## Deep Facial Animation Retargeting from Humans to 3D Stylized Characters

#### Facial Expression Research Group (FERG), UW CSE

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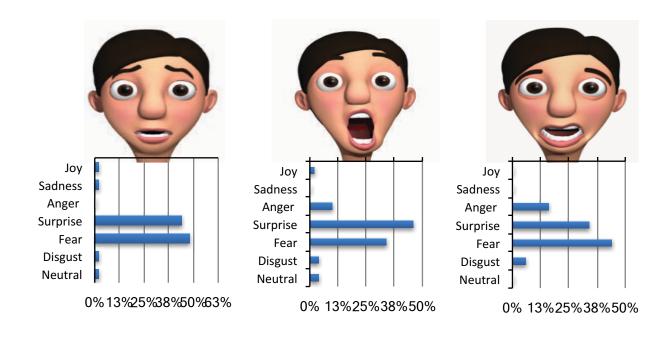


# Facial expressions: The art of non-verbal communication



## Creating recognizable expressions

- Accurate facial expression depiction is critical and difficult for storytelling.
- We asked professional animators to make this character look surprised. None of the expressions achieved greater than 50% recognition on Mechanical Turk.



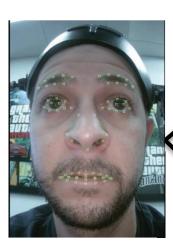
#### Introduction

- ➤ **Goal:** accurate and perceptually correct facial expression transfer
  - Capture human facial expression from 2D images
  - Transfer the expression to multiple 3D stylized characters

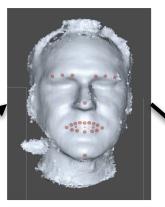
#### Related Work:

- Landmark mapping (manual labor intensive)
- Blendshape interpolation (memory intensive)
- We propose a generative method for expression transfer that ensures:
  - Perceptually validity by using deep neural network features
  - Geometric consistency by using geometric features

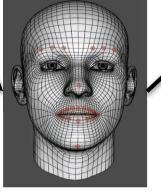
## Related work: Landmark Mapping



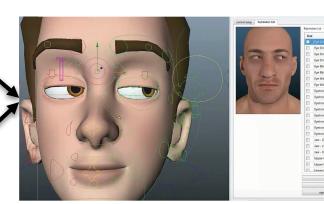
Human input followed by landmark detection



New 3D model generation

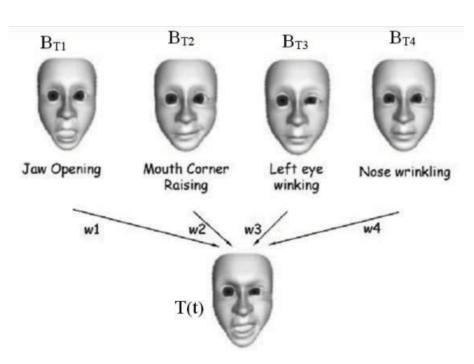


Existing 3D model morphing



Associate character rig controls to expression sets/landmark groups

### Related work: Blendshape Interpolation



Performance capture using blendshapes; Regression-based estimation of weights



$$\mathbf{S}(t) = \sum_{i=1}^{k} \omega_{i}(t).\,\mathbf{B}_{\mathrm{si}}$$

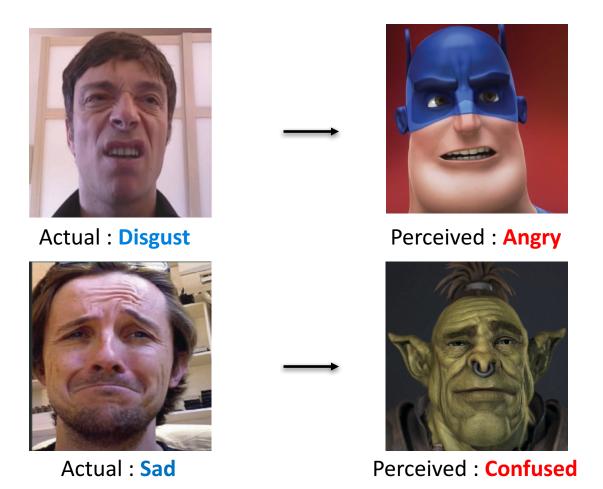




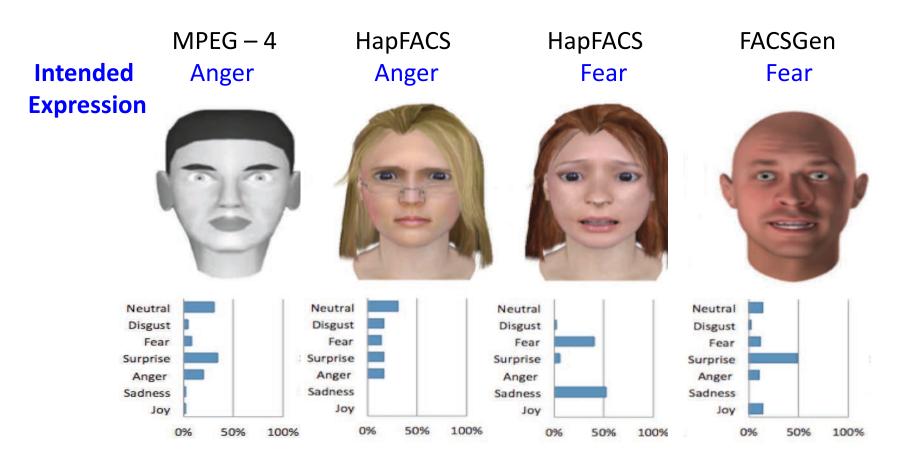
$$\mathbf{T}(t) = \sum_{i=1}^{k} \omega_{i}(t).\,\mathbf{B}_{\mathrm{T}}$$

## Drawbacks of existing methods

Geometric features transfer, expressions often do not!



## Drawbacks of existing methods

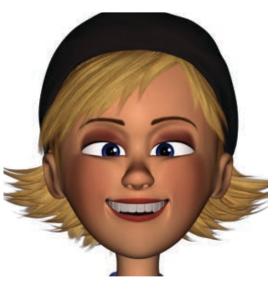


MPEG-4: Pereira, F.C., Ebrahimi, T.: The MPEG-4 Book. Prentice Hall PTR, Upper Saddle River, NJ, USA (2002)
HapFACS: Amini, R., Lisetti, C.: HapFACS: an open source API/Software to generate FACS- Based expressions for ECAs animation (ACII). (2013) 270–275
FACSGen: Roesch, E.B., Tamarit, L., Reveret, L., Grandjean, D., Sander, D., Scherer, K.R.: FACSGen: a tool to synthesize emotional facial expressions through systematic manipulation of facial action units. Journal of Nonverbal Behavior (2011) 1–16







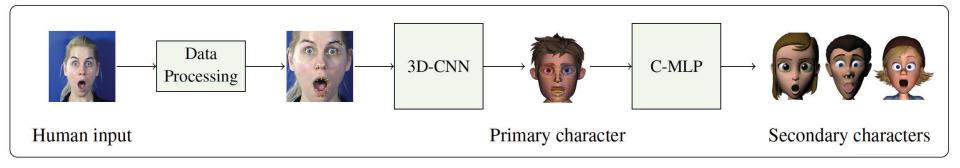




#### **Contributions**

- A novel perceptually valid method to map 2D human face images to 3D stylized character rig controls.
- The ability to utilize this mapping to generate 3D characters with clear unambiguous facial expressions.
- A semi-supervised method to enable expression transfer between multiple characters.

## Our Approach



**3D-CNN**: Deep convolutional neural network for human to character transfer

**C-MLP**: Lightweight multi-layer perceptrons for character to character transfer *Advantage*: No retraining of 3D-CNN required for a new character

Challenge: No existing dataset of matching 2D image – 3D character model pairs

## **Dataset and Preprocessing**

Expression classes: Anger, Disgust, Fear, Neutral, Joy, Sadness, Surprise

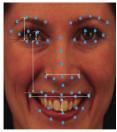






SFEW, CK+, MMI, KDEF and DISFA ~100k images





Bounding box and landmark detection; frontalization and geometry extraction







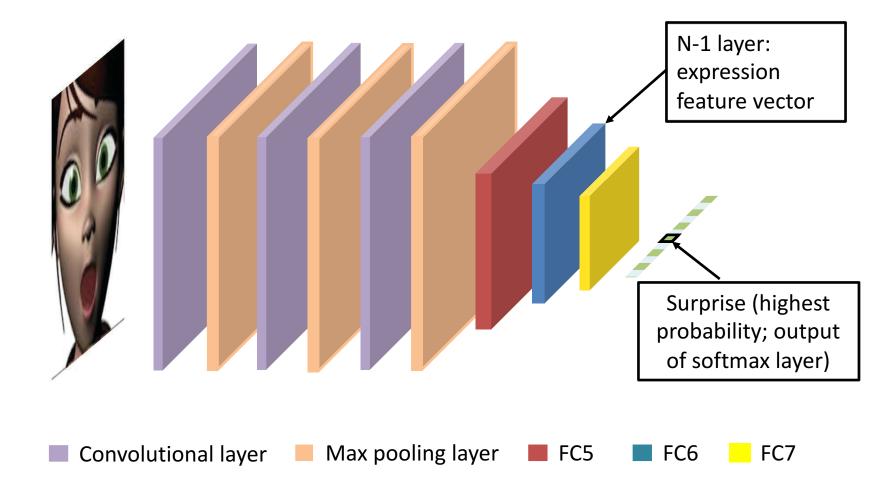
FERG-DB + 3 new characters ~55K images





3D rig parameters; geometry extraction

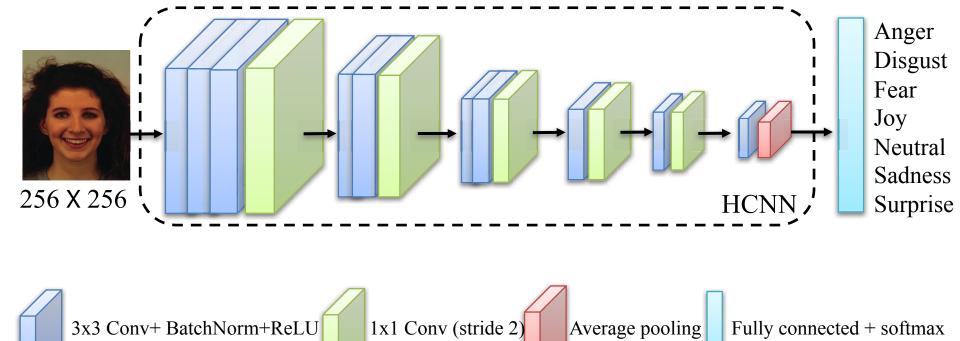
## **CNN** for expression recognition



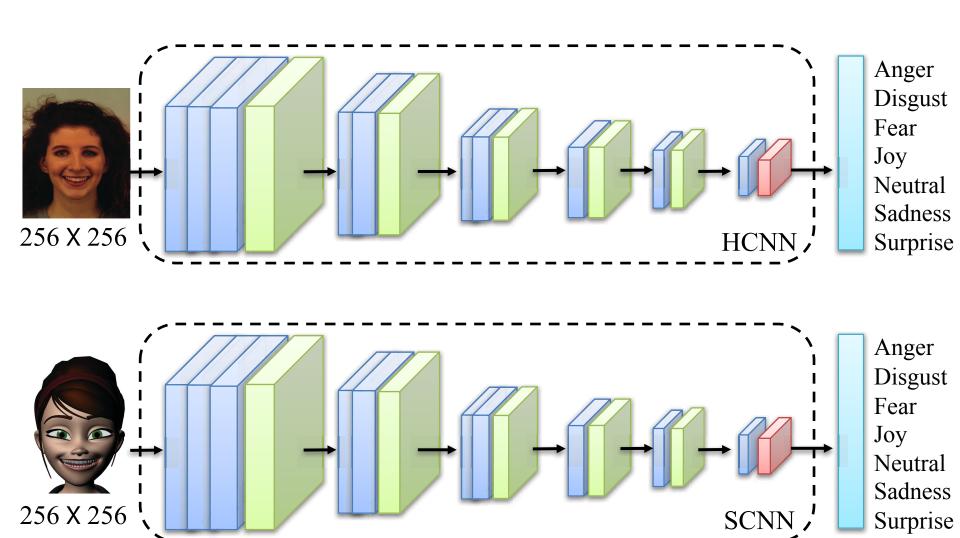
#### Different activation functions in CNN

Name	Plot	Equation	Derivative		
Identity		f(x) = x	f'(x) = 1		
Binary step		$f(x) = \begin{cases} 0 & \text{for } x < 0 \\ 1 & \text{for } x \ge 0 \end{cases}$	$f'(x) = \begin{cases} 0 & \text{for } x \neq 0 \\ ? & \text{for } x = 0 \end{cases}$		
Logistic (a.k.a Soft step)		$f(x) = \frac{1}{1 + e^{-x}}$	f'(x) = f(x)(1 - f(x))		
TanH		$f(x) = \tanh(x) = \frac{2}{1 + e^{-2x}} - 1$	$f'(x) = 1 - f(x)^2$		
ArcTan		$f(x) = \tan^{-1}(x)$	$f'(x) = \frac{1}{x^2 + 1}$		
Rectified Linear Unit (ReLU)		$f(x) = \begin{cases} 0 & \text{for } x < 0 \\ x & \text{for } x \ge 0 \end{cases}$	$f'(x) = \begin{cases} 0 & \text{for } x < 0 \\ 1 & \text{for } x \ge 0 \end{cases}$		
Parameteric Rectified Linear Unit (PReLU) <sup>[2]</sup>		$f(x) = \begin{cases} \alpha x & \text{for } x < 0 \\ x & \text{for } x \ge 0 \end{cases}$	$f'(x) = \begin{cases} \alpha & \text{for } x < 0 \\ 1 & \text{for } x \ge 0 \end{cases}$		
Exponential Linear Unit (ELU) <sup>[3]</sup>		$f(x) = \begin{cases} \alpha(e^x - 1) & \text{for } x < 0 \\ x & \text{for } x \ge 0 \end{cases}$	$f'(x) = \begin{cases} f(x) + \alpha & \text{for } x < 0 \\ 1 & \text{for } x \ge 0 \end{cases}$		
SoftPlus		$f(x) = \log_e(1 + e^x)$	$f'(x) = \frac{1}{1 + e^{-x}}$		

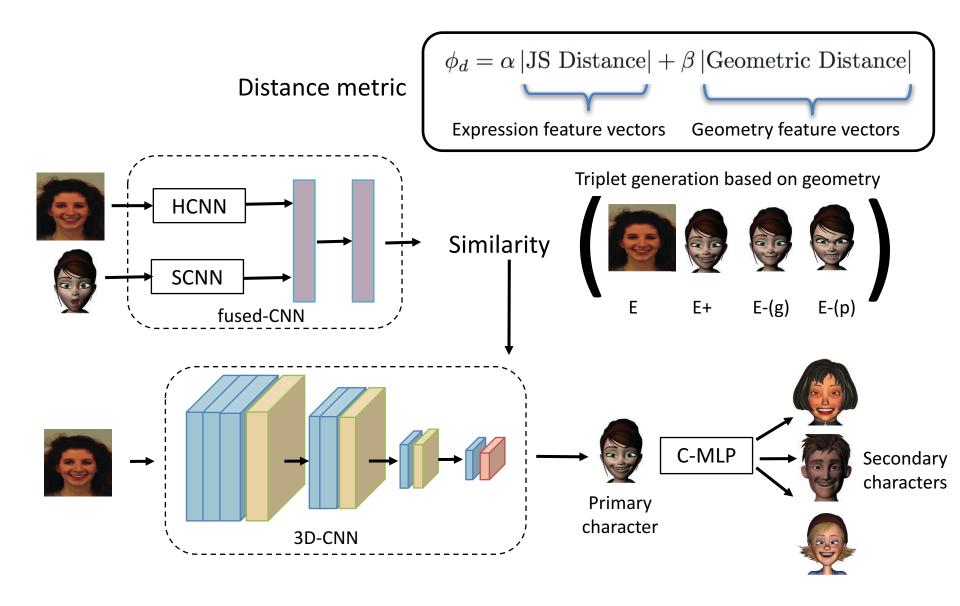
#### **Network Architecture**



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## **Network Architecture**



# Results **Human input - Character output**

**Expression: Sadness** 







Expression: Anger



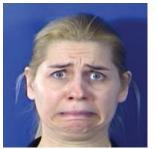


Expression: Surprise





Expression: Fear





Expression: Disgust



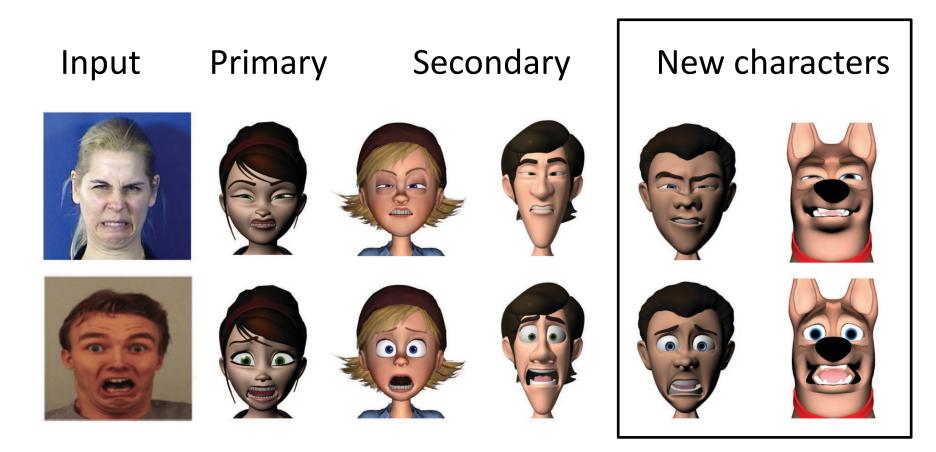


Expression: Joy

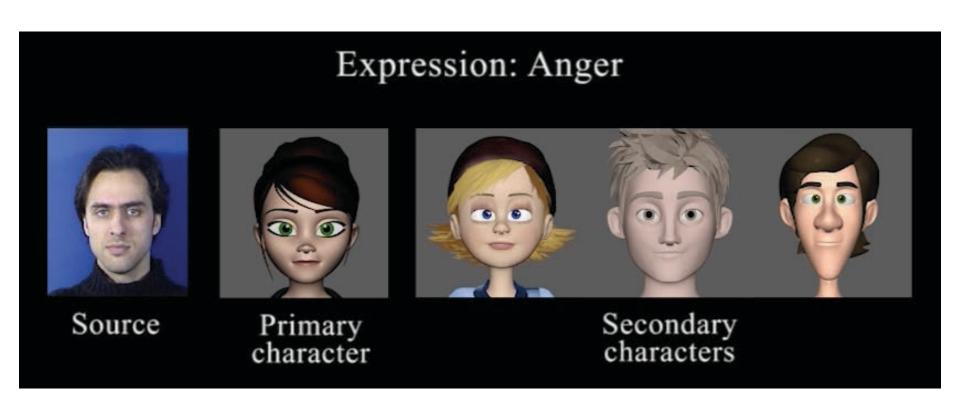




## Single Human to Multiple Characters



## Results: Video frame sequences

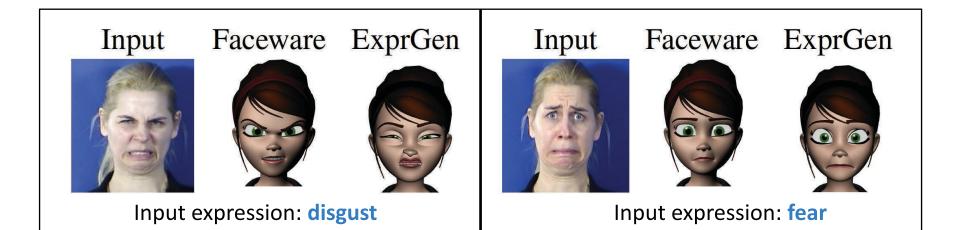


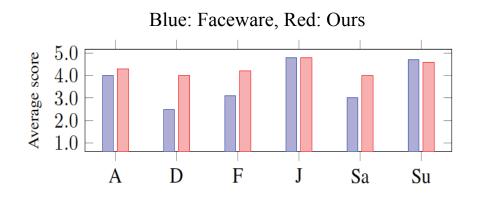
### **Evaluation**

		Perceived character expression (%)								
			Α	D	F	J	N	Sa	Su	
Perceived human	expression (%)	Α	71.32	16.28	5.43	1.55	3.10	0.78	1.55	
		D	14.29	67.35	4.08	1.02	4.08	8.16	1.02	
		F	2.88	6.47	64.03	2.16	3.60	3.60	17.27	
		J	0.92	1.83	0.92	90.83	1.83	0.92	2.75	
		N	1.09	3.26	2.17	4.35	76.09	10.87	2.17	
	ex	Sa	1.80	3.60	2.70	1.80	18.02	71.17	0.90	
		Su	0.52	1.04	7.77	1.55	0.52	0.52	88.08	

Confusion matrix of expression recognition for 1000 test cases with 30 MT subjects

#### **Evaluation**

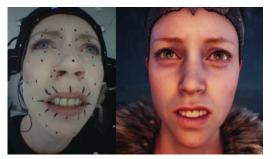




Quantitative comparison of expression transfer results Faceware (blue bars) and ExprGen (red bars).

## **Applications**

Efficient visual storytelling:







video games

motion capture films

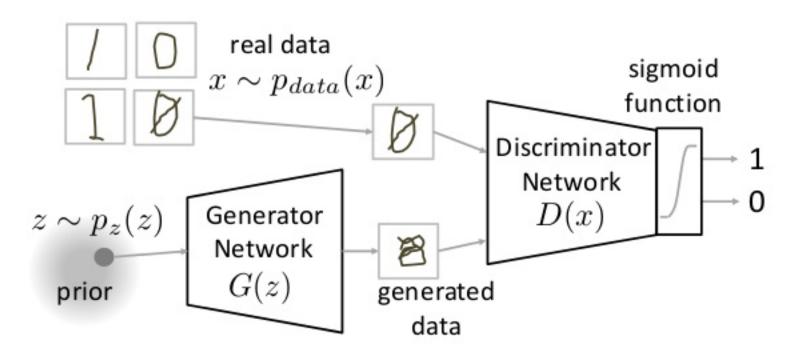
social VR experience

- Human-computer or human-robot interaction
- Helping autistic kids to recognize and convey emotions

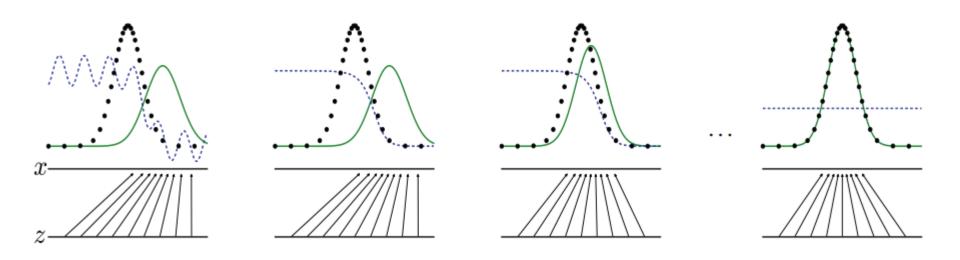
#### Generative Adversarial Networks

$$\min_{G} \max_{D} V(D,G)$$

$$V(D,G) = \mathbb{E}_{x \sim p_{data}(x)}[\log D(x)] + \mathbb{E}_{z \sim p_z(z)}[\log(1 - D(G(z)))]$$



## **GAN** training



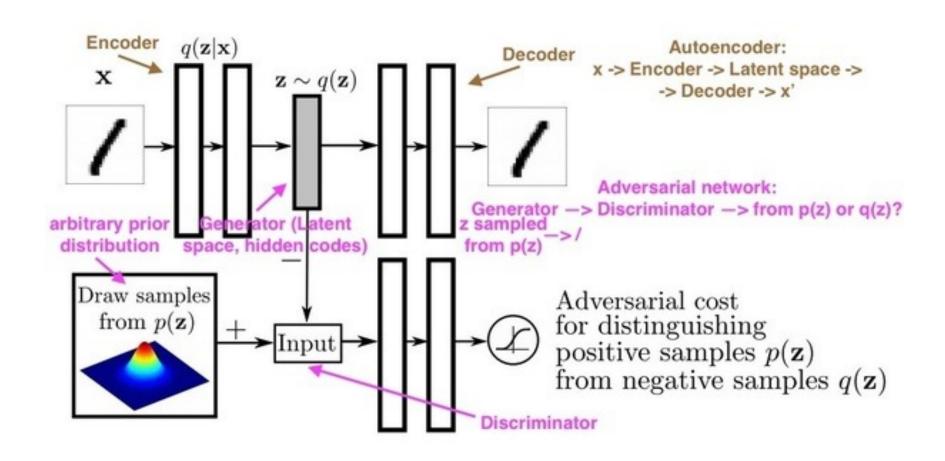
x: actual distribution of training data (black dotted line)

z: random noise input

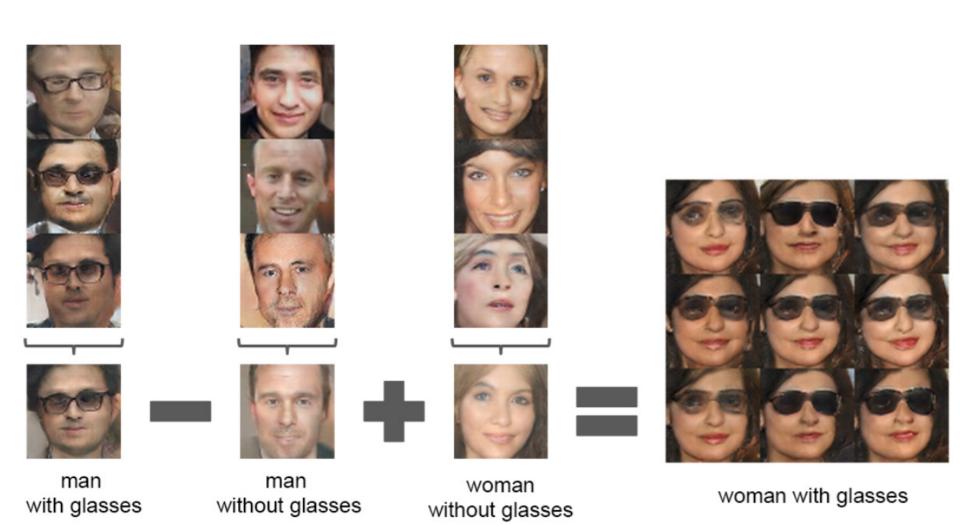
green solid line: distribution of generated samples

blue dotted line: discriminator loss

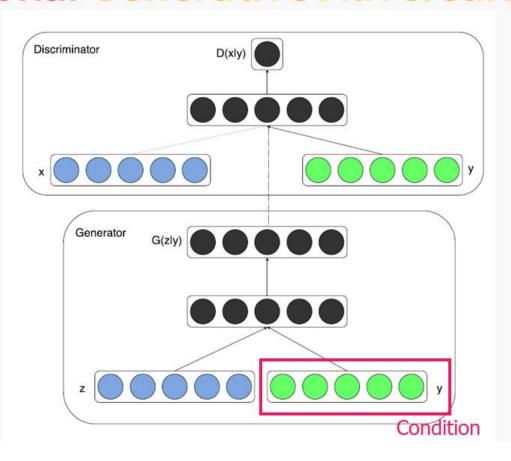
## GAN alternative; use of Autoencoder



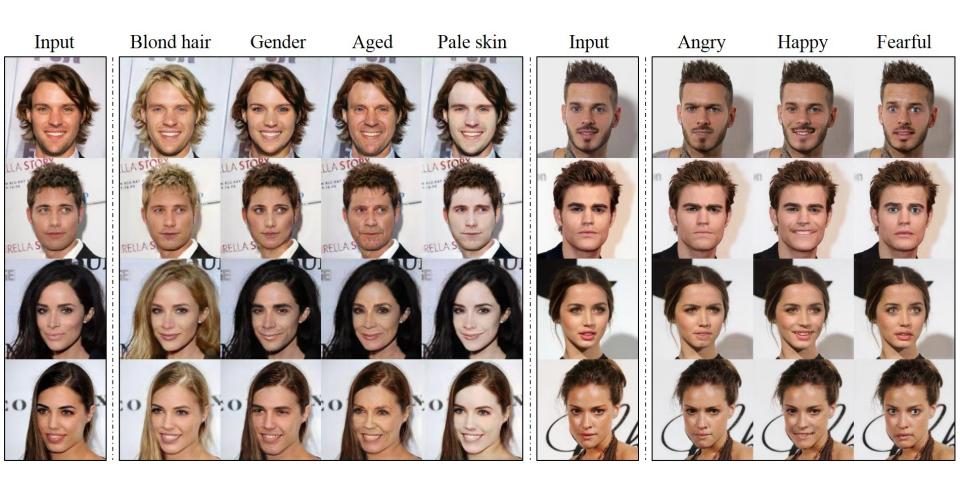
#### Vector arithmetic with latent space in GAN



#### **Conditional Generative Adversarial Nets**



#### Outputs of GAN conditioned on attributes



### References

- [Our ACCV'16 paper]: http://grail.cs.washington.edu/projects/deepexpr/deepali\_accv2016.pdf
- [Our WACV'18 paper]: https://homes.cs.washington.edu/~bindita/2Dto3Dexpr\_WACV.pdf
- **♦** [Faceware]: <a href="http://facewaretech.com/">http://facewaretech.com/</a>; geometry mapping technique based software
- [Blendshape interpolation]:

  https://www.researchgate.net/publication/320472350\_Semi\_Automatic\_Retargeting\_for\_Facial\_Ex

  pressions\_of\_3D\_Characters\_with\_Fuzzy\_logic\_Based\_on\_Blendshape\_Interpolation
- [List of GAN papers]: https://github.com/nightrome/really-awesome-gan
- ❖ [Facial Animation survey]: http://freesouls.github.io/2015/04/16/3d-facial-animation/index.html
- [Details about CNNs]: http://cs231n.github.io/convolutional-networks/

# Thank you!













**Questions?**