

Training GANs with Stronger Augmentations via Contrastive Discriminator

by Harum Naseem Bhinder & Muhammad Raahim Khan

Background Literature

THE GAP BETWEEN
SUPERVISED AND
UNSUPERVISED
LEARNING CAN BE
SIGNIFICANTLY CLOSED
BY LARGE SCALE
CONTRASTIVE LEARNING

ACTIVE RESEARCH
CONDUCTED ON
VARIOUS DATA
AUGMENTATION
TECHNIQUES



PROBLEM STATEMENT

DESPITE ALL THE
ACTIVE RESEARCH,
STILL UNCLEAR WHICH
AUGMENTATIONS
COULD ACTUALLY
IMPROVE GANS

Why is it important?

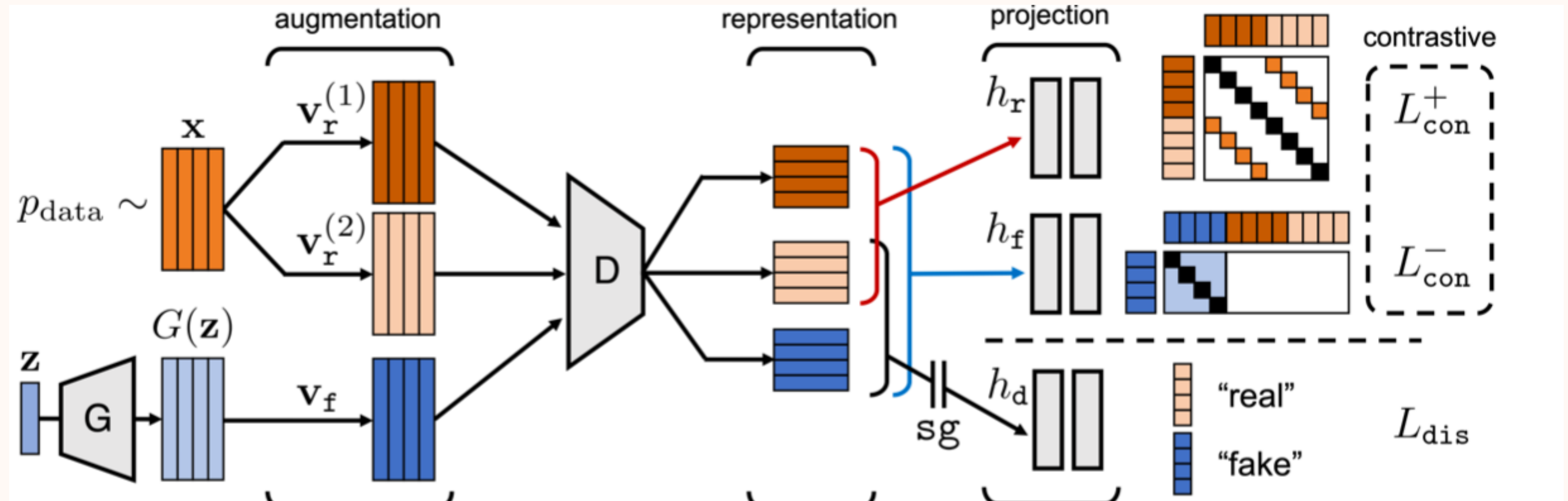
WE CAN TRAIN THE
MODEL TO LEARN A LOT
ABOUT OUR DATA
WITHOUT ANY
ANNOTATIONS OR
LABELS

LEARN THE HIGH-LEVEL
FEATURES OF THE
OBJECTS IN OUR
WORLD



SOLUTION PROPOSED

SUGGESTED
INCORPORATING A
CONTRASTIVE
REPRESENTATION
LEARNING SCHEME
INTO THE GAN
DISCRIMINATOR
CALLED CONTRAD



ARCHITECTURE

THE ENCODER NETWORK IS TRAINED TO MINIMIZE 2 CONTRASTIVE LOSSES AND IS THEN MERGED UNDER THE STANDARD FRAMEWORK OF GAN

How is it different?

ENABLES THE
DISCRIMINATORS TO
WORK WITH MUCH
STRONGER
AUGMENTATIONS

DOES NOT INCREASE
THE TRAINING
INSABILITY AND SO
PREVENTS
DISCRIMATOR
OVERFITTING

Results Demonstrated

GANS WITH CONTRAD
CONSISTENTLY IMPROVED
FRECHET INCEPTION DISTANCE
(FID) AND INCEPTION SCORE (IS)

MAINTAINED HIGHLY
DISCRIMINATIVE FEATURES IN THE
DISCIMNATOR IN TERMS OF LINEAR
EVALUATION

Results Demonstrated

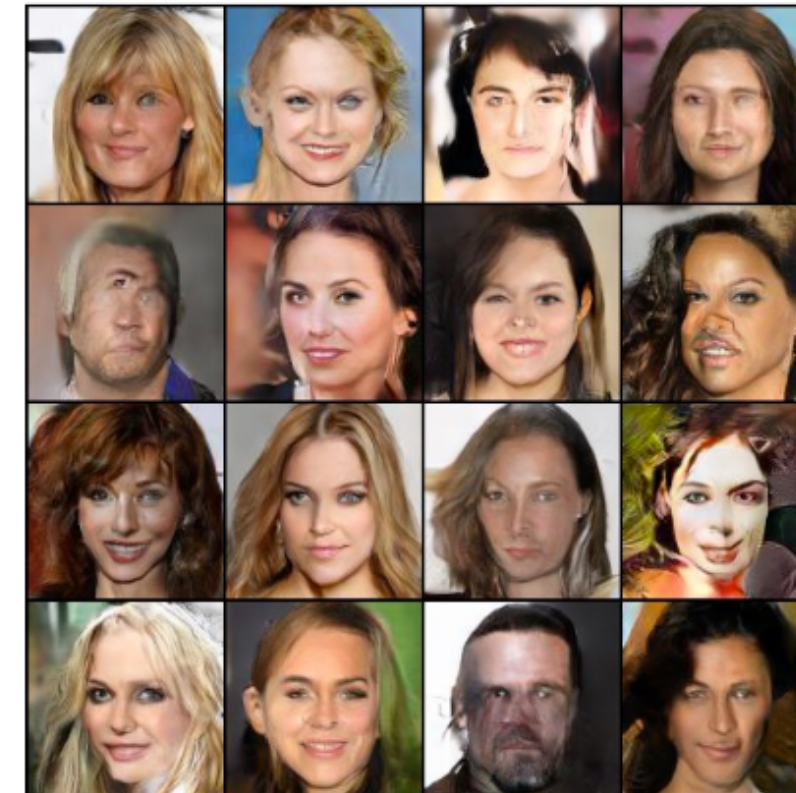
W/O (FID: 24.8)



CR (FID: 20.9)




bCR (FID: 19.5)



ContraD (FID: 17.1)





GANs trained in an unsupervised manner can induce
many conditional generative models via a simple
latent sampling by leveraging the learned features
of ContraD

OTHER CONTRIBUTION

Limitations

NO LIMITATIONS WERE POINTED OUT BY THE AUTHORS AND DUE TO THE LACK OF DEPTH IN THE METHODOLOGY STEPS WE DIDN'T HAVE ENOUGH BASIS TO DEDUCE LIMITATIONS

Future Work

PROPOSE AN IDEA OF
INCORPORATING FAKE SAMPLES
FOR CONTRASTIVE LEARNING

FURTHER EXPLORE MINIMIZING
GAN LOSS BY ADDING A SMALL
HEADER

THANK YOU!