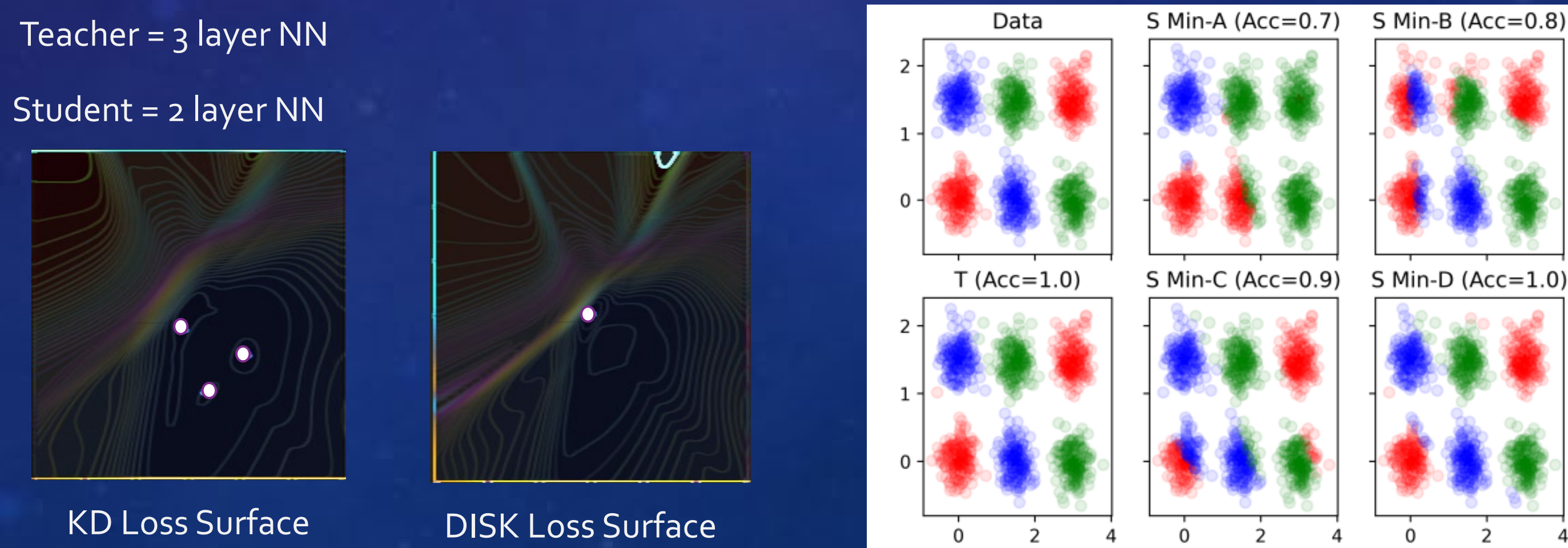
### DiSK Setup



### DiSK smoothens the student's loss-landscape, often eliminating suboptimal minima



Example: 2D Gaussians



## Empirical Evaluation

### CIFAR-100 (Tiny Students): Up to 4% higher accuracy compared to KD

Teacher	Teacher MACs	Student	Student MACs	CE	KD	DiSK
ResNet10-l	64M	Resnet10-s	4M	52.16	54.92	<b>58.14</b>
71.99%		Resnet10-m	16M	65.24	66.96	<b>70.03</b>
ResNet18	555M	Resnet10-s	4M	52.16	55.76	<b>58.11</b>
76.56		Resnet10-m	16M	65.24	68.09	<b>69.86</b>

### CIFAR-100 (Standard Students): Up to 2.5% higher accuracy compared to KD

Teacher	Teacher MACs	Student	Student MACs	CE	KD	DiSK
ResNet32x4	1083M	ShuffleNetV2	45M	73.74	79.13	<b>80.23</b>
81.45%		MobileNetV2x2	22M	69.24	76.05	<b>77.24</b>
Wide-ResNet	327M	ShuffleNetV2	45M	73.74	75.81	<b>78.33</b>
78.41		MobileNetV2x2	22M	69.24	73.92	<b>76.32</b>

### ImageNet-1K: More than 1% higher accuracy compared to KD

Teacher	Teacher MACs	Student	Student MACs	CE	KD	DiSK
ResNet50	4.12B	ResNet18	1.82B	69.73	71.29	<b>72.35</b>
ViT-Large	59.65B	ViT-Tiny	1.07B	75.45	76.61	<b>77.86</b>
ViT-Large	59.65B	DeiT-Tiny	1.07B	72.2	74.5	<b>75.59</b>

### DiSK can be integrated with other procedures such as feature matching

Teacher	Teacher MACs	Student	Student MACs	FitNet	SimKD	SimKD + DiSK
Wide-ResNet	327M	ResNet8x4	177M	75.02	76.75	<b>77.13</b>
78.41		Wide-ResNet-40-1	83M	74.17	75.56	<b>76.21</b>

### [https://github.com/anilkagak2/DiSK\\_Distilling\\_Scaffolded\\_Knowledge](https://github.com/anilkagak2/DiSK_Distilling_Scaffolded_Knowledge)