



# Bird Transformers: Modifying Birds Using Natural Language

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“Make this bird green!”



## Specialized Skills

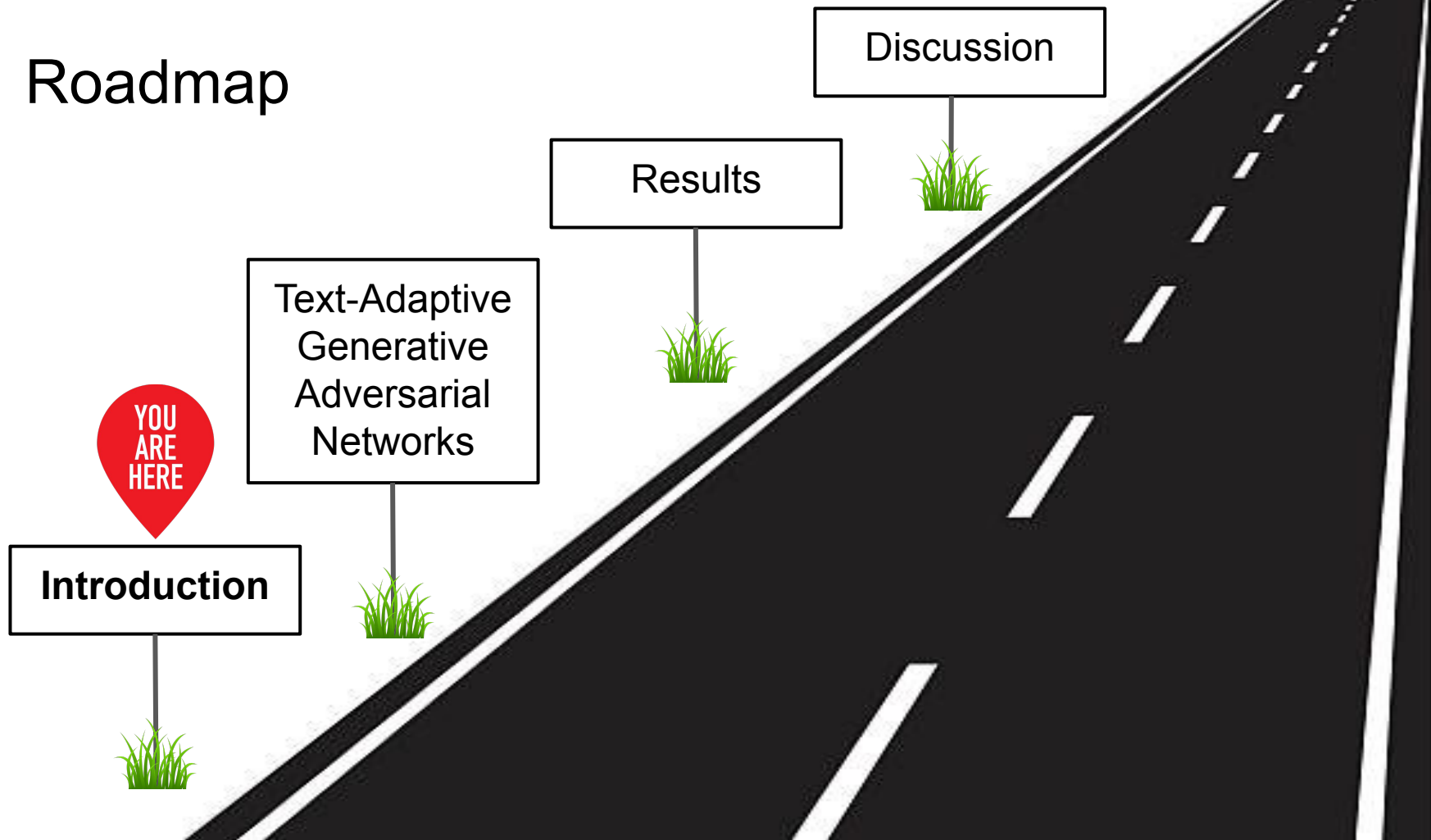


## Computers



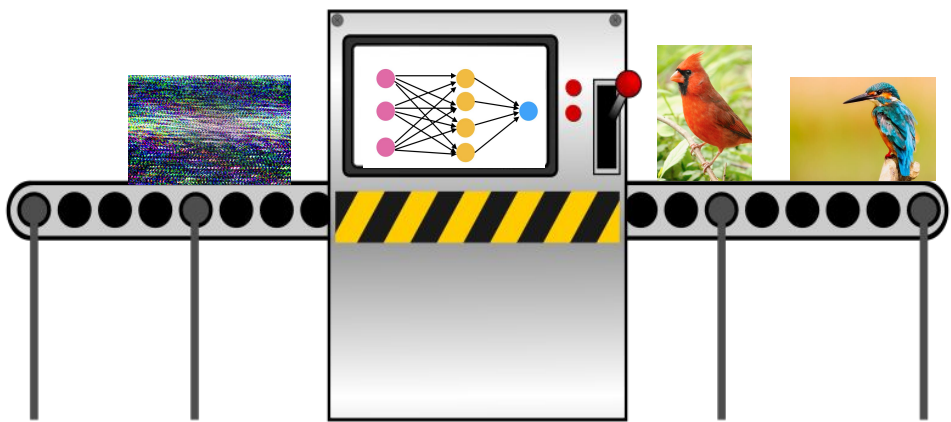
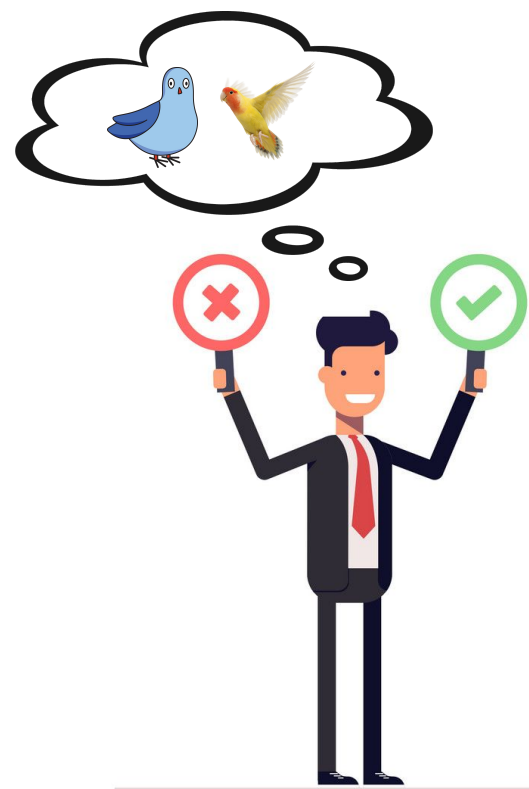
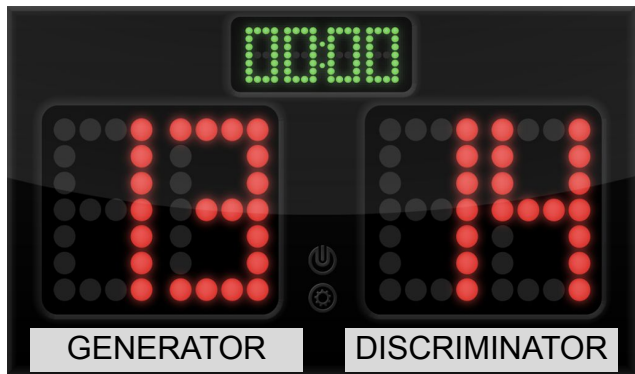
# Image Modification with Natural Language

# Roadmap



# Generative Adversarial Networks (GANs)





Generator

Discriminator

# Roadmap

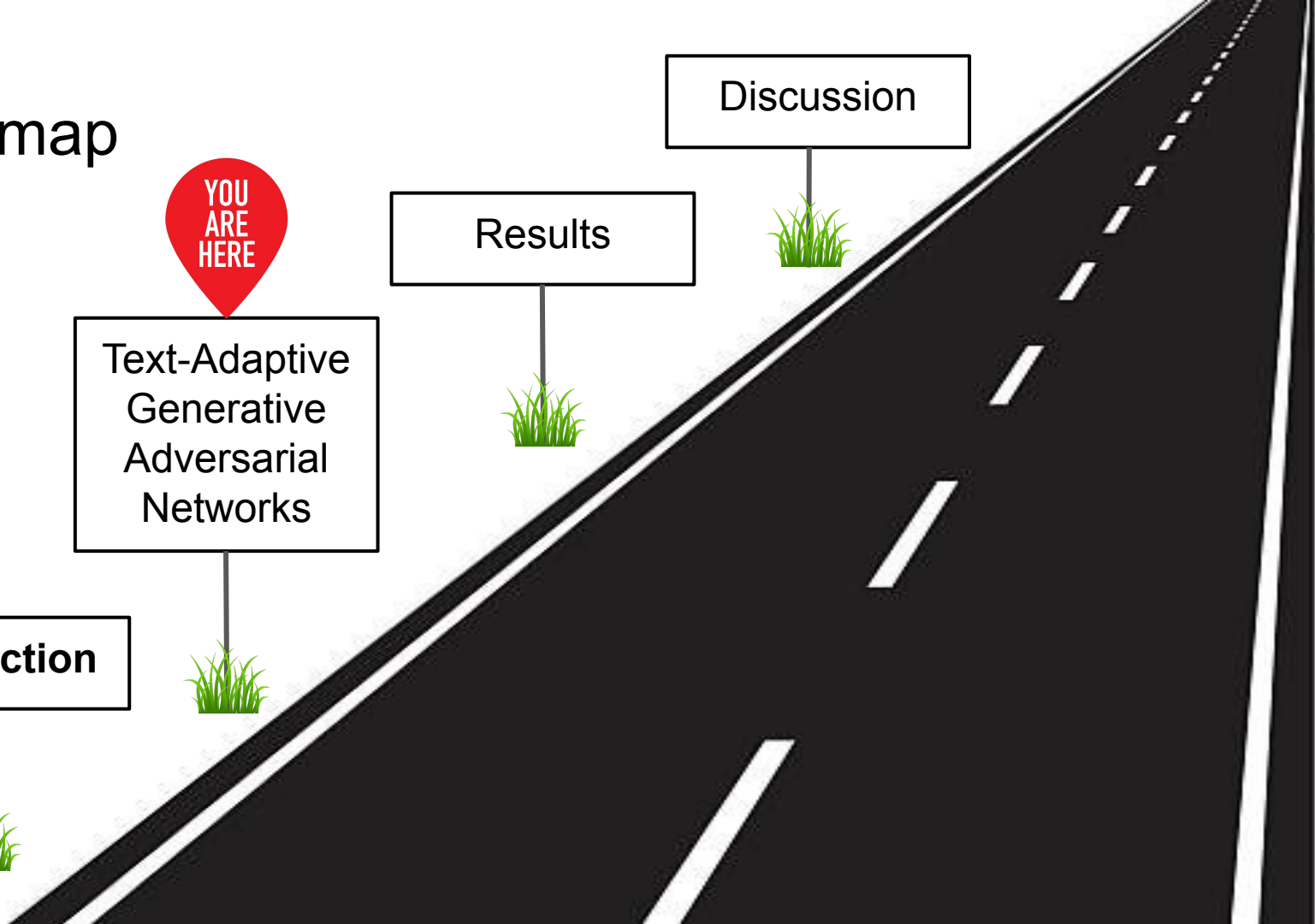


Text-Adaptive  
Generative  
Adversarial  
Networks

Results

Discussion

Introduction





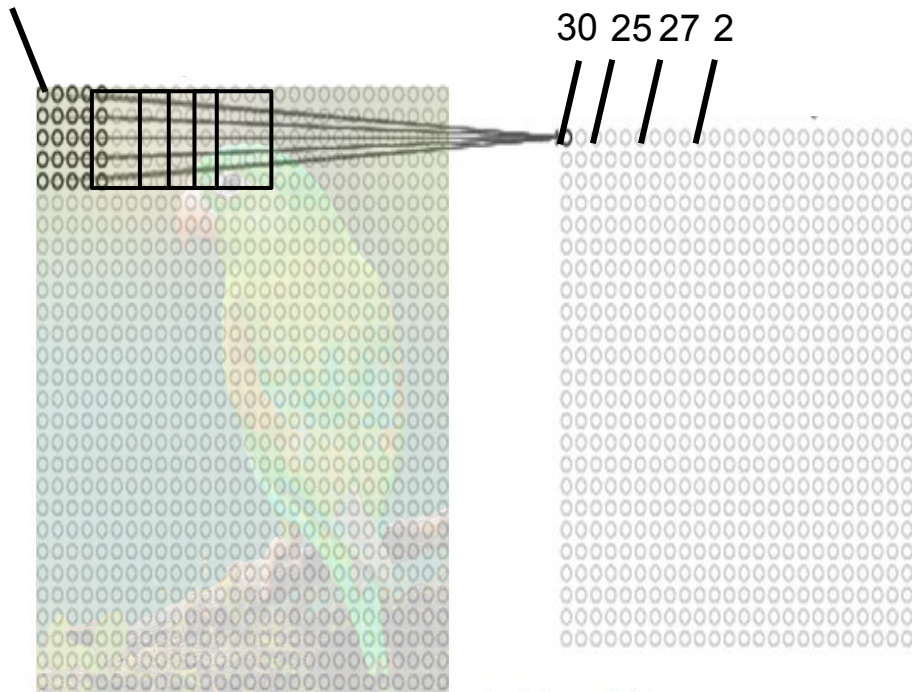
# Image Encoding



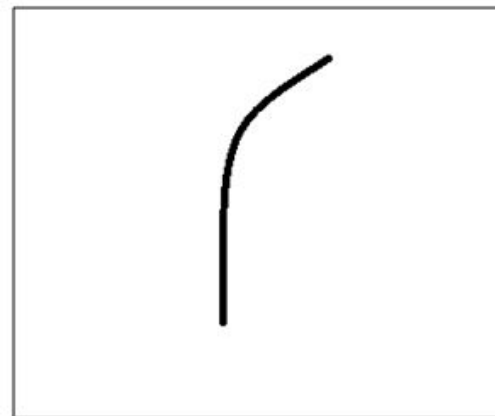
0	0	0	0	0	30	0
0	0	0	0	30	0	0
0	0	0	30	0	0	0
0	0	0	30	0	0	0
0	0	0	30	0	0	0
0	0	0	30	0	0	0
0	0	0	0	0	0	0

# Image Encoding

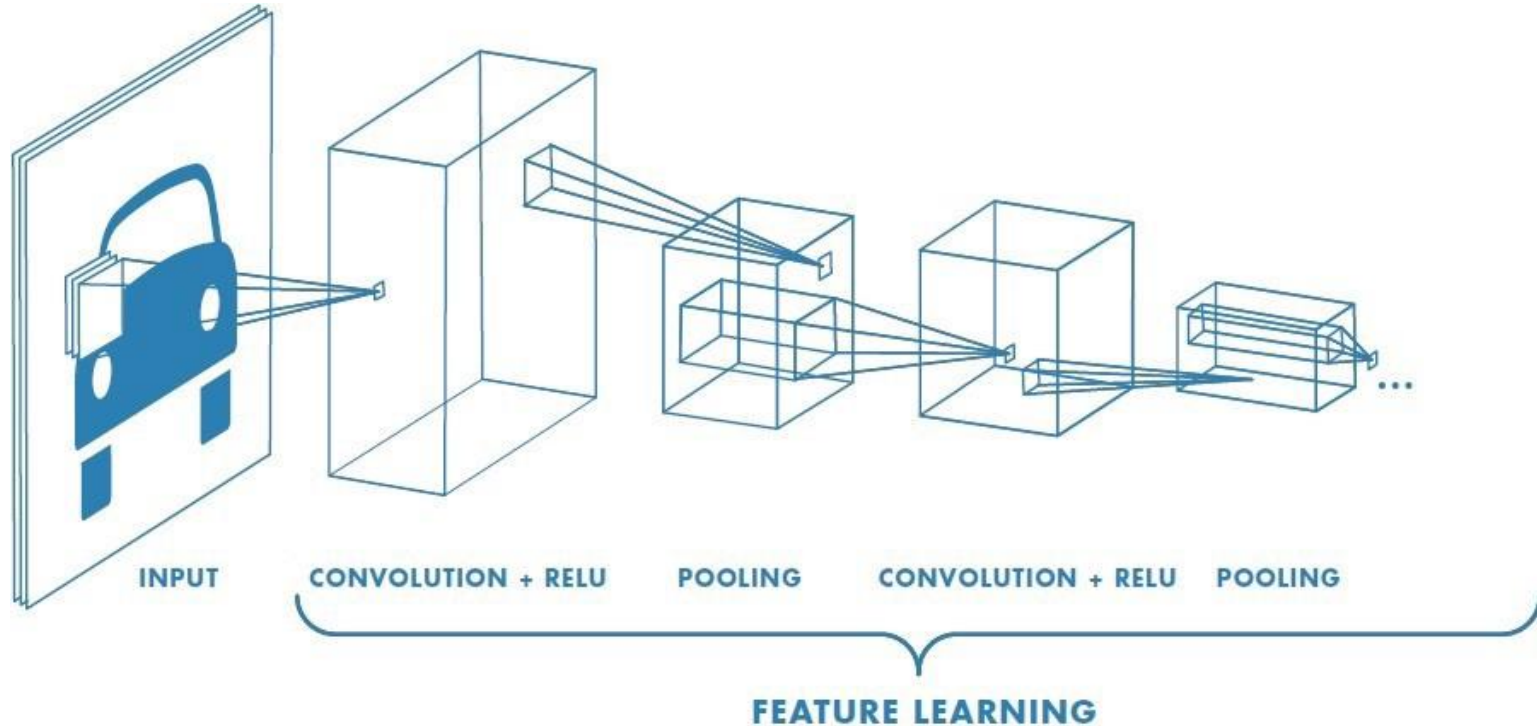
(250, 24, 170)



0	0	0	0	0	30	0
0	0	0	0	30	0	0
0	0	0	30	0	0	0
0	0	0	30	0	0	0
0	0	0	30	0	0	0
0	0	0	30	0	0	0
0	0	0	0	0	0	0



# Convolutional Neural Network



# Intuition: modifying image with text

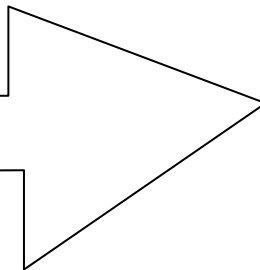


A yellow bird with grey wings and a black crown

Black crown

Grey wings

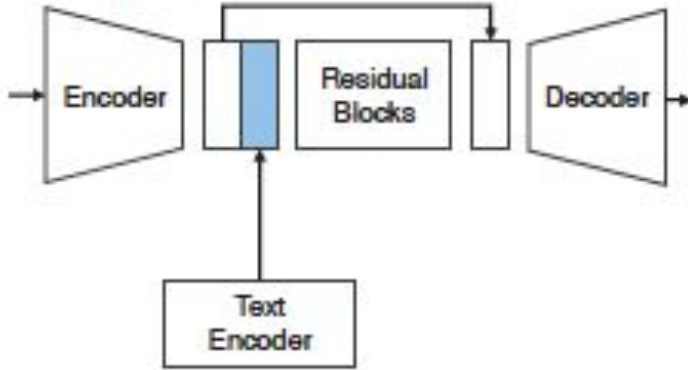
Yellow



A bird with a mix of black and white

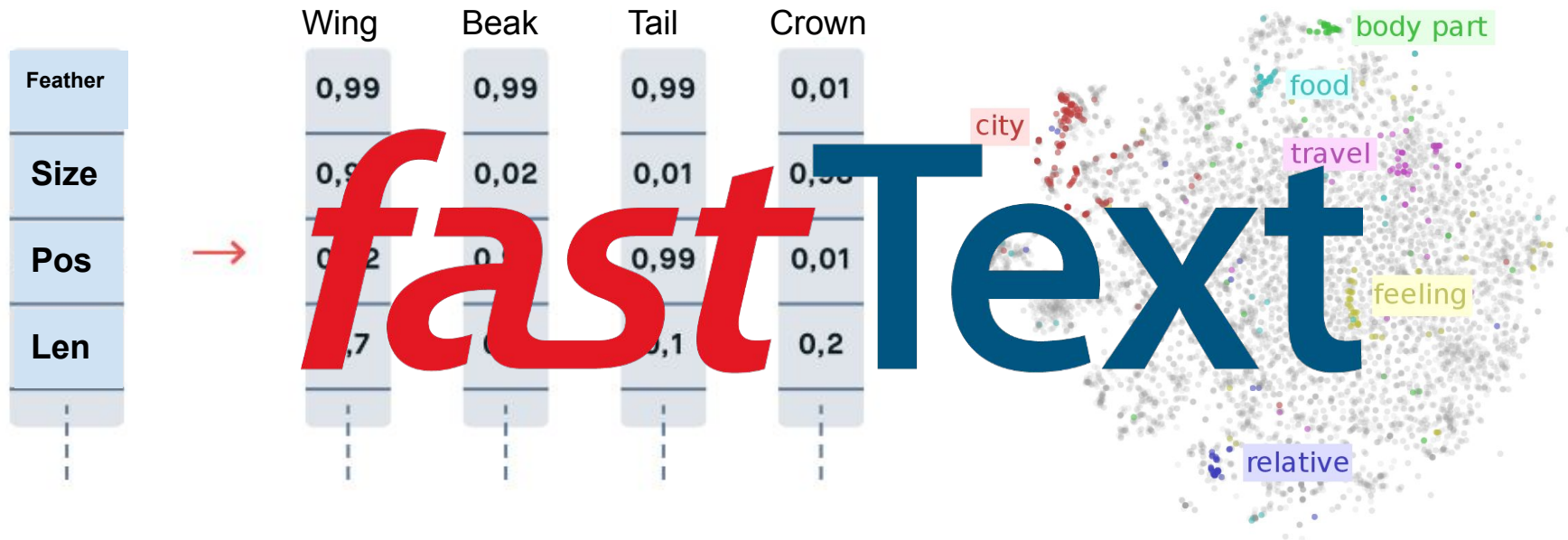


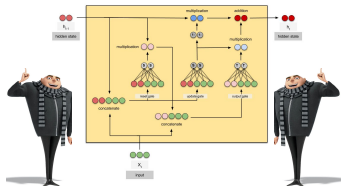
Generator



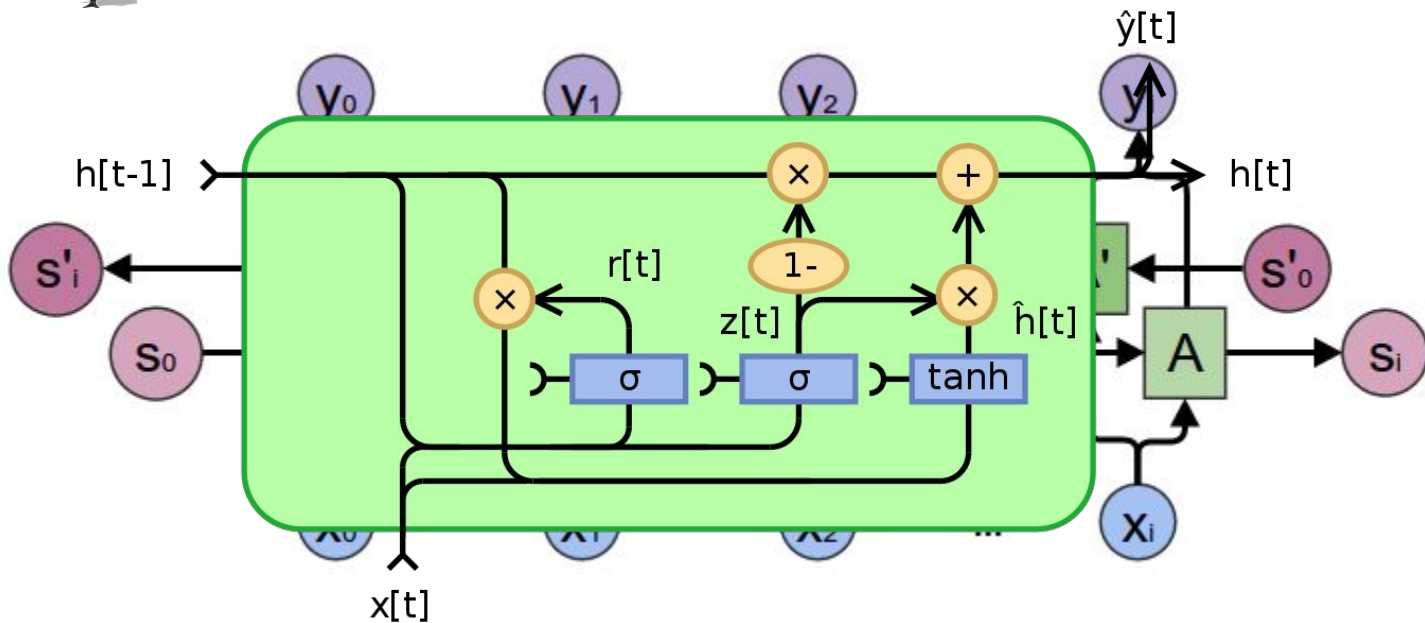
The bird is a mix of black and white..

# Text Encoding

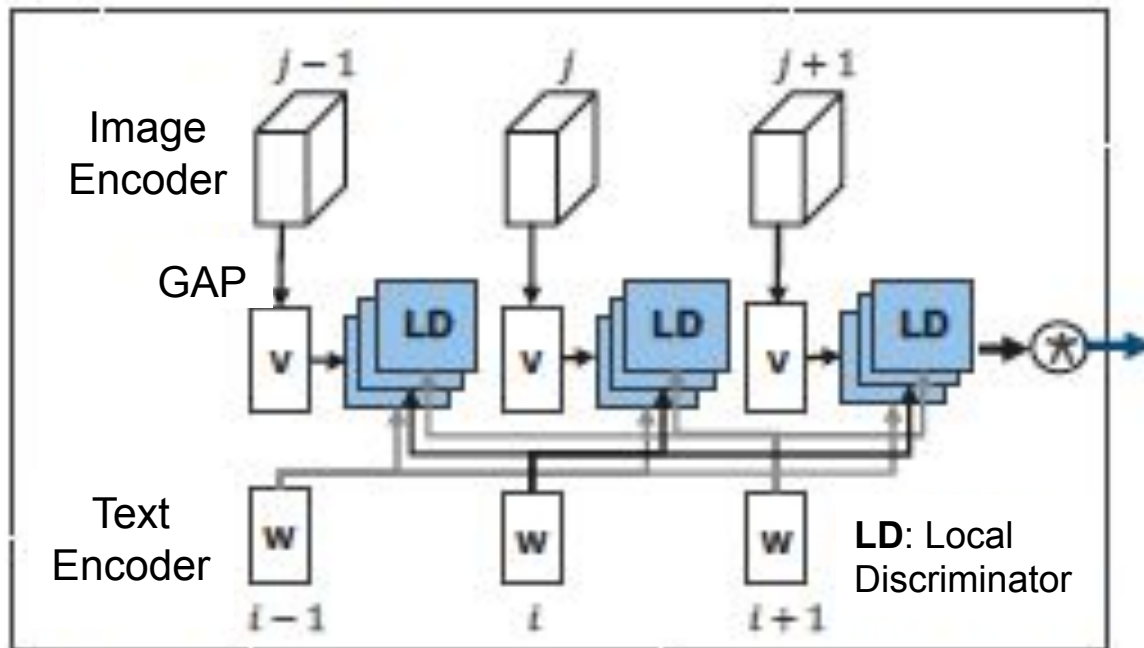




# Bidirectional GRU



# Discriminator



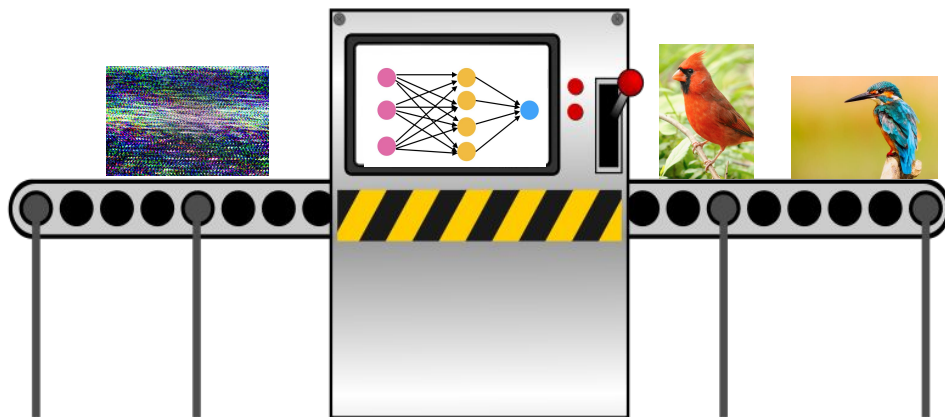


# Loss

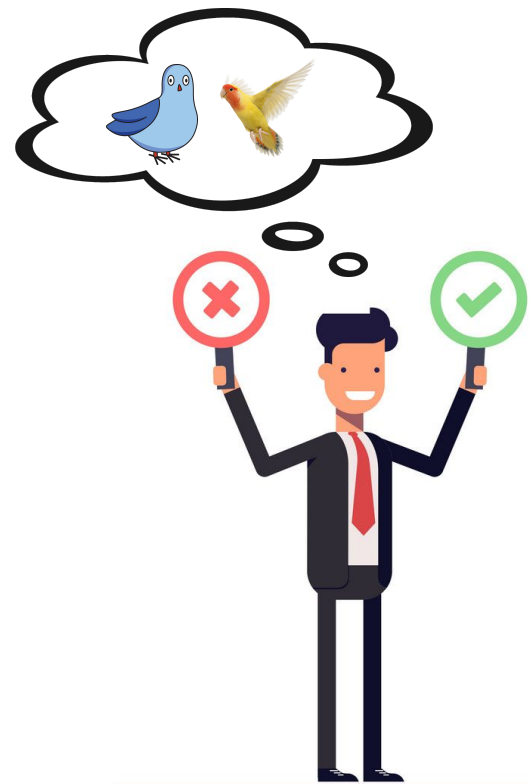
- Reconstruction loss
- Unconditional loss
- Conditional loss

# Recap

What's a GAN?



Generator



Discriminator

# Recap

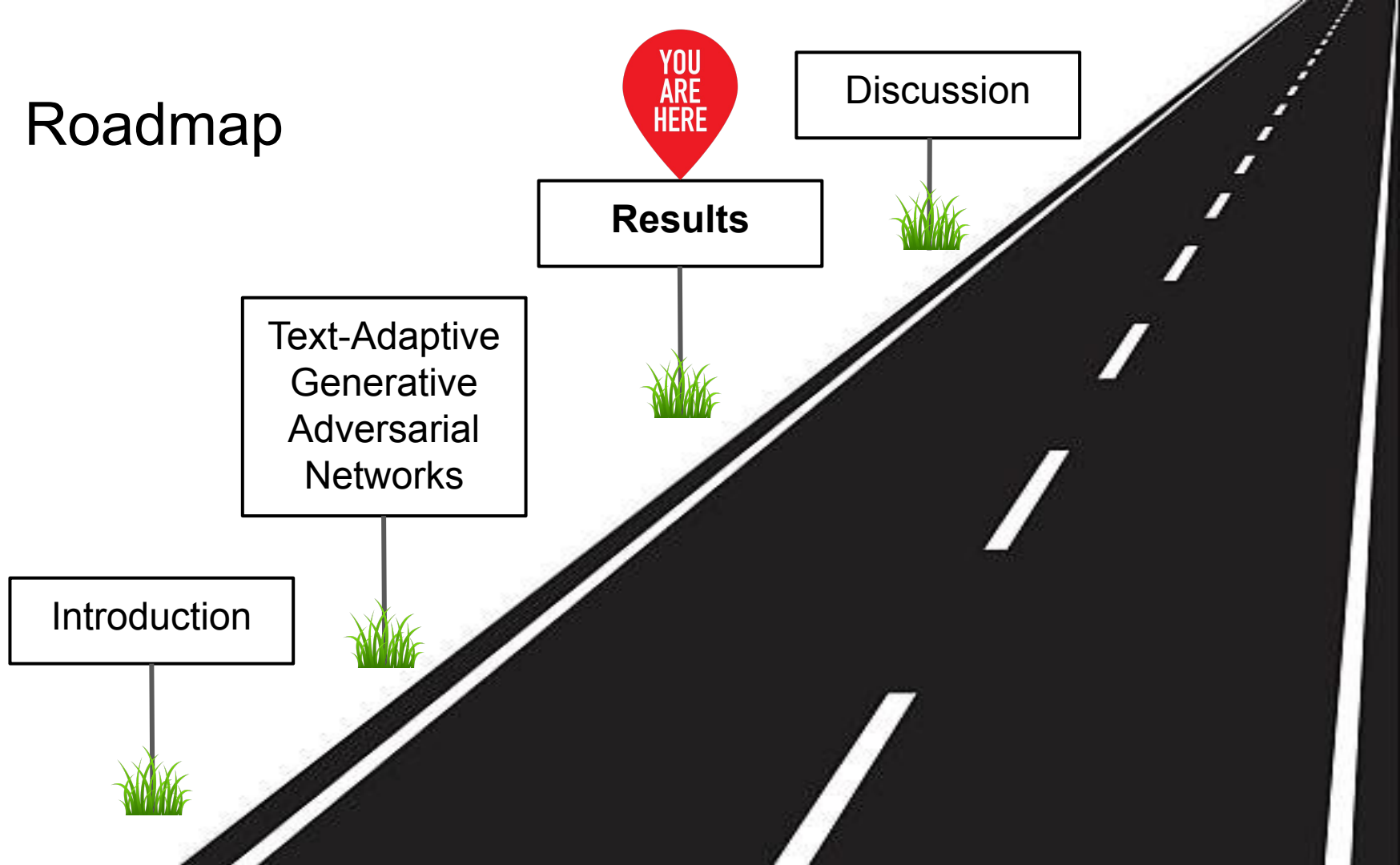
The problem?

- We want to manipulate images based on natural language descriptions

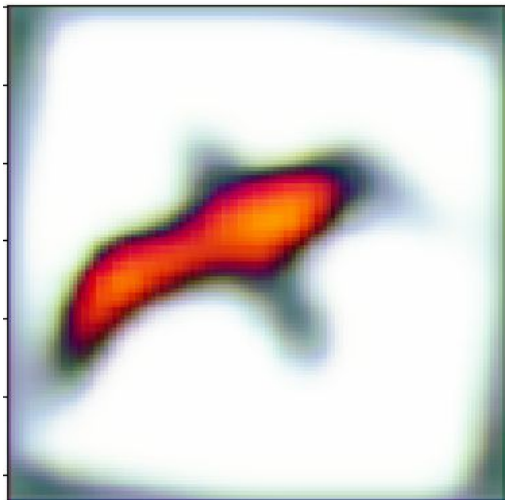
How does TAGAN differ?

- **Key difference:** the text-adaptive discriminator creates word-level local discriminators.

# Roadmap



# Results (Generator)



5 epochs



20 epochs

# Results (Discriminator pt1.)

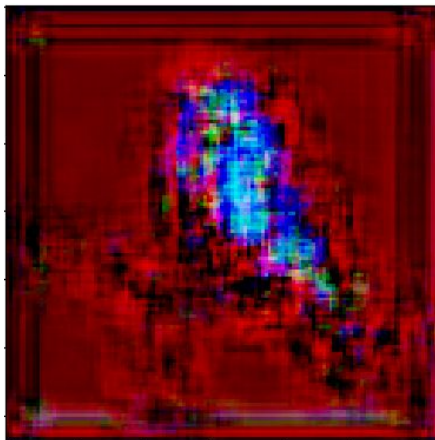
Generator with no text



Network with small dataset



Epoch 1



Epoch 6





Epoch 10



Epoch 20



# 30 Epochs



Real Image

**Caption:** this particular bird has a **belly that is white and has black spots**



Fake Image

# TAGAN Qualitative Results

**TAGAN Results:**

Original



The bird has **wings that are blue** and has a **white belly**.



**Our Results:**

Original



The small bird is **yellow and brown speckled** with a **pointed beak**



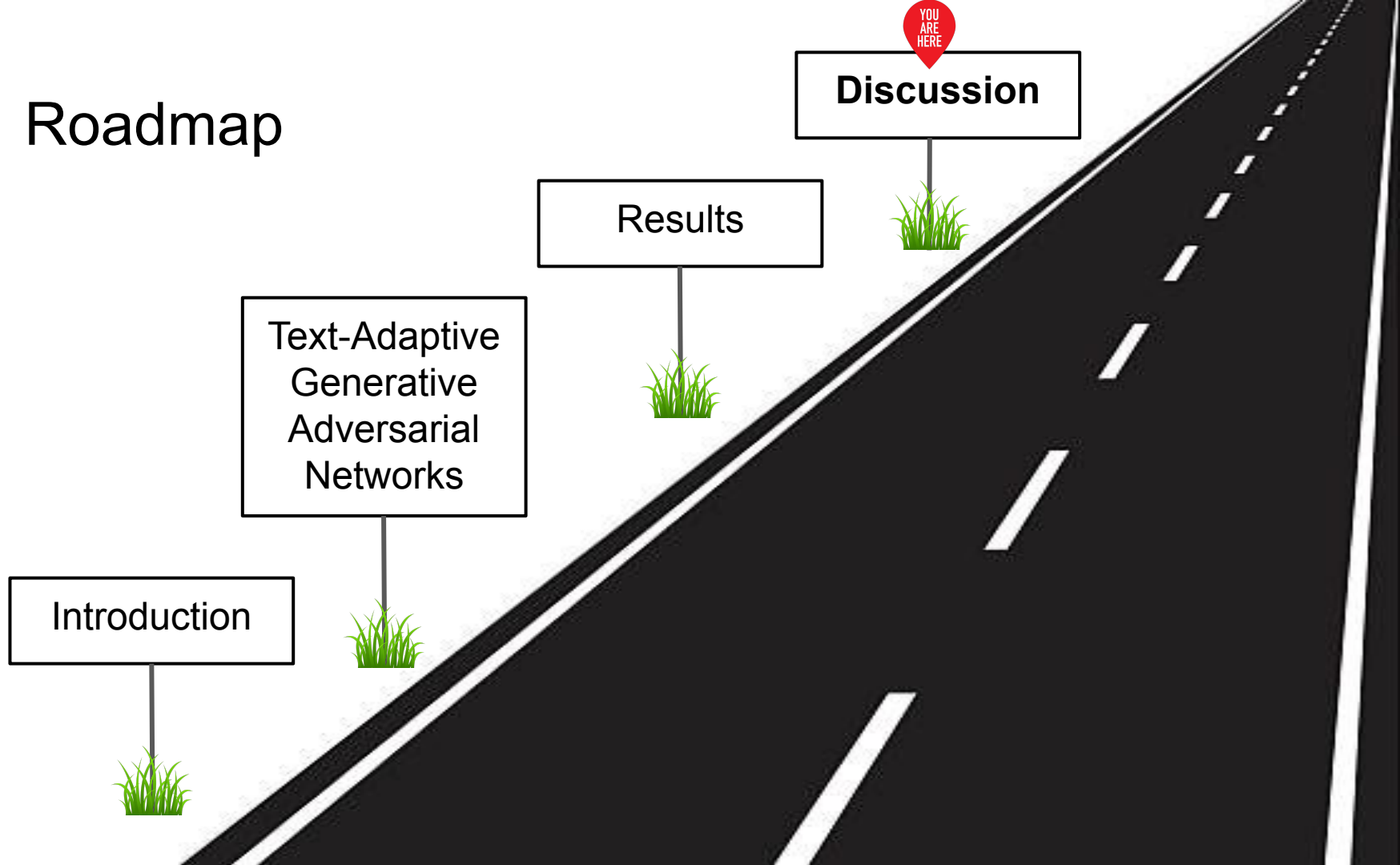
# TAGAN Quantitative Results

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Method	Accuracy	Naturalness	$L_2$ Error
SISGAN	2.33	2.34	0.30
AttnGAN	2.19	2.11	0.25
TAGAN	1.49	1.56	0.11
Ours	N/A	N/A	?

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# Roadmap



# Discussion

- Why are our results different?
- Can we make them not different?
- If we made them not different, what would we then make different?

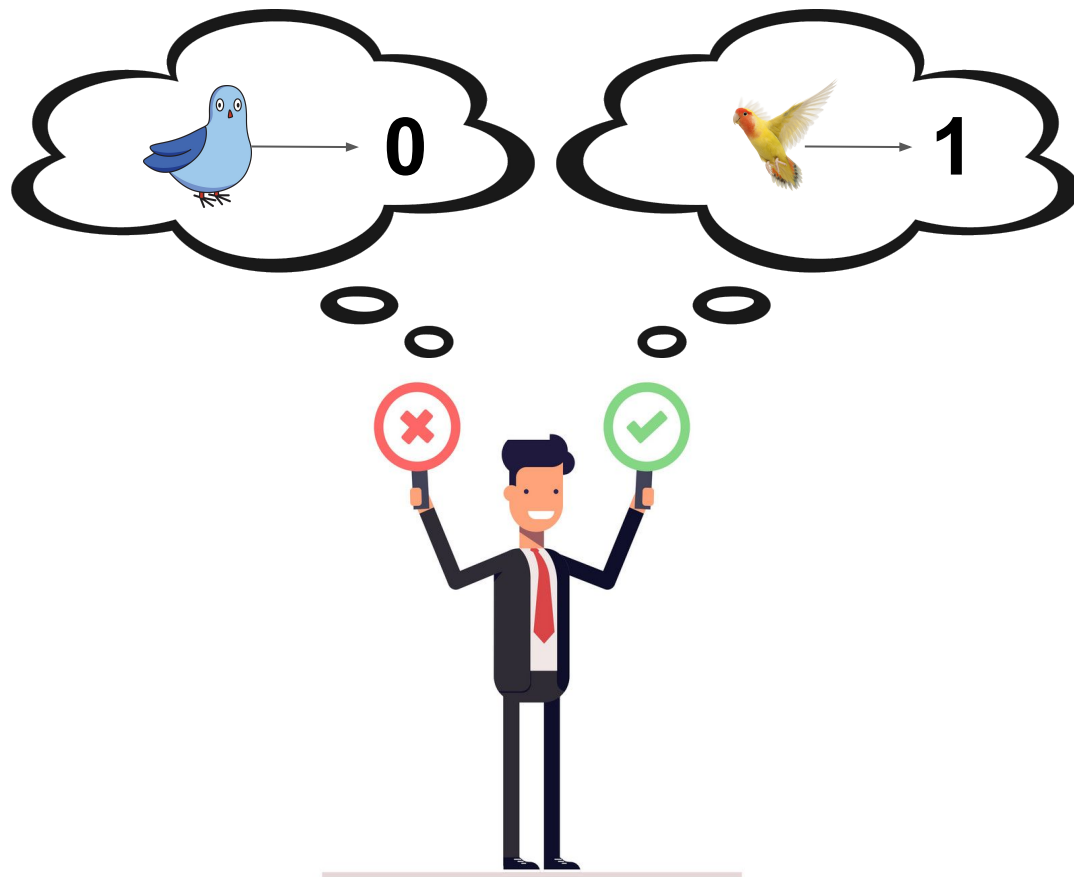
# Discussion

- **Why are our results different?**
- Can we make them not different?
- If we made them not different, what would we then make different?

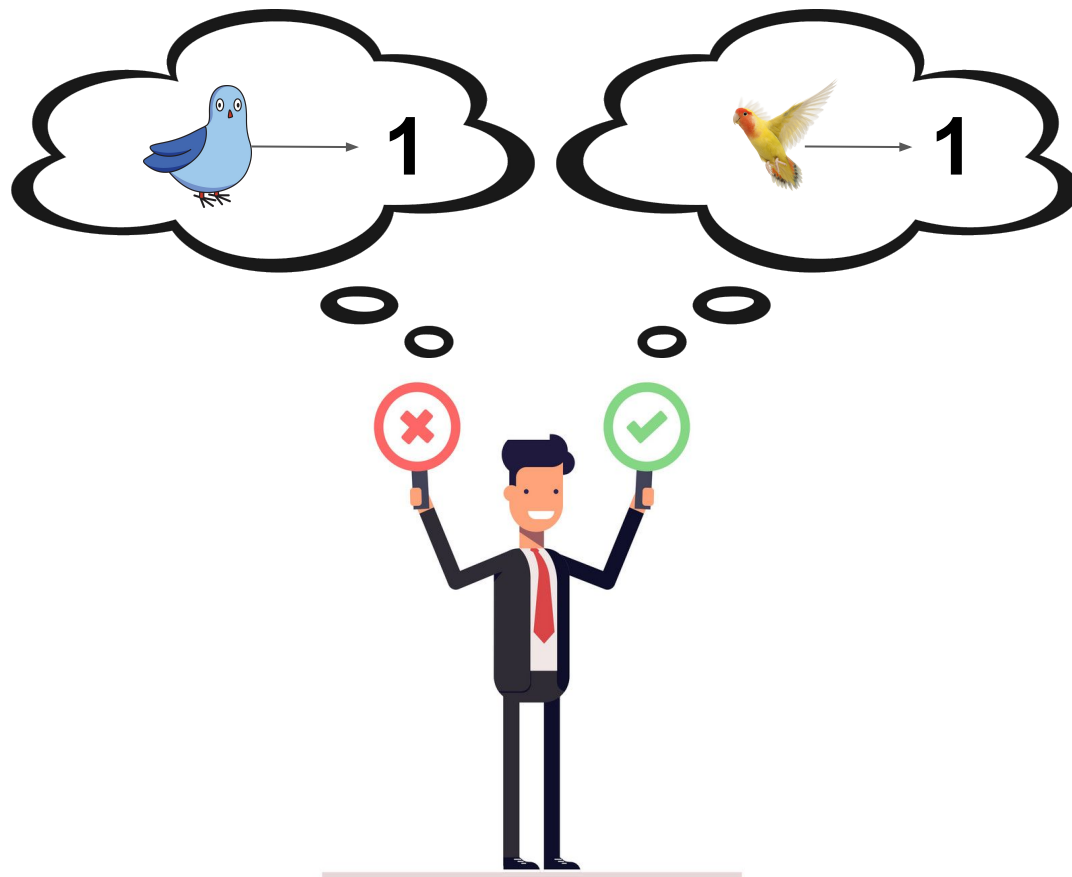


Q: Why are our results different?

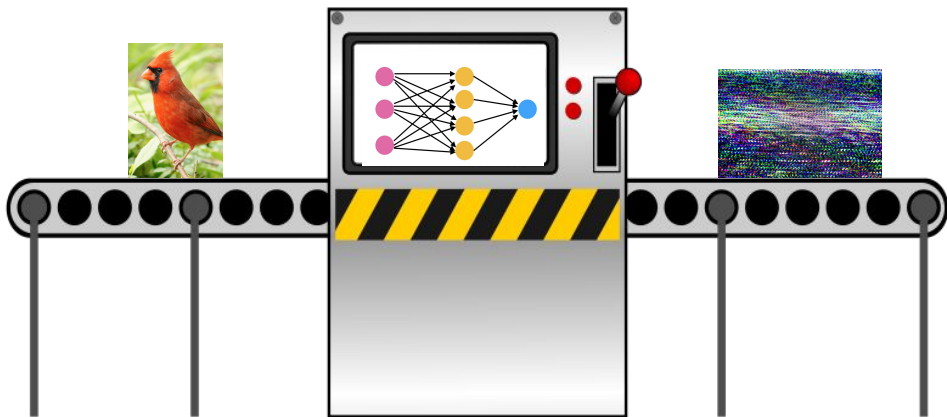
A: **Unstable training**



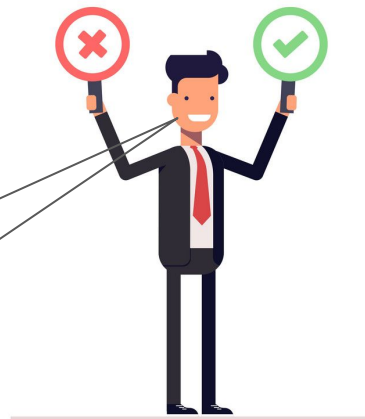
Discriminator



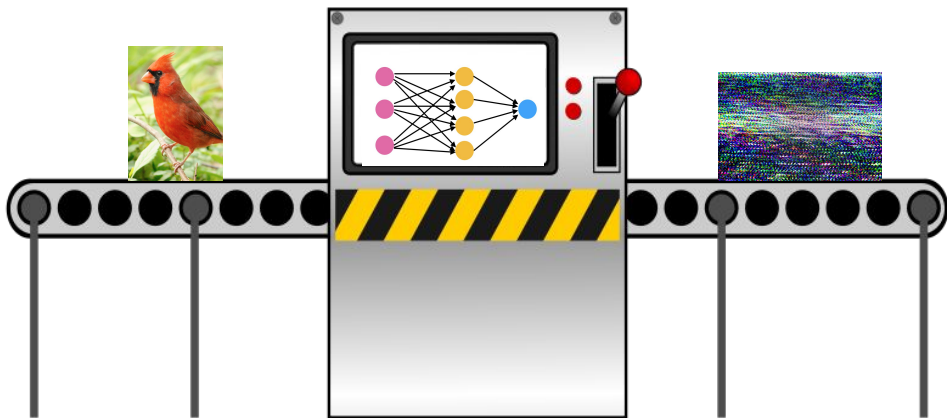
Discriminator



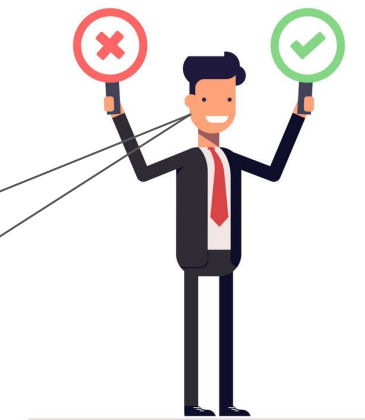
10/10  
wow  
great bird



“this particular bird has a belly that  
is white and has black spots”



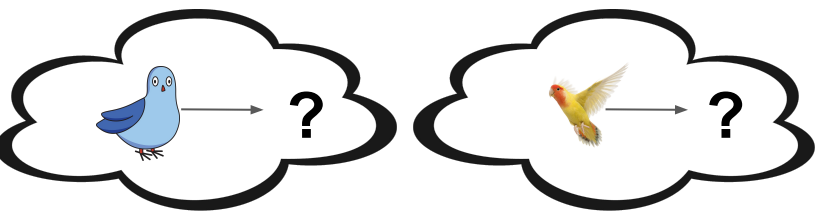
10/10 yes  
precisely  
correct



# Discussion

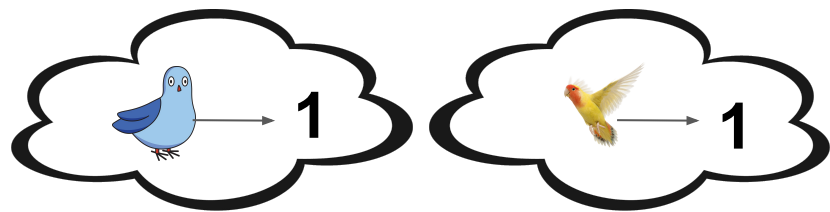
- Why are our results different? *A: Unstable training*
- **Can we make them not different?**
- If we made them not different, what would we then make different?

# Could it be our problem?



(Buggy)  
Discriminator

Training



(Buggy)  
Discriminator

# Could it be the paper's problem?

Module	Layers	Input size	Output size
Image Encoder	Conv2d(4, 2), LeakyReLU(0.2)	$3 \times 128 \times 128$	$64 \times 64 \times 64$
	Conv2d(4, 2), BN, LeakyReLU(0.2)	$64 \times 64 \times 64$	$128 \times 32 \times 32$
conv3	Conv2d(4, 2), BN, LeakyReLU(0.2)	$128 \times 32 \times 32$	$256 \times 16 \times 16$
conv4	Conv2d(4, 2), BN, LeakyReLU(0.2)	$256 \times 16 \times 16$	$512 \times 8 \times 8$
conv5	Conv2d(4, 2), BN, LeakyReLU(0.2)	$512 \times 8 \times 8$	$512 \times 4 \times 4$
Unconditional Discriminator	Conv2d(4, 0), Softmax	$512 \times 4 \times 4$	$1 \times 1 \times 1$
Text Encoder	Bidirectional GRU	# of words $\times$ 300	# of words $\times$ 512
$\beta_{ij}$	Linear, Softmax	# of words $\times$ 512	# of words $\times$ 3
$\alpha_i$	See Eq. (3) in the paper	# of words $\times$ 512	# of words $\times$ 1
$f_{w_{i,j}}$	Linear (See Eq. (2) in the paper)	N/A	N/A
From conv3	Conv2d(3, 1), BN, LeakyReLU(0.2)	$256 \times 16 \times 16$	$256 \times 16 \times 16$
(a)	Global Average Pooling	$256 \times 16 \times 16$	$256 \times 1 \times 1$
From conv4	Conv2d(3, 1), BN, LeakyReLU(0.2)	$512 \times 8 \times 8$	$512 \times 8 \times 8$
(b)	Global Average Pooling	$512 \times 8 \times 8$	$512 \times 1 \times 1$
From conv5	Conv2d(3, 1), BN, LeakyReLU(0.2)	$512 \times 4 \times 4$	$512 \times 4 \times 4$
(c)	Global Average Pooling	$512 \times 4 \times 4$	$512 \times 1 \times 1$
Conditional Discriminator	See Eq. (5) in the paper with $(\alpha_i, \beta_{ij}, f_{w_{i,j}}, (a), (b), (c))$	N/A	$1 \times 1 \times 1$

# Could it be the model's problem?

Not only are GANs famous for training instability, the authors themselves mention having such problems:

- “Note that we do not penalize generated outputs using the conditional discriminator in Eq. (6) due to **instability of training**.”
- “We set  $\lambda_1$  and  $\lambda_2$  to 10 and 2 respectively considering both the visual quality and the **training stability**.”

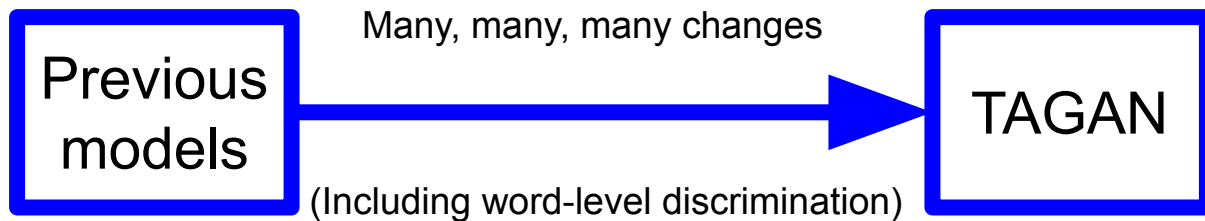




# Discussion

- Why are our results different? *A: Unstable training*
- Can we make them not different? *A: It depends, but probably*
- **If we made them not different, what would we then make different?**

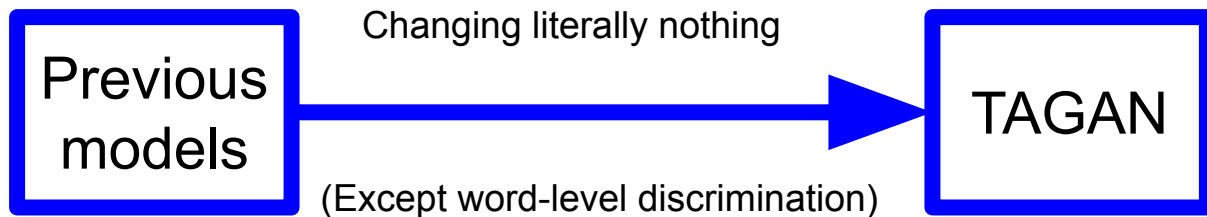
# Putting the science back in CS: the value of controls



The  
TAGAN  
authors

Word-level  
discrimination  
works!!!

# Putting the science back in CS: the value of controls



# Putting the science back in CS: the value of controls

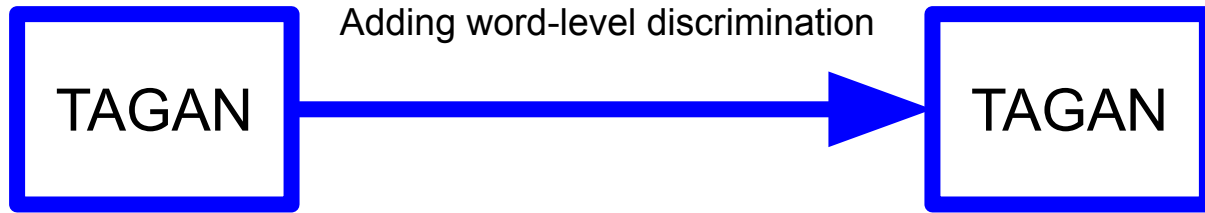


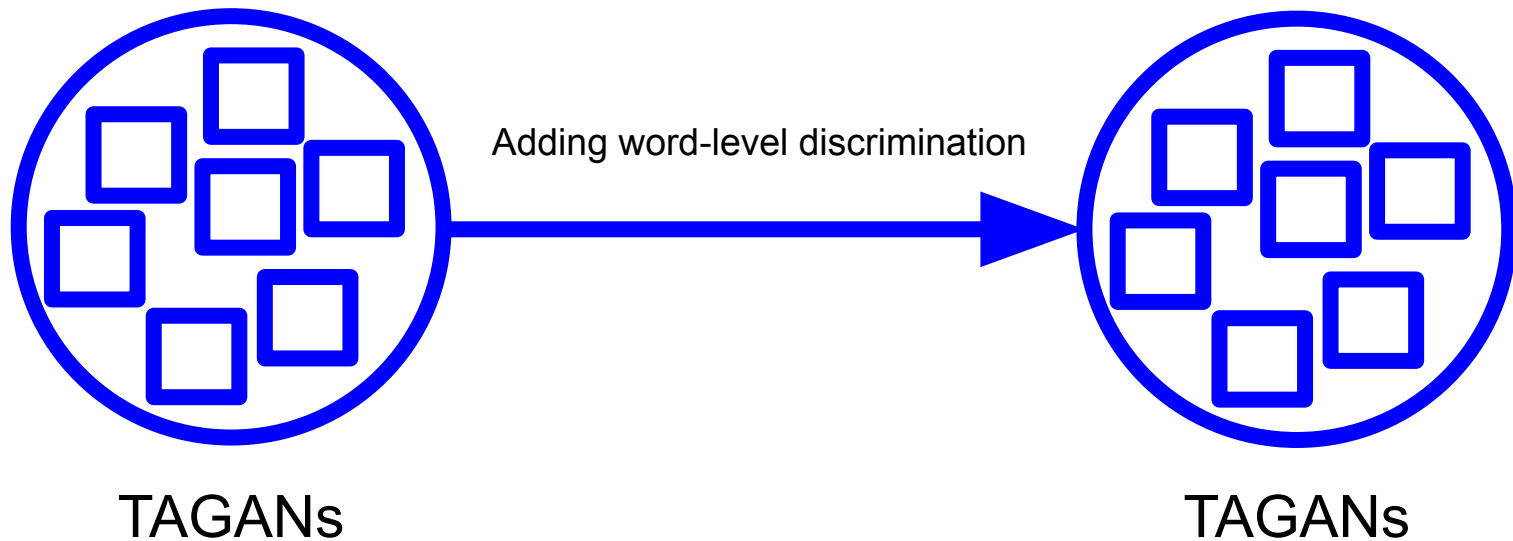
Image Encoder	Conv2d(4, 2), LeakyReLU(0.2)	$3 \times 128 \times 128$	$64 \times 64 \times 64$
	Conv2d(4, 2), BN, LeakyReLU(0.2)	$64 \times 64 \times 64$	$128 \times 32 \times 32$
conv3	Conv2d(4, 2), BN, LeakyReLU(0.2)	$128 \times 32 \times 32$	$256 \times 16 \times 16$

```

self.conv123 = nn.Sequential(
    nn.Conv2d(3, 64, 4, 2, padding=1, bias=False),
    nn.LeakyReLU(negative_slope=0.2, inplace=True),
    nn.Conv2d(64, 128, 4, 2, padding=1, bias=False),
    nn.BatchNorm2d(128),
    nn.LeakyReLU(negative_slope=0.2, inplace=True),
    nn.Conv2d(128, 256, 4, 2, padding=1, bias=False),
    nn.BatchNorm2d(256),
    nn.LeakyReLU(negative_slope=0.2, inplace=True)
)

```

# Putting the science back in CS: DIY replication



# Discussion

- Why are our results different? *A: Unstable training*
- Can we make them not different? *A: It depends, but probably*
- If we made them not different, what would we then make different? *A: Science*



# Acknowledgements



Seonghyeon Nam,  
Yunji Kim, and Seon  
Joo Kim

# Thank you!

Questions?

