# Differential Privacy Applied to Facial Recognition



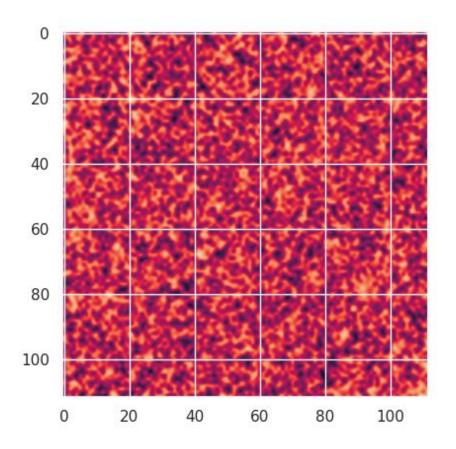


Source: DALL-E

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### Who is this?









- Homomorphic encryption
- Discrete Cosine Transform
- Learnable privacy budgets our motivation



#### **Datasets**



Source: VGG-Face2: A dataset for recognising faces across pose and age.







Differential Privacy Perturbation Module



**Utility Check Module** 



## Frequency Domain Transformation Module

- Preserve visual privacy
- Humans rely on low-frequency information for image recognition
- Neural Networks rely on low and high frequency information for IR
- Using DCT to identify and alter human-readable aspects of images





- Transforms signals from one representation to another
- Common use is in compression for JPEG

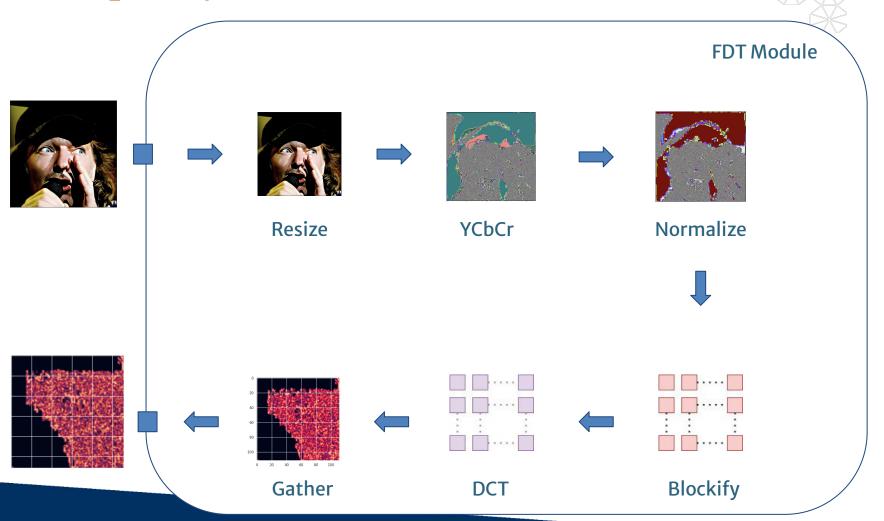
115.75	116.33	115.18	115.75	121.95	128.61		130.08
112.75	114.10	113.89	115.75	122.73	129.81	130.89	127.08
109.75	110.73	112.41	114.98	123.24	130.10	129.18	122.78
109.98	111.73	112.81	114.95	122.94	130.73		127.89
111.98	113.02	113.10	113.89	120.05	128.01		233.28
114.97	114.83	115.33	115.18	116.75	120.94	127.33	129.81
114.20	113.82	114.73	116.41	116.97	119.05	123.96	129.32
100.90	102.05	106.13	112.33	114.89	115.98	122.06	128.24



-70.25	-54.59	7.77	5.53	4.39	0.02	-0.33	0.54
15.14	-0.79	-3.70	10.23	-6.71	-0.45	-0.04	0.29
-7.62	1.88	0.44	4.06	3.77	0.15	-0.65	-0.17
10.67	6.17	8.45	-3.00	-0.06	-0.03	0.05	-0.06
-2.19	-7.95	0.51	-0.12	0.22	-0.68	-0.05	-0.10
3.58	4.43	-0.29	0.37	-0.12	0.18	0.11	0.09
-4.52	-0.29	-1.11	0.05	-0.04	0.69	-0.22	0.54
0.79	0.04	0.26	0.50	0.21	-0.42	0.51	-0.17



# Frequency Domain Transformation Module





## Differential Privacy Perturbation Module



- Traditional DP vs. image DP
- Element wise distance:

$$d_{i,j,k}(x_1, x_2) = \frac{|x_1 - x_2|}{r_{max}^{i,j,k} - r_{min}^{i,j,k}} \quad \forall x_1, x_2 \in R_{i,j,k}$$

Whole representation distance:

$$d(X_1, X_2) = \max_{i,j,k} (d_{i,j,k}(x_1, x_2))$$
  
$$\forall X_1, X_2 \in \mathbb{R}^{H,W,C}$$





- Used a model pre-trained on unperturbed images
  - 733 images from 12 different people from different ethnicities
- The perturbed images were validated against the model and accuracy was measured





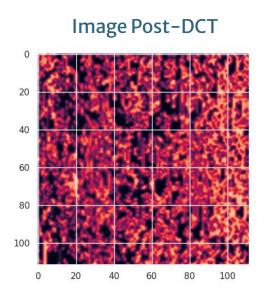
#### **Privacy Preservation and Image Recognition**

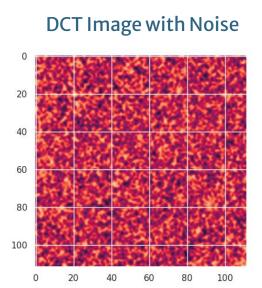
# Results





- Image post-DCT unreadable to human eye
- DCT image with noise prevents attacker from reconstructing the original image









CNN Model Accuracy Comparison				
Baseline Images	Images with Privacy Implementation			
72.58%	13.51%			

- Privacy/utility trade-off
- Utility better than random guess
- Future Work





Who was that celebrity?

# Discussion





#### Who is it?

Original Image

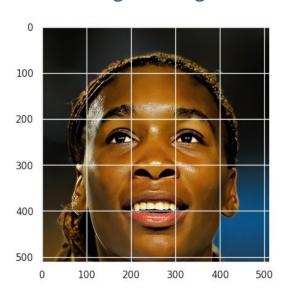
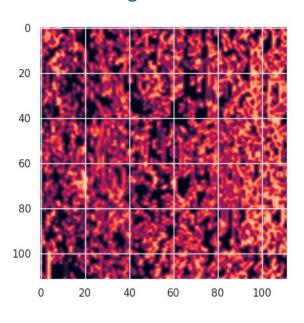


Image Post-DCT



**Venus Williams** 

DCT Image with Noise

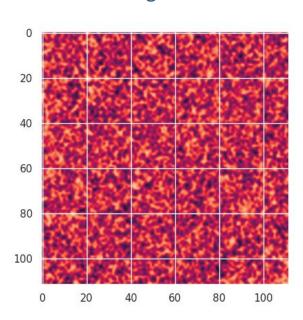


Image source: VGG-Face2: A dataset for recognising faces across pose and age.



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