



Javna agencija za znanstvenoraziskovalno
in inovacijsko dejavnost Republike Slovenije

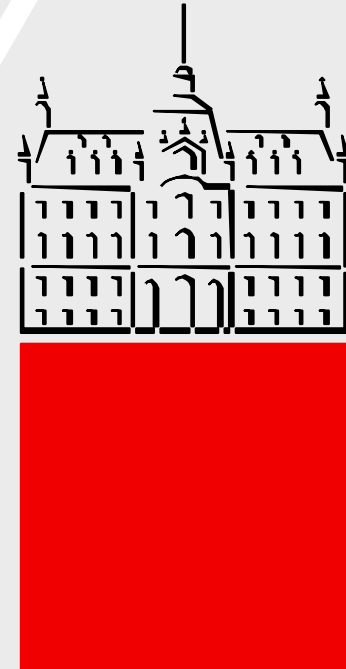
DifFIQA: Face Image Quality Assessment Using Denoising Diffusion Probabilistic Models

Prof. Vitomir Štruc, PhD

Faculty of Electrical Engineering
University of Ljubljana, Slovenia

vitomir.struc@fe.uni-lj.si

EAB Face Image Quality Workshop



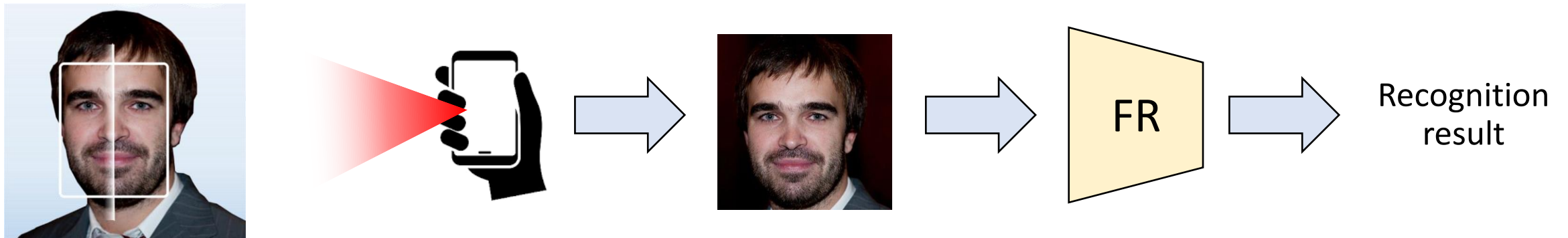


Face Image Quality Assessment

- Face Image Quality Assessment (FIQA)

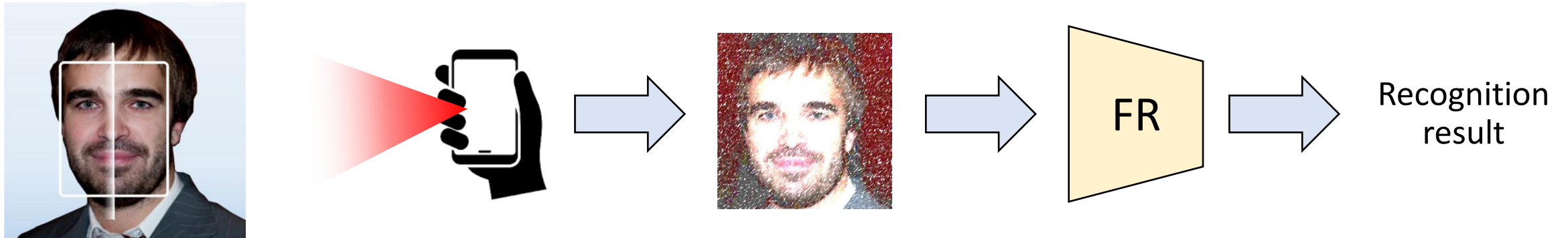
Face Image Quality Assessment

- Face Image Quality Assessment (FIQA)
 - Key use case in Face Recognition



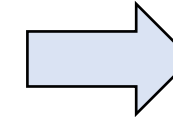
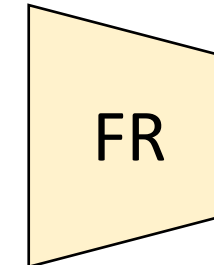
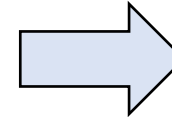
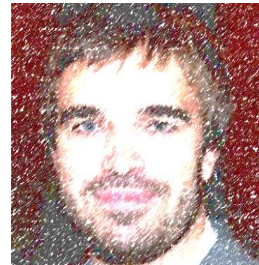
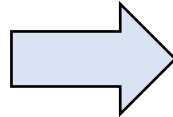
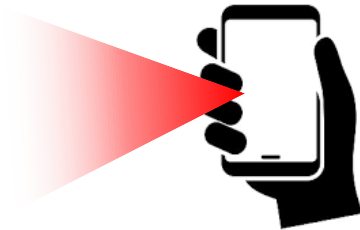
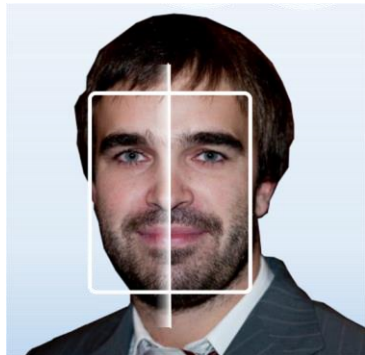
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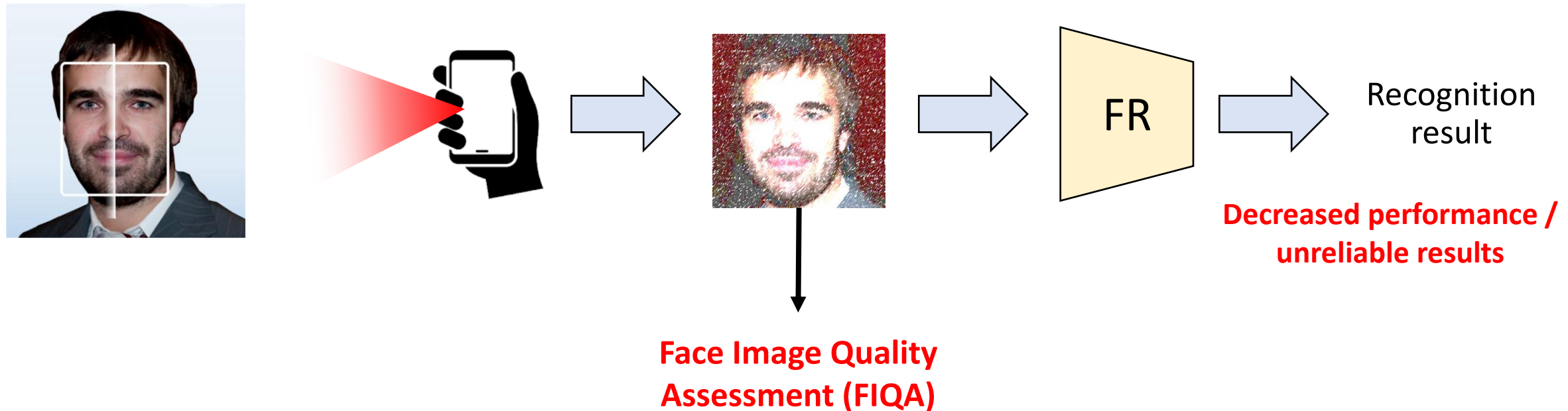


Recognition result

**Decreased performance /
unreliable results**

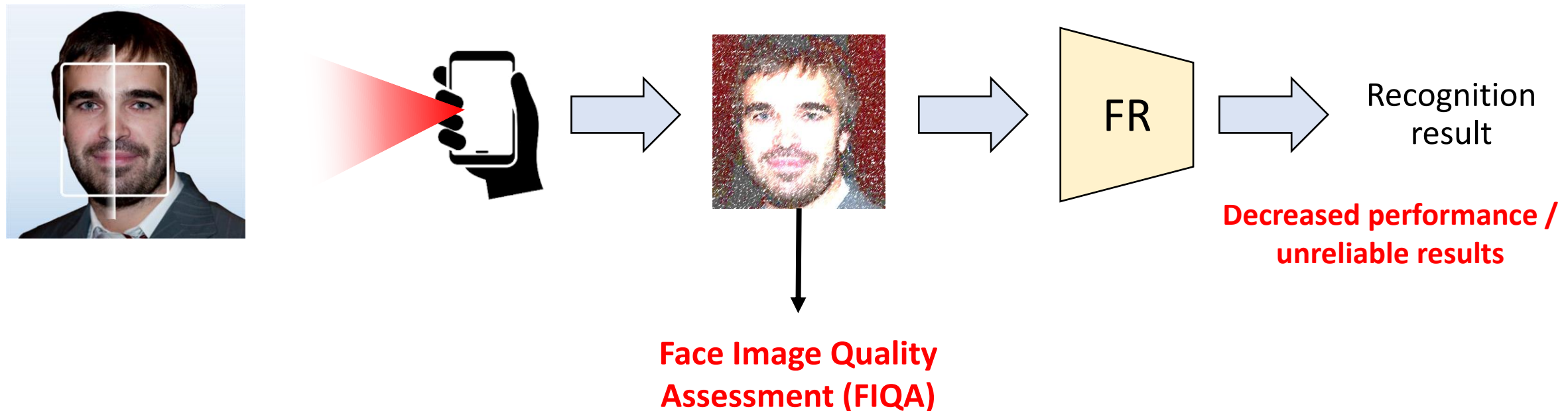
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- Reject sample, use quality for recognition, etc.

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High perceptual
quality

High face image
quality

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High perceptual
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High perceptual
quality
Low face image
quality

Face Image Quality Assessment

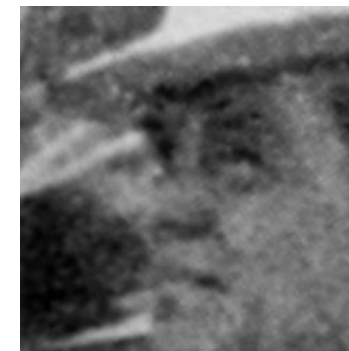
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High perceptual
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High perceptual
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Low face image
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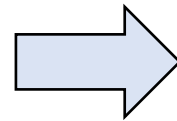
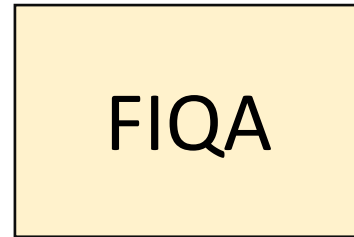
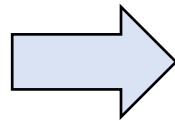
Low perceptual
quality
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Face Image Quality Assessment

EAB Face Image Quality Workshop

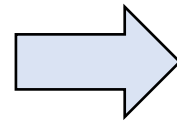
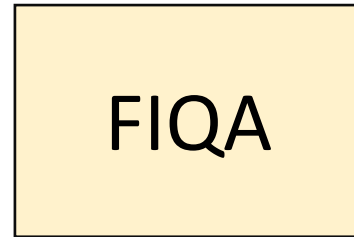
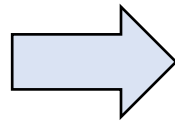


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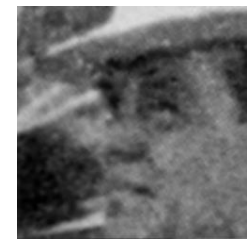


q

Face Image Quality Assessment

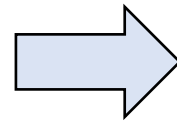
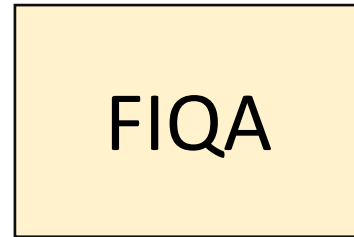
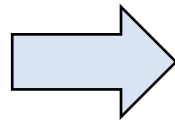
 q  q_1

>

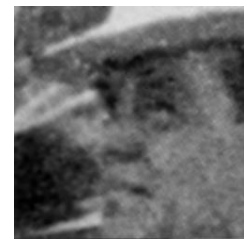
 q_2

**Higher scores
mean better quality**

Face Image Quality Assessment

 q  q_1

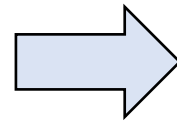
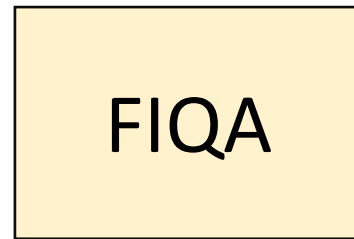
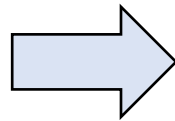
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 q_2

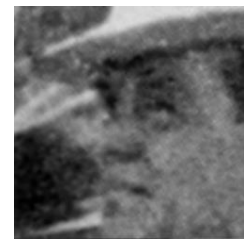
**Higher scores
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Existing solutions:

Face Image Quality Assessment

 q  q_1

>

 q_2

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Existing solutions:

Analytical:

Compute scores directly from the input sample.

Regression based:

Train regression model on extracted pseudo-quality labels.

Model based:

Combine recognition and quality estimation tasks.

DifFIQA: Quality Assessment Using Denoising Diffusion Probabilistic Models

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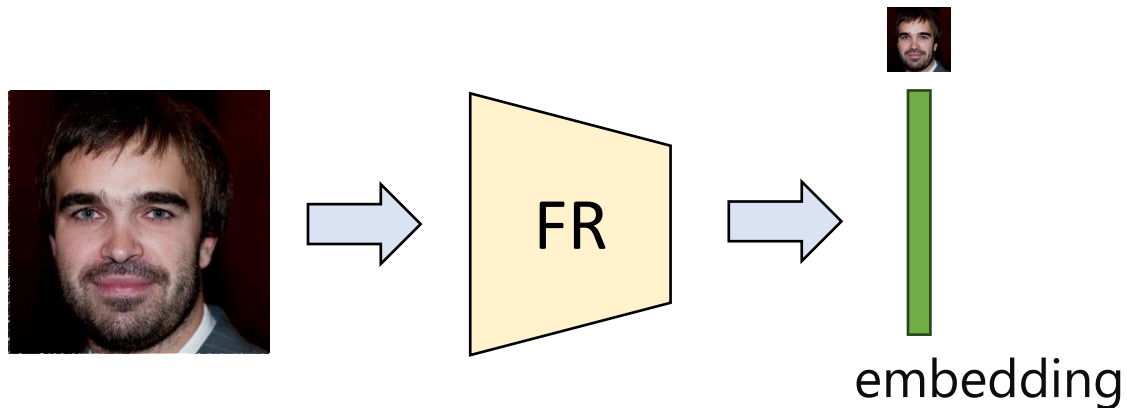
Perturbation robustness

Reconstruction quality

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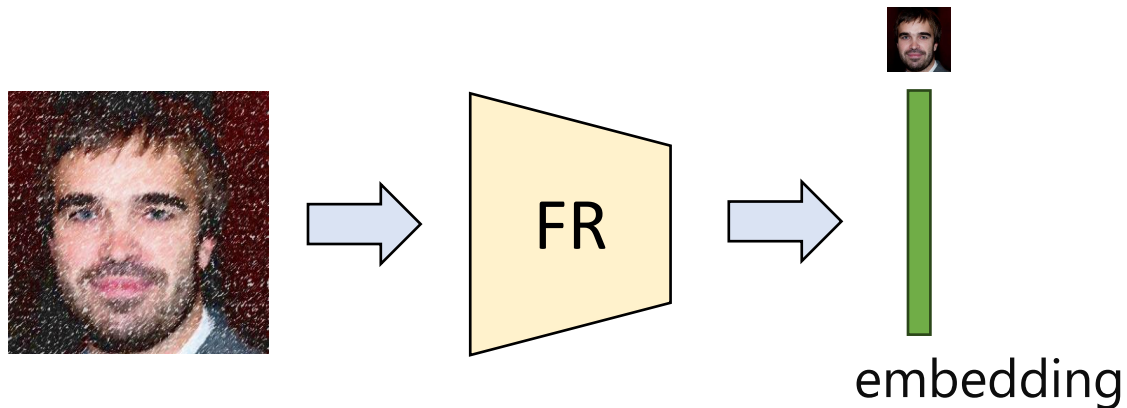


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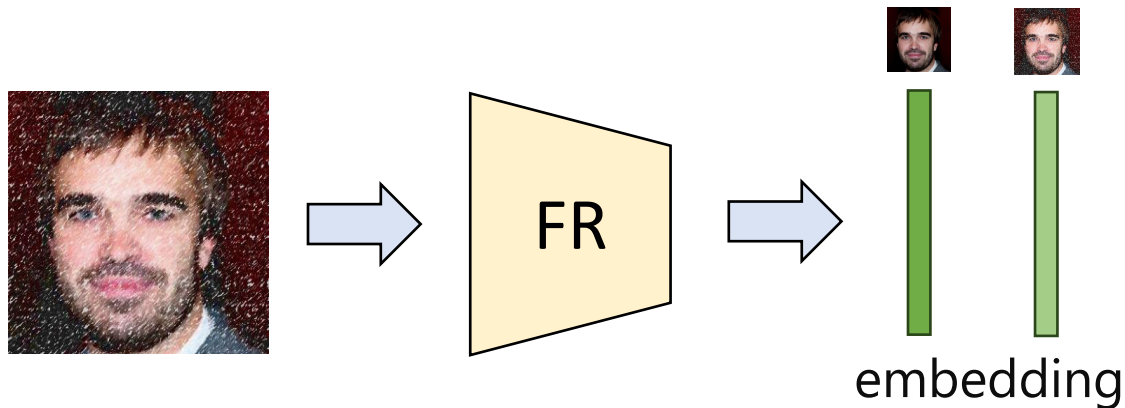


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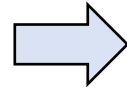
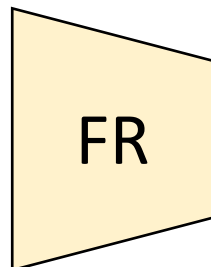
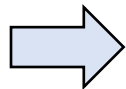
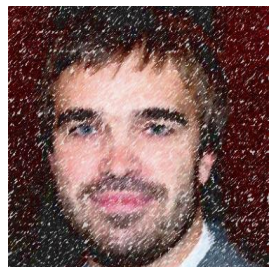


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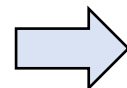
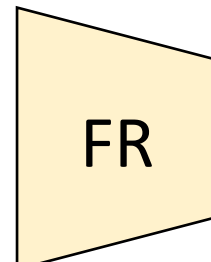
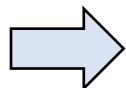
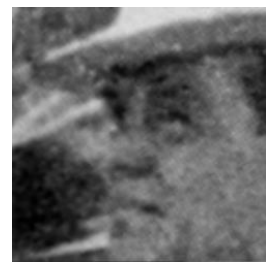
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embedding

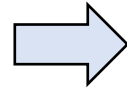
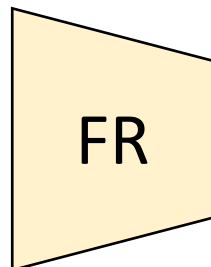
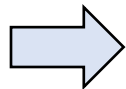
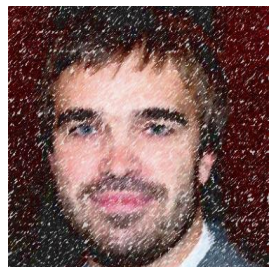


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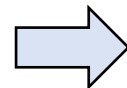
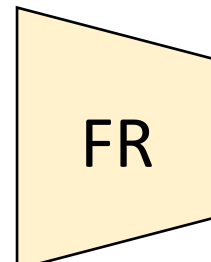
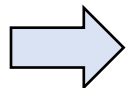
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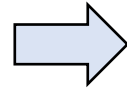
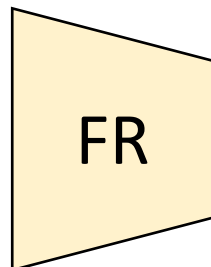
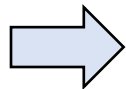
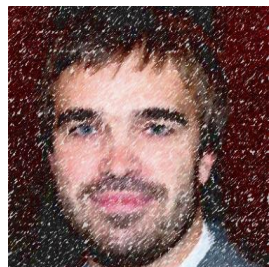


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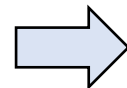
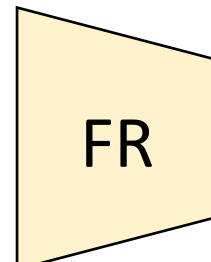
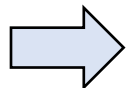
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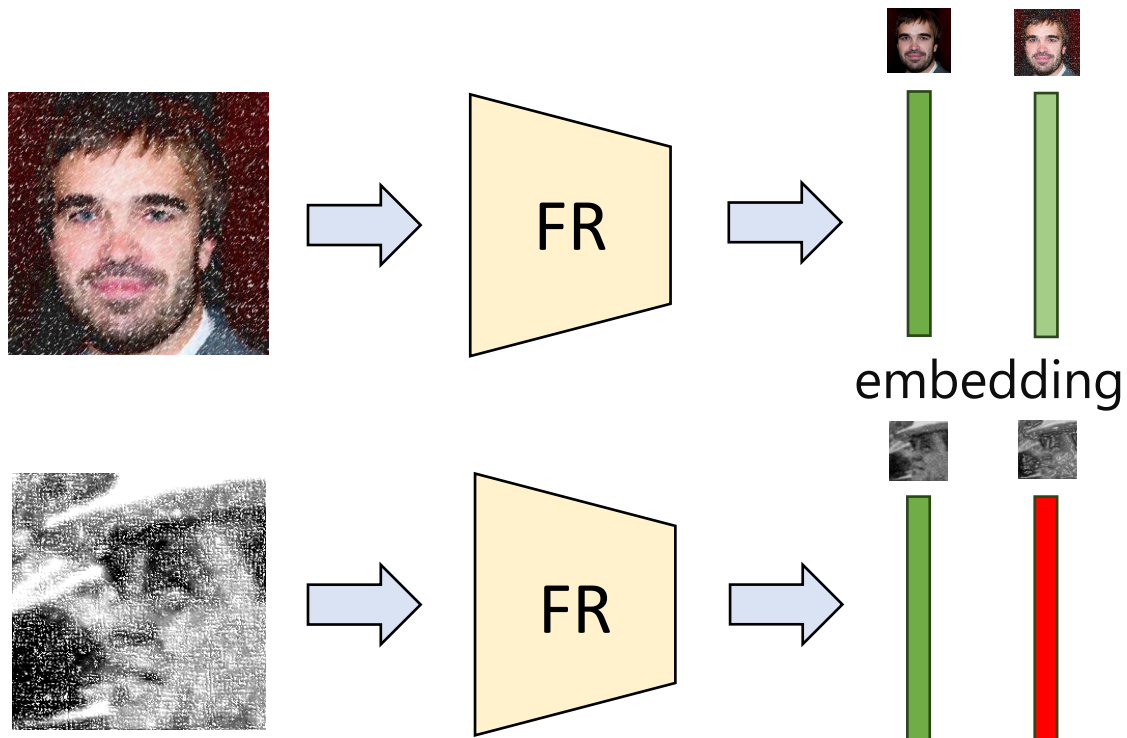


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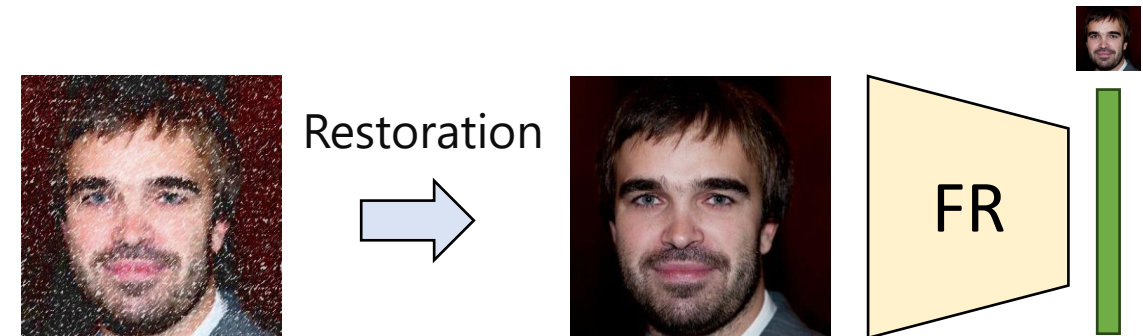
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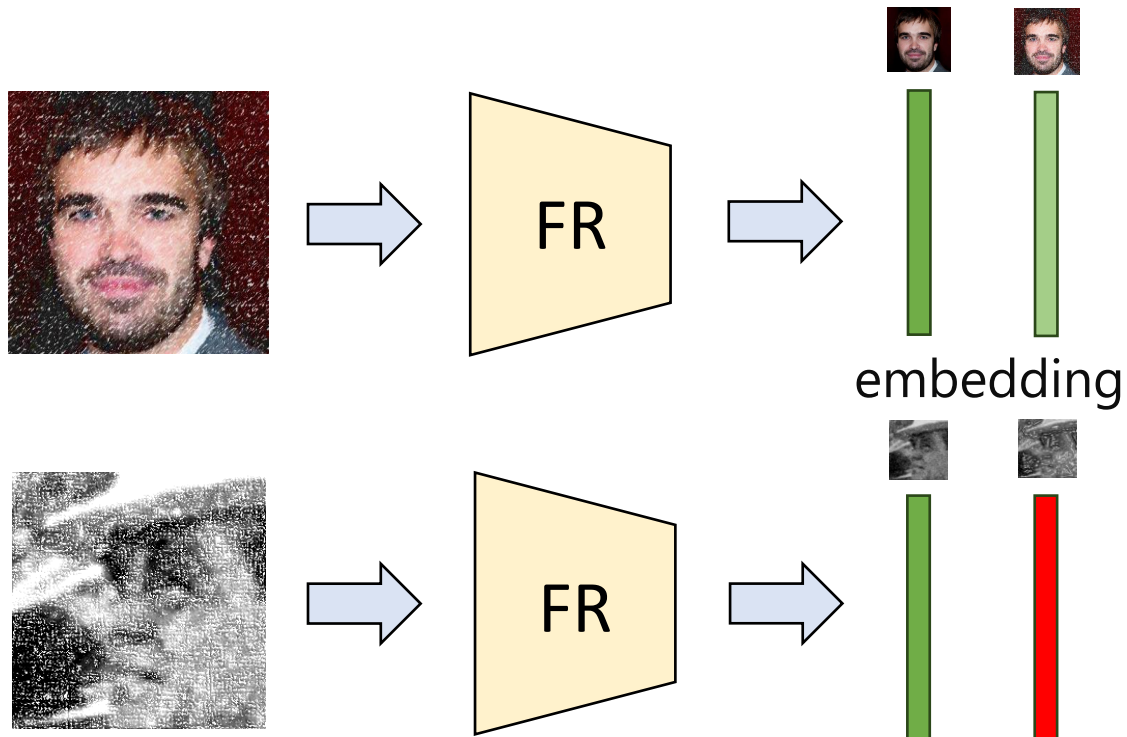
Restoration quality



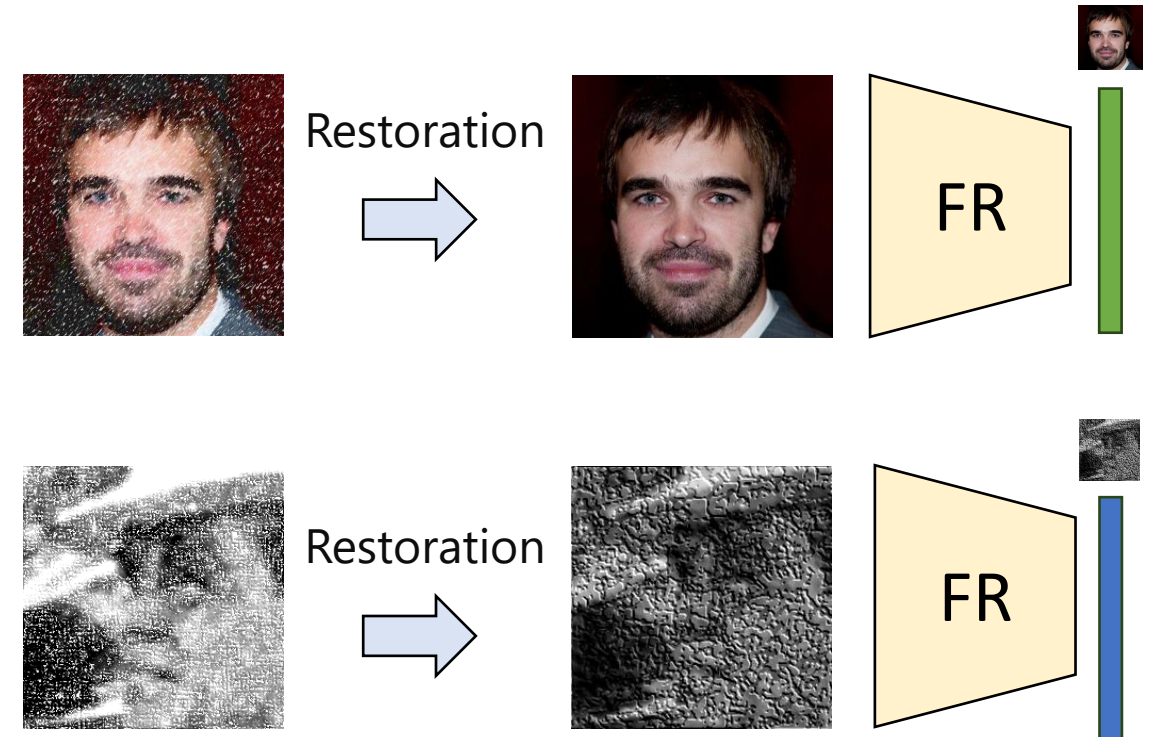
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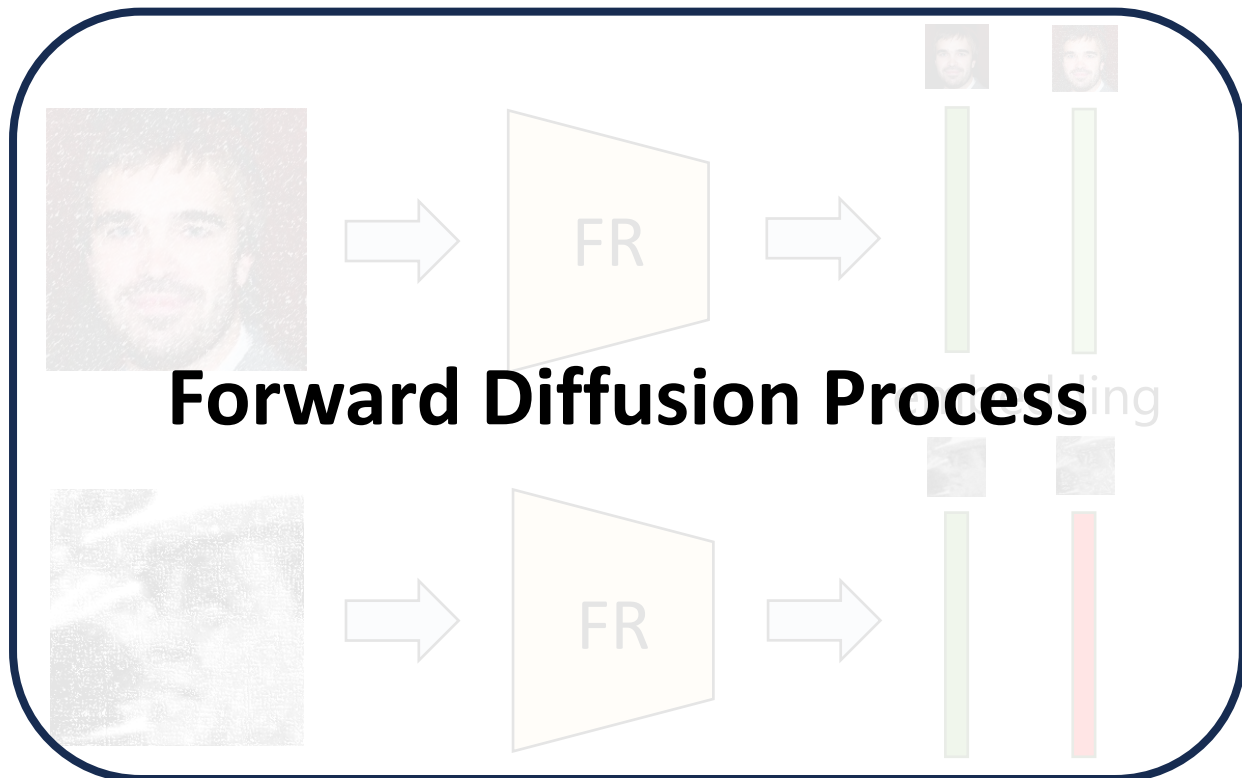
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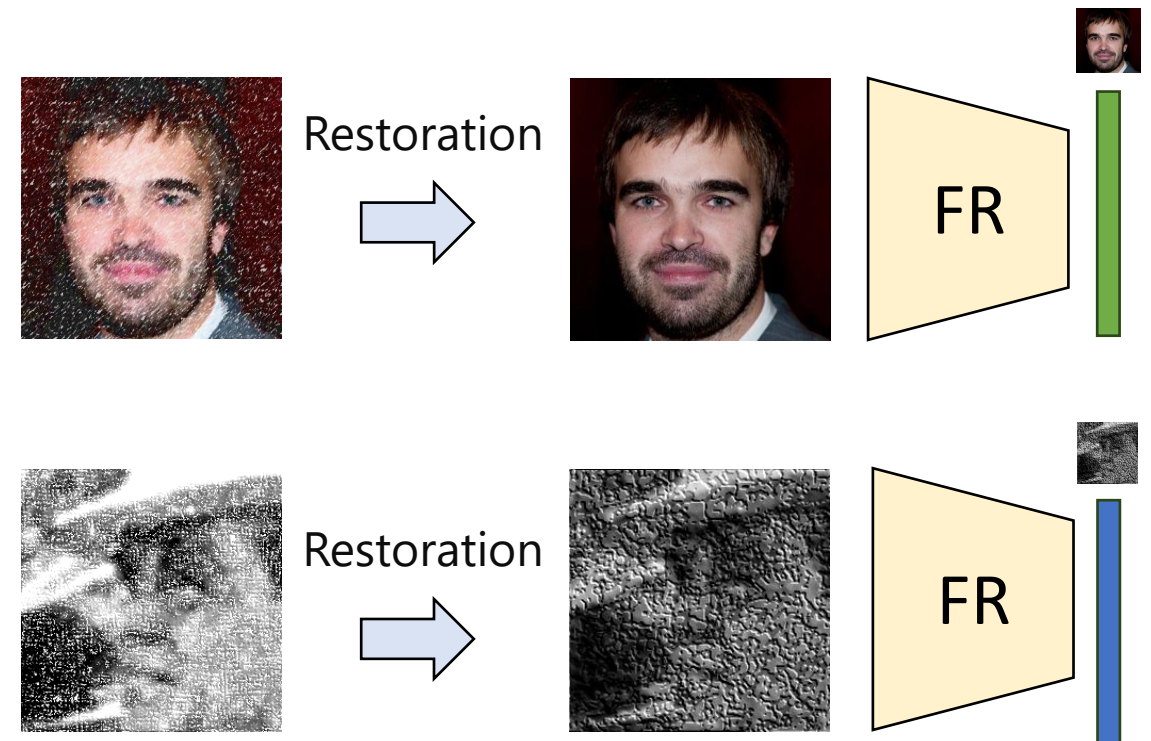
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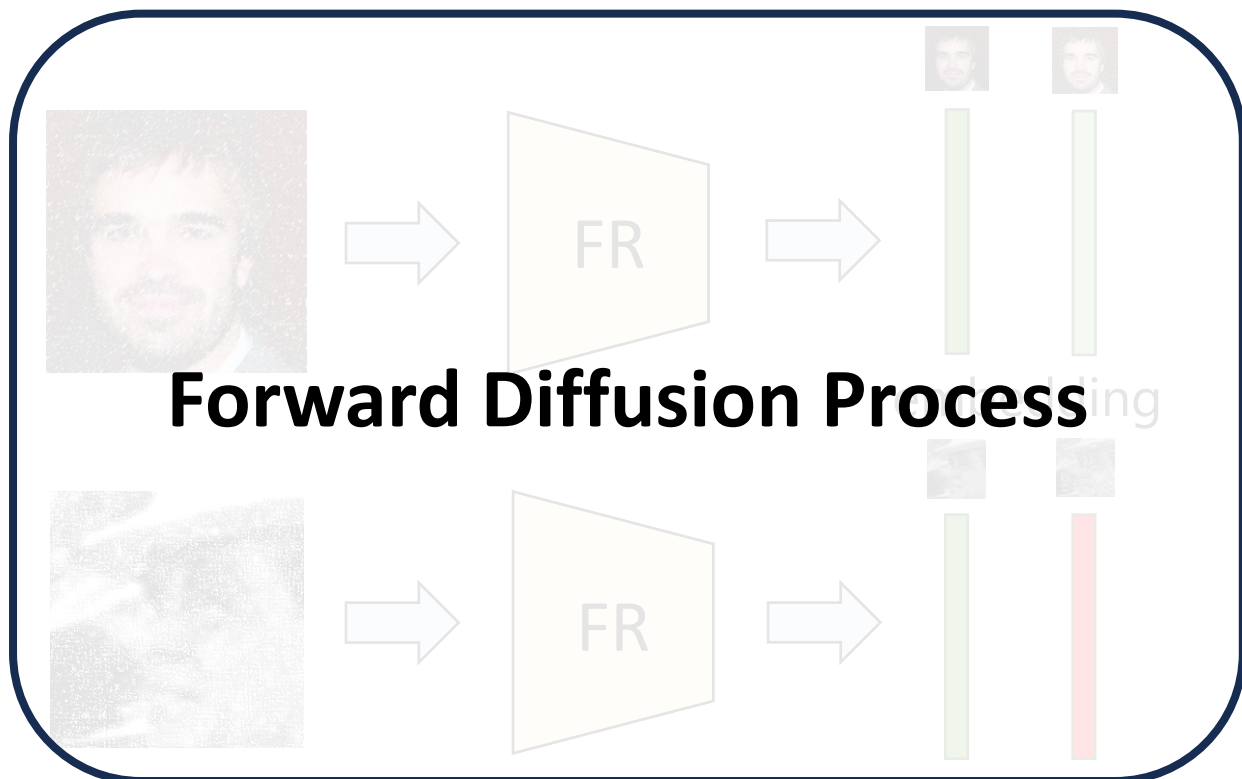
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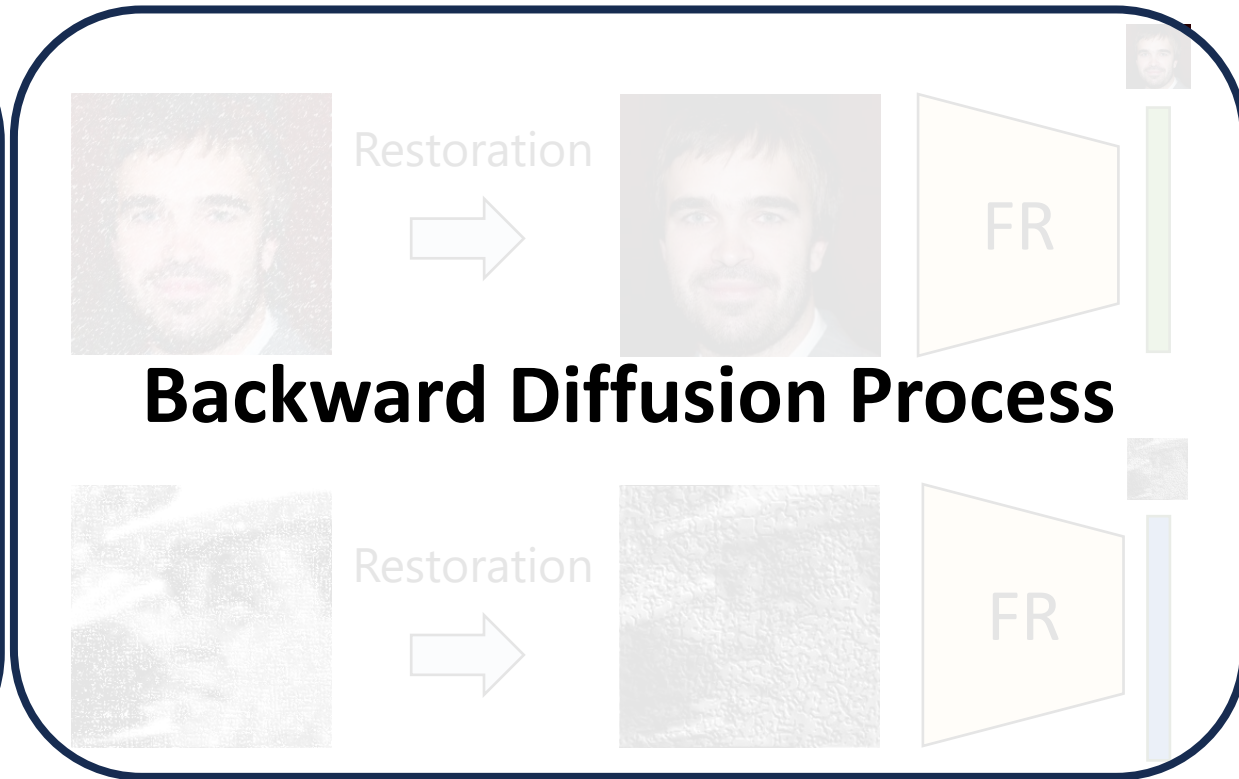
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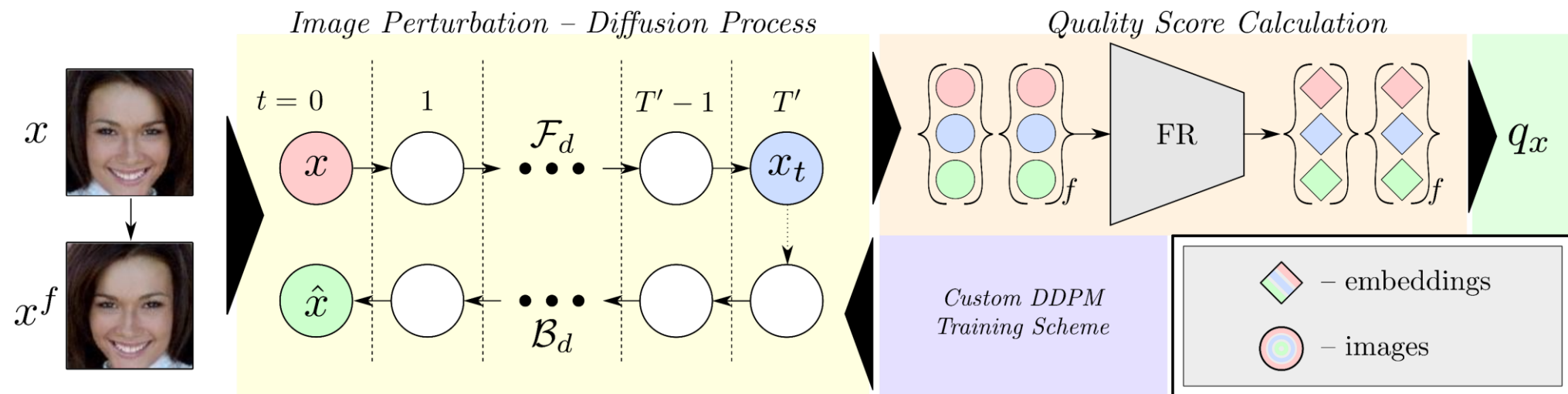


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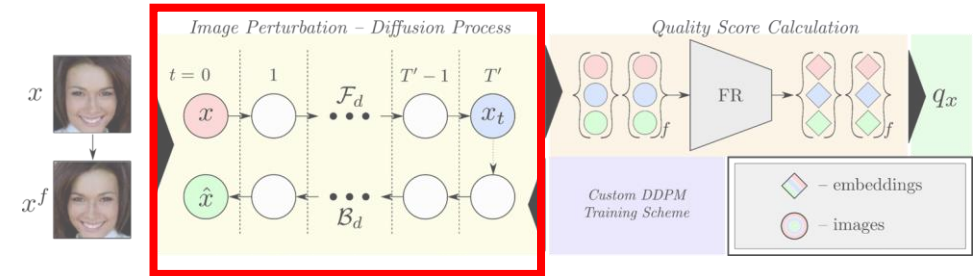
DifFIQA: Quality Assessment Using Denoising Diffusion Probabilistic Models

- DifFIQA (high-level) overview:
 - Step 1: Forward and backward diffusion (i.e., perturbations and reconstructions)
 - Step 2: Quality score calculation



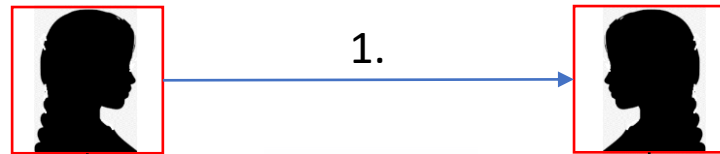
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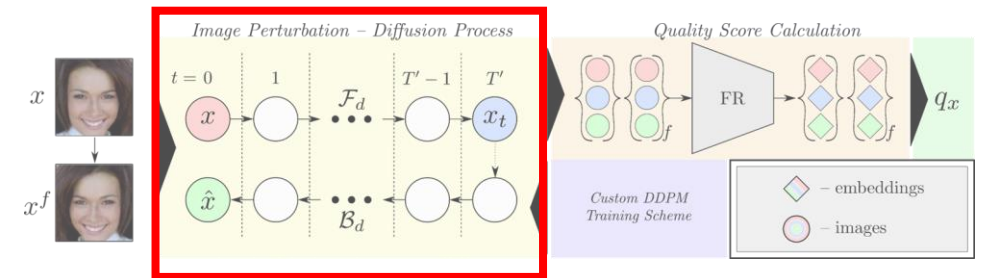


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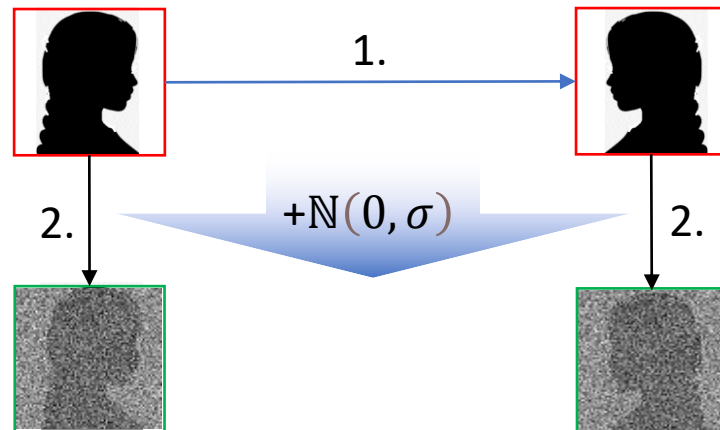


Flip image



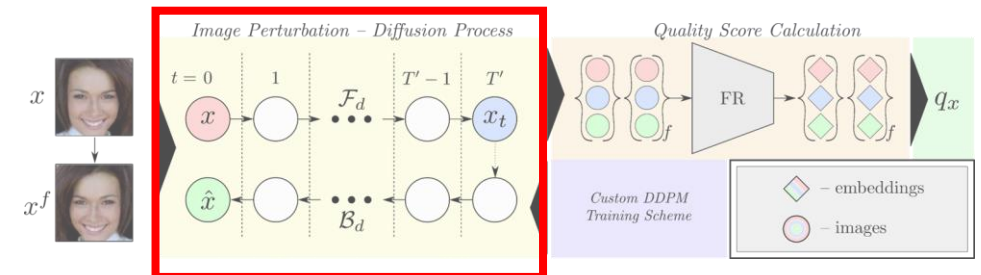
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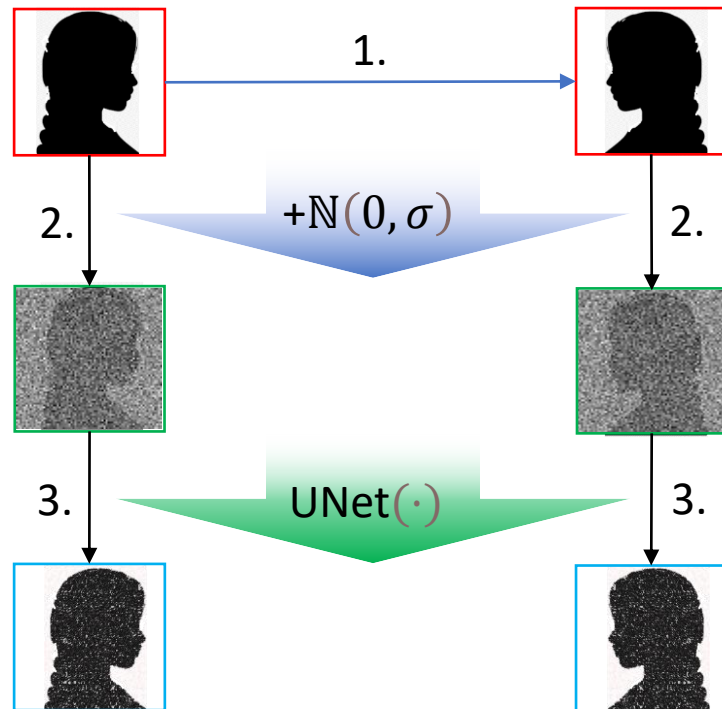
Flip image

Forward diffusion – adds noise



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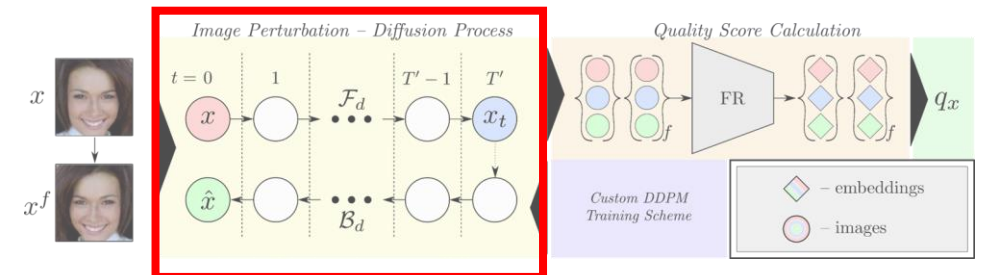
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Flip image

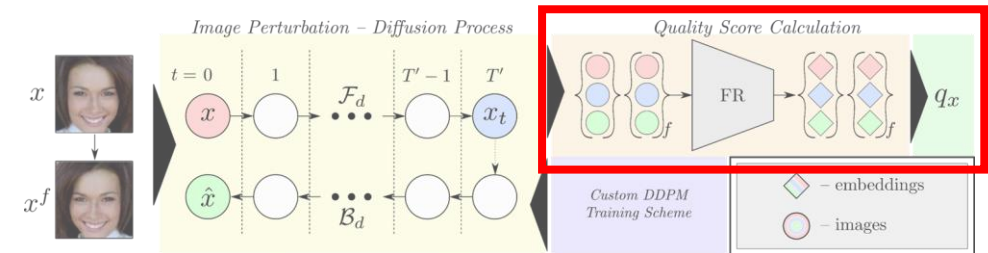
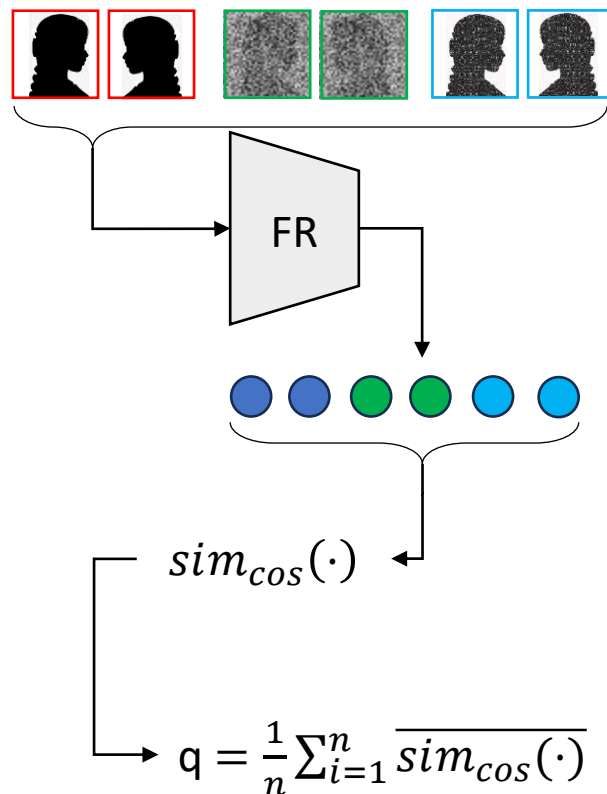
Forward diffusion – adds noise

Backward diffusion



DifFIQA: Quality Assessment Using Denoising Diffusion Probabilistic Models

- Step 2: Quality score calculation



1. Construct image embeddings:

- using any target FR model,
- for all three image pairs.

2. Calculate embedding similarity:

- using cosine similarity,
- comparing the input image embedding to all others.

3. Repeat process n-times and calculate average.



DifFIQA: Quality Assessment Using Denoising Diffusion Probabilistic Models

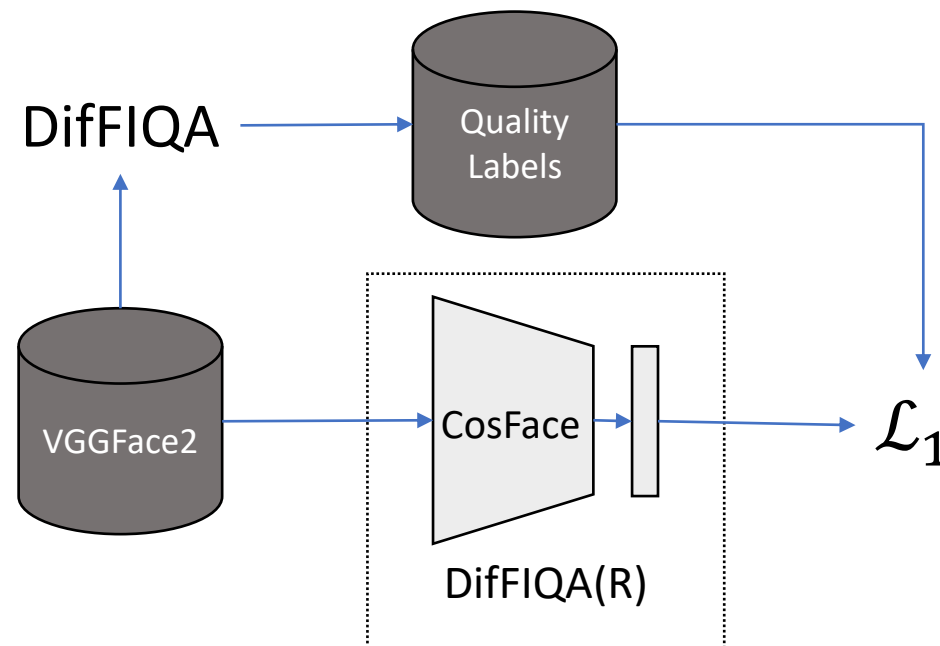
- Problem: Diffusion is slow!

DifFIQA: Quality Assessment Using Denoising Diffusion Probabilistic Models

- Problem: Diffusion is slow!
- DifFIQA(R) - Knowledge distillation

DifFIQA: Quality Assessment Using Denoising Diffusion Probabilistic Models

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Fine-tune a pretrained FR model using pseudo-quality labels extracted using DifFIQA.

Experimental Setup

- **Standard evaluation approach:**

- **Error-Versus Discard Characteristic (EDC) curves**

- *Show how the performance of FR models improves with increasing discard rates.*

- **partial Area Under the Curve (pAUC)**

- *Lower value indicates better performance.*
 - *Limit the discard ratio to (0.2) or (0.3).*

- **Benchmarks:**

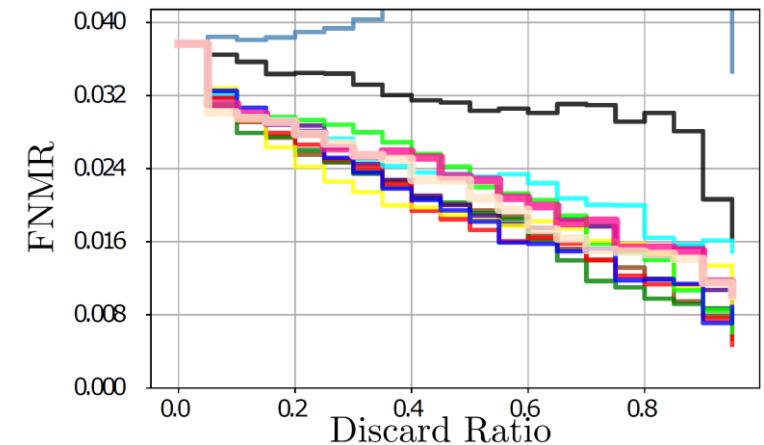
- *Adience, CALFW, CFP-FP, CPLFW, IJB-C, LFW, XQFW*

- **FR models:**

- *AdaFace, ArcFace, CurricularFace, CosFace*

- **FIQA methods:**

- *FaceQnet, SDD-FIQA, PFE, PCNet, MagFace, LightQNet, SER-FIQ, FaceQAN, CR-FIQA, FaceQgen*



Results (pAUC)

- Average results over all benchmarks and FR models

Results using discard ratio of (0.2).

FaceQnet [17]	SDD-FIQA [33]	PFE [39]	PCNet [44]	MagFace [31]	LightQNet [7]	SER-FIQ [40]	FaceQAN [3]	CR-FIQA [5]	FaceQgen [15]	DifFIQA	DifFIQA(R)
0.9458	0.8244	0.8197	0.8989	0.8253	0.8183	0.7985	0.7519	0.7567	0.8527	0.7591	0.7518

Results using discard ratio of (0.3).

FaceQnet [17]	SDD-FIQA [33]	PFE [39]	PCNet [44]	MagFace [31]	LightQNet [7]	SER-FIQ [40]	FaceQAN [3]	CR-FIQA [5]	FaceQgen [15]	DifFIQA	DifFIQA(R)
0.9315	0.7483	0.7497	0.8691	0.7635	0.7412	0.7292	0.6847	0.6800	0.7954	0.6822	0.6768

- Knowledge distillation improves the results of the base approach
- DifFIQA(R) outperforms SOTA methods.

Ablation Study

- **Showcase performance changes when:**
 - **(A1)** *Removing image flipping.*
 - **(A2)** *Include only perturbations from the backward diffusion pass.*
 - **(A3)** *Use a different number of timesteps for noise generation.*

Model variant	LFW	CPLFW	CALFW	XQLFW	$\overline{\text{pAUC}}$
(A1): w/o Image Flipping	0.702	0.727	0.888	0.535	0.713
(A2): w/o Forward Pass	0.730	0.684	0.897	0.531	0.710
(A3): DifFIQA ($t = 20$)	0.657	0.694	0.945	0.628	0.731
DifFIQA (complete)	0.695	0.669	0.900	0.546	0.702

DifFIQA best overall performance on a varied set of benchmarks.

Run-time

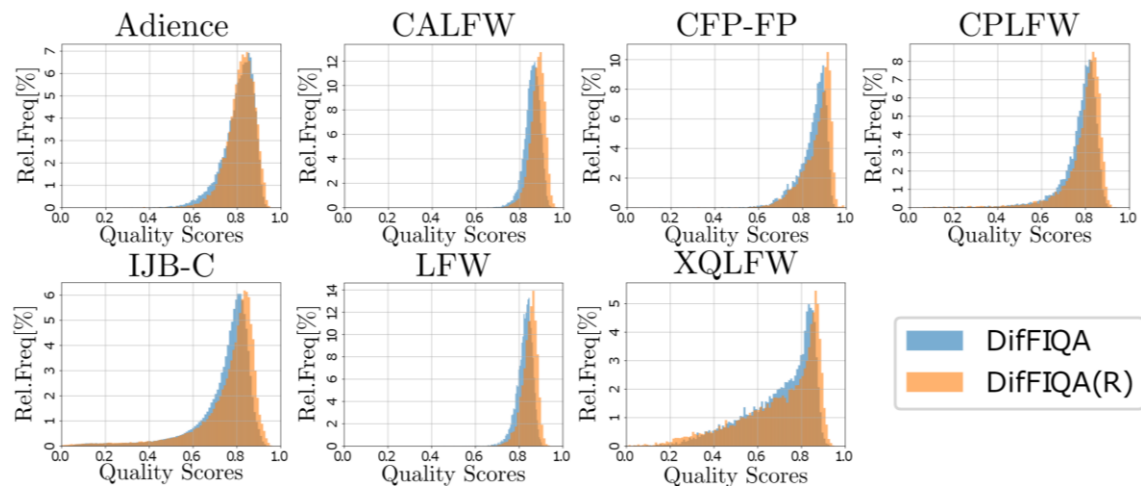
- **Hardware: Intel i9-10900KF CPU, 64 GB RAM in Nvidia 3090 GPU**

FIQA model	Ours	
	DifFIQA	DifFIQA(R)
Runtime ($\mu \pm \sigma$)	1074.62 \pm 11.45	1.24 \pm 0.36

- Speed-up due to knowledge distillation is around **1000x**

DifFIQA and DifFIQA(R) Comparison

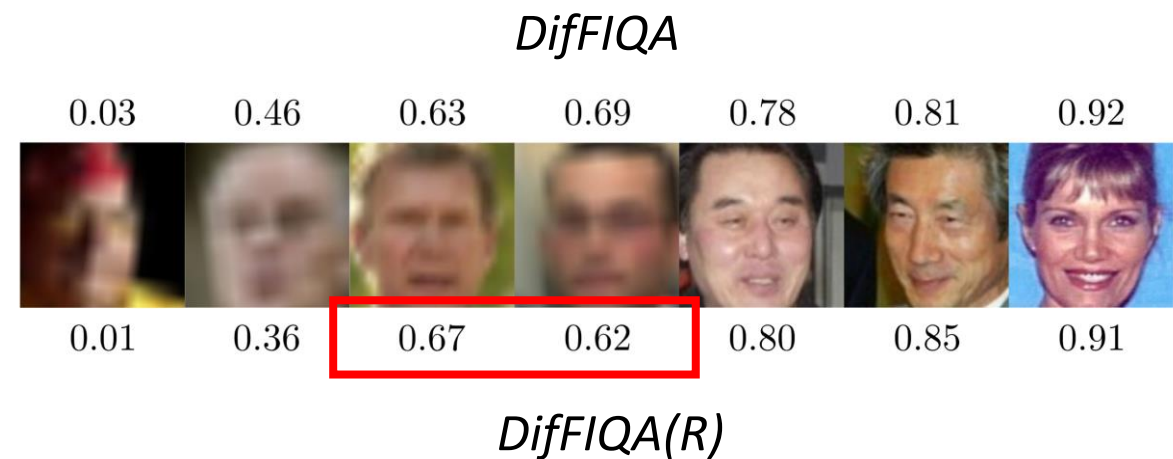
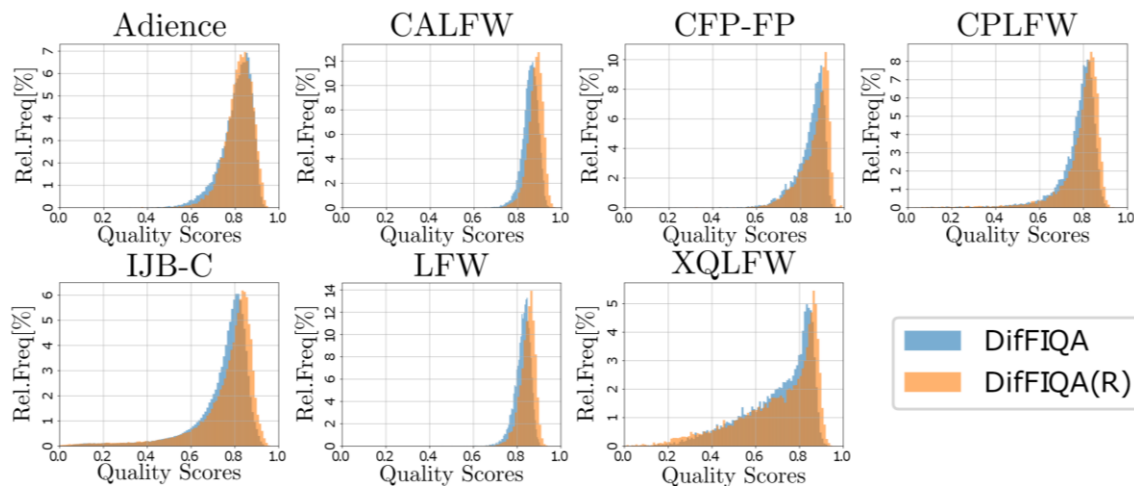
Comparison of quality score distributions



DifFIQA and DifFIQA(R) Comparison

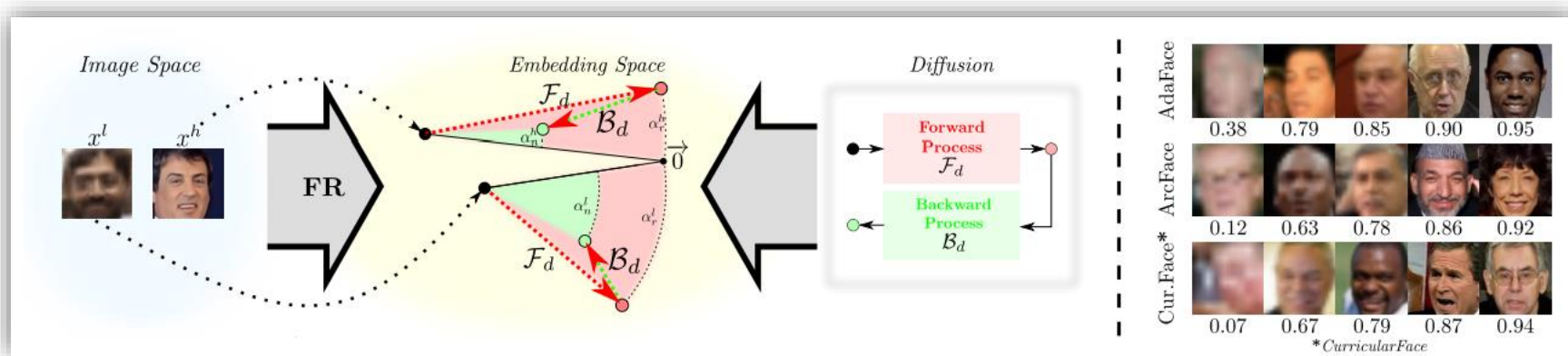
Comparison of quality score distributions

Visual comparison



Conclusions

- DifFIQA: a new FIQA approach based on DDPMs
- DifFIQA(R): distilled version for fast prediction
- State-of-the-art performance across multiple datasets and face recognition models
- Outlook and future work
 - Distillation with lighter models
 - Distillation with score-optimization for fine-grained FIQA
 - Incorporating detector variability
 - FIQA for videos



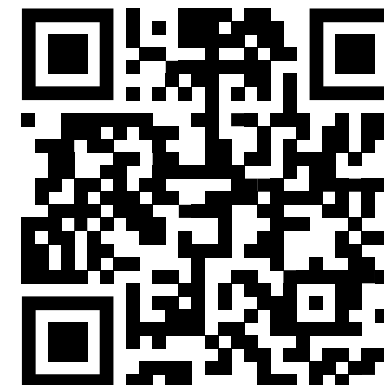
Paper



Thank You!

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Code



EDC curves

