

# 27th Vienna Deep Learning Meetup



22<sup>nd</sup> May 2019 @ Robert Bosch AG  
**#VDLM**



**BOSCH**

# Vienna Deep Learning Meetup



## The Organizers:



René Donner  
contextflow



Thomas Lidy  
Musimap



Alex Schindler  
AIT & TU Wien



Jan Schlüter  
OFAI & UTLN

# Topics for Today

## Introduction

### Deep Learning for Predictive Quality & Predictive Maintenance

*Daniel Ressi, Data Scientist – craftworks GmbH*

## Break

### Advances in Automotive In-Cabin Monitoring: Present & Future

*Florian Seitner, CEO and Michael Hödlmoser, CTO – emotion 3D*

### Hot Topics & Latest News

*Alex Schindler*

## Networking and Discussion

# Announcements

# Speech, Music and Mind 2019

Detecting and Influencing Mental States with Audio Targeted  
as a Satellite Workshop of Interspeech 2019

**14 SEPTEMBER, VIENNA AUSTRIA**

## Keynote Speakers



Prof. Carlos Busso Recabarren,  
University of Texas at Dallas



dhr. dr. J.A. (Ashley) Burgoyne

Facultät der Geisteswissenschaften  
Capacitierungsgruppe Musikwissenschaft

## Important Dates

- **24 June 2019:** Last date to submit your paper
- **29 Jul 2019:** Acceptance Notification
- **16 Aug 2019:** Registration deadline
- **30 Aug 2019:** Camera-ready papers due



Mr. Venkata Subramanian  
Viraraghavan  
TCS Research and Innovation



Alexander Schindler  
Vienna University of Technology



Joao Cabral  
Trinity College Dublin

### **Detecting stress, emotion or mental states of people from speech**

- Multi-modal approaches: using other modes such as video and sensor data in addition to speech
- Relevance of language models for mental state detection
- Cross-corpus detection on non-acted speech databases in multiple languages and realistic environments

### **Effects of Audio on stress, emotion and mental states of people**

- Audio-Visual Perception of music
- Analysis of brain signal responses to audio and visual stimulus
- Evaluation and Applications: augmented reality, art installations, music animations, computer games, etc

### **Other topics that are of interest in the context of stress, emotion and mental states**

- Novel signal processing or machine learning techniques
- Sounds at inaudible frequencies
- Novel protocols for assessing mental states, inducing stress or emotion
- Applications related to the above topics

# Break

# Hot Topics & Latest News

a short block at every meetup  
to briefly present recent papers and news in Deep Learning

Send us contributions ([tom.lidy@gmail.com](mailto:tom.lidy@gmail.com))  
or come with slides to do a short block yourself!

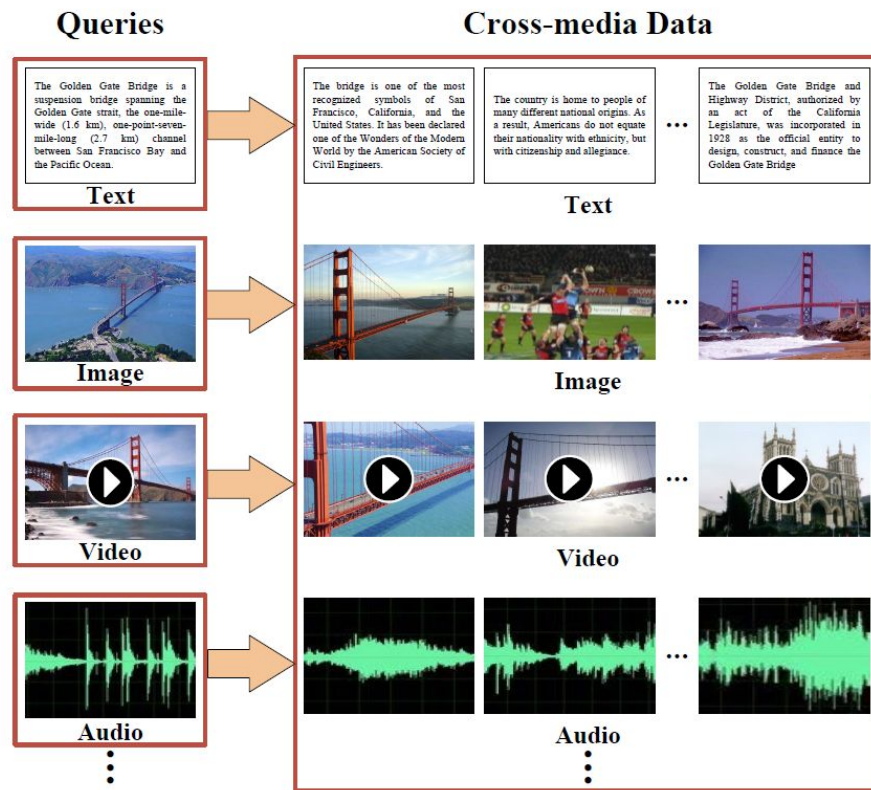


# Cross-Modal Retrieval

# Media Query / Retrieval

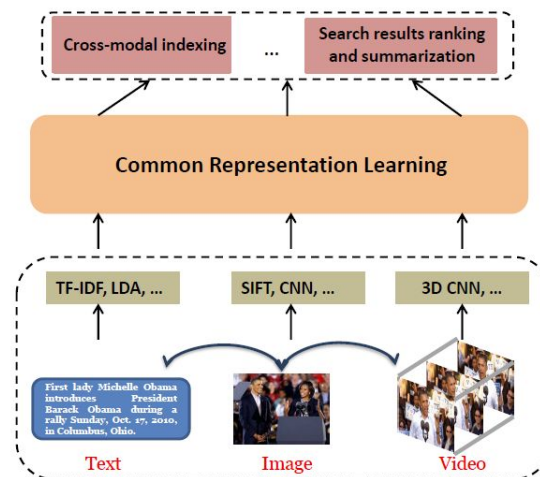
- Features extracted for one modality do not describe the other

⇒ **Heterogeneity Gap**

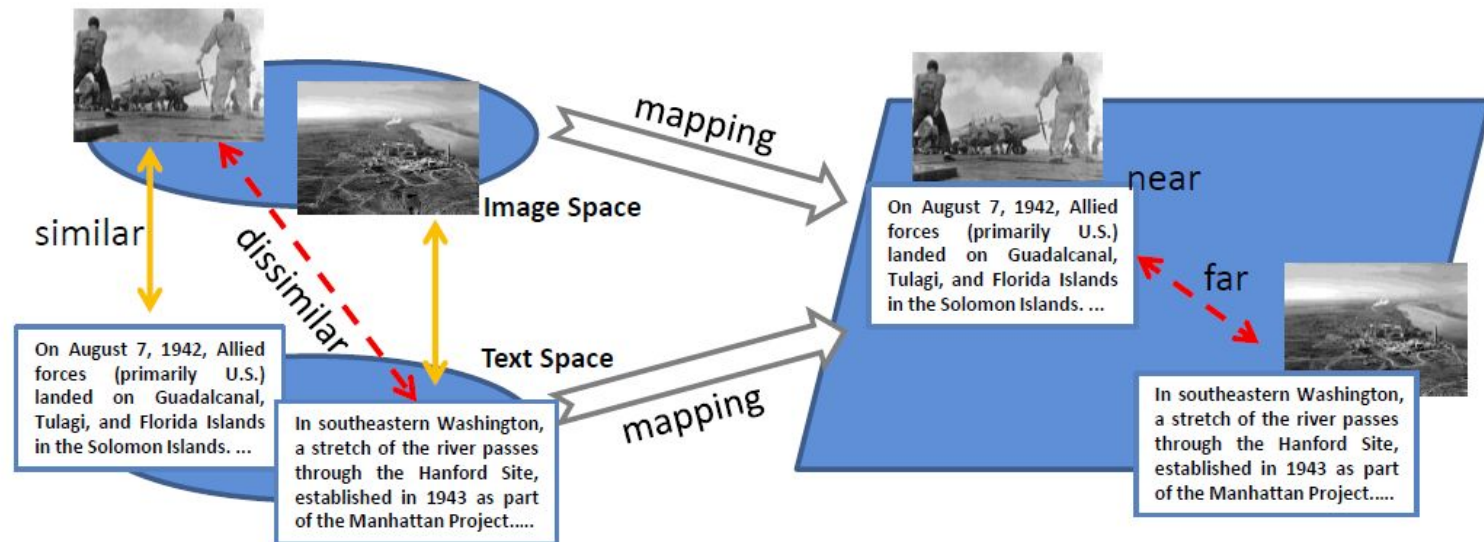


# Cross-Media Query / Retrieval

Categories	
Common space learning	Traditional statistical correlation analysis
	DNN-based
	Cross-media graph regularization
	Metric learning
	Learning to rank
	Dictionary learning
Cross-media similarity measurement	Cross-media hashing
	Others
Cross-media similarity measurement	Graph-based
	Neighbor analysis
Others	Relevance feedback analysis
	Multimodal topic model



# Basic Idea



The basic idea of heterogeneous metric learning, which learns projections using similar/dissimilar pairs.

# Problem: Exclusive & Shared Information



Image Modality

## Text Modality

Friends playing a  
game of Frisbee  
in a green park

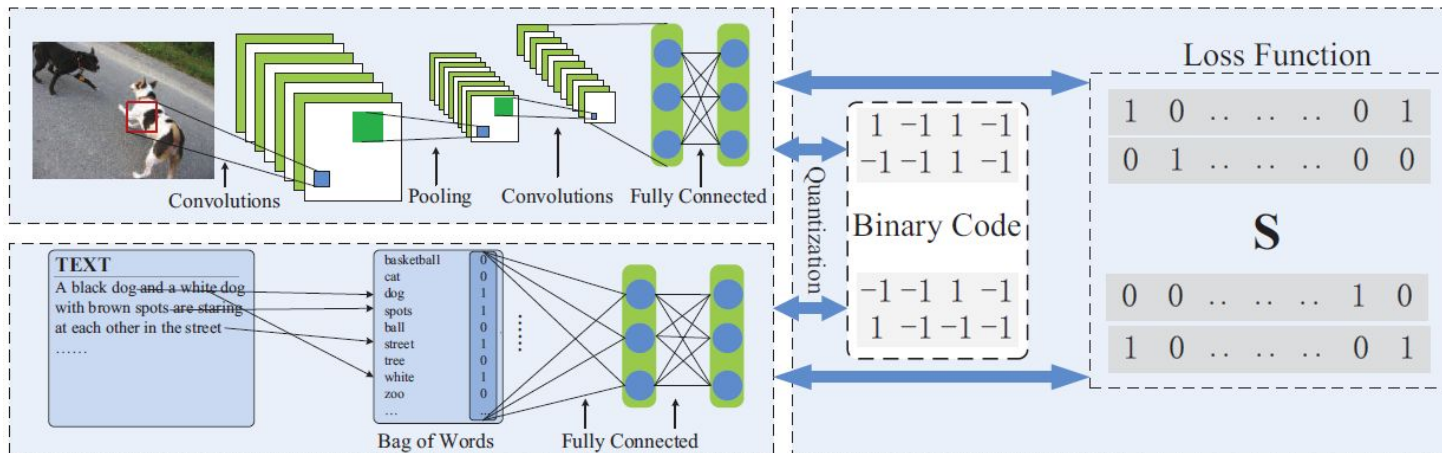
## Shared Information

- *Friends*
- *Frisbee*
- *green*
- *Park*

## Exclusive Information

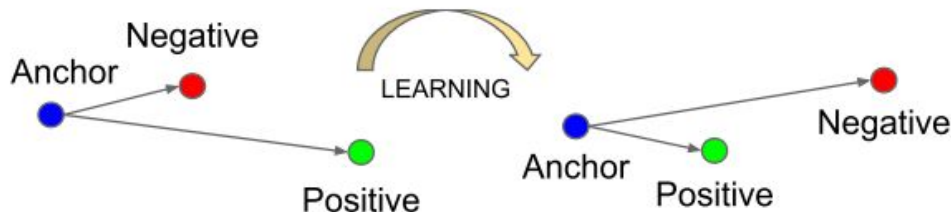
- Text: *playing, game, stopwords (a, of, in)*
- Image: *black/blue shorts, red/white shirt, baseball cap, fence, autumn*

# Cross-Modal Hashing



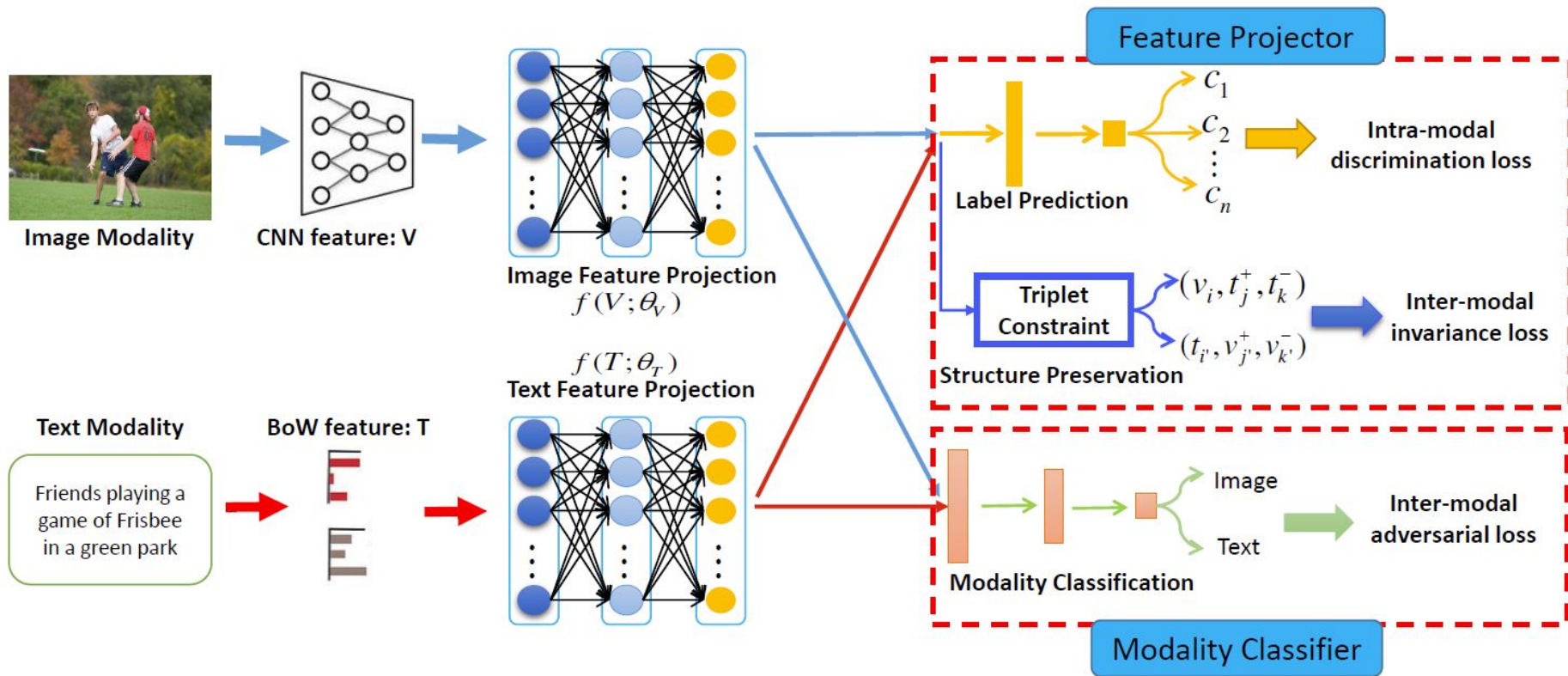
Jiang, Qing-Yuan, and Wu-Jun Li. "Deep cross-modal hashing." Proceedings of the IEEE conference on computer vision and pattern recognition. 2017.

## Triplet Loss / Hinge Loss



Schroff, F., Kalenichenko, D., & Philbin, J. (2015). FaceNet: A unified embedding for face recognition and clustering. In Proceedings of the IEEE Computer Society Conference on Computer Vision and Pattern Recognition (Vol. 07–12–June, pp. 815–823). <https://doi.org/10.1109/CVPR.2015.7298682>

# Adversarial Cross-Modal Retrieval





# Results

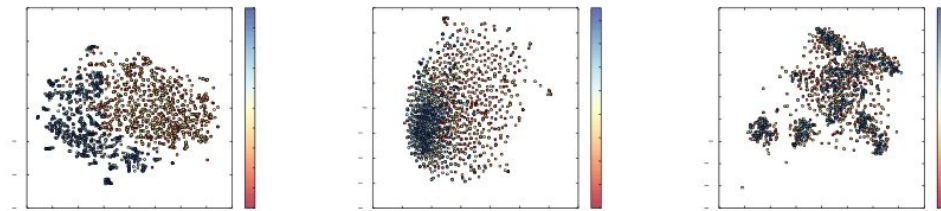
- Metrics
- Projections
- No examples

- Generally promising results
- Huge improvements in recent years
- Huge potential

Task	Method	MIRFLICKR-25K			IAPR TC-12			NUS-WIDE		
		16 bits	32 bits	64 bits	16 bits	32 bits	64 bits	16 bits	32 bits	64 bits
$I \rightarrow T$	DCMH	<b>0.7410</b>	<b>0.7465</b>	<b>0.7485</b>	<b>0.4526</b>	<b>0.4732</b>	<b>0.4844</b>	<b>0.5903</b>	<b>0.6031</b>	<b>0.6093</b>
	SePH	0.6573	0.6603	0.6616	0.4112	0.4158	0.4203	0.4787	0.4869	0.4888
	STMH	0.5921	0.5950	0.5980	0.3580	0.3732	0.3819	0.3973	0.4082	0.4153
	SCM	0.6290	0.6404	0.6480	0.3833	0.3898	0.3878	0.4650	0.4714	0.4822
	CMFH	0.5818	0.5808	0.5805	0.3683	0.3734	0.3786	0.3568	0.3624	0.3661
$T \rightarrow I$	CCA	0.5695	0.5663	0.5641	0.3345	0.3254	0.3193	0.3414	0.3336	0.3282
	DCMH	<b>0.7827</b>	<b>0.7900</b>	<b>0.7932</b>	<b>0.5185</b>	<b>0.5378</b>	<b>0.5468</b>	<b>0.6389</b>	<b>0.6511</b>	<b>0.6571</b>
	SePH	0.6480	0.6521	0.6545	0.4024	0.4074	0.4131	0.4489	0.4539	0.4587
	STMH	0.5802	0.5846	0.5855	0.3445	0.3570	0.3690	0.3607	0.3738	0.3842
	SCM	0.6195	0.6302	0.6366	0.3698	0.3734	0.3696	0.4370	0.4428	0.4504
	CMFH	0.5787	0.5774	0.5784	0.3619	0.3687	0.3769	0.3623	0.3670	0.3723
	CCA	0.5690	0.5659	0.5639	0.3340	0.3255	0.3197	0.3392	0.3320	0.3272

Table 4. MAP. The best accuracy is shown in boldface. The baselines are based on CNN-F features.

Task	Method	MIRFLICKR-25K			IAPR TC-12			NUS-WIDE		
		16 bits	32 bits	64 bits	16 bits	32 bits	64 bits	16 bits	32 bits	64 bits
$I \rightarrow T$	DCMH	<b>0.7410</b>	<b>0.7465</b>	<b>0.7485</b>	<b>0.4526</b>	<b>0.4732</b>	<b>0.4844</b>	0.5903	0.6031	0.6093
	SePH	0.7123	0.7194	0.7232	0.4442	0.4563	0.4639	<b>0.6037</b>	<b>0.6136</b>	<b>0.6211</b>
	STMH	0.6132	0.6219	0.6274	0.3775	0.4002	0.4130	0.4710	0.4864	0.4942
	SCM	0.6851	0.6921	0.7003	0.3692	0.3666	0.3802	0.5409	0.5485	0.5553
	CMFH	0.6377	0.6418	0.6451	0.4189	0.4234	0.4251	0.4900	0.5053	0.5097
$T \rightarrow I$	CCA	0.5719	0.5693	0.5672	0.3422	0.3361	0.3300	0.3604	0.3485	0.3390
	DCMH	<b>0.7827</b>	<b>0.7900</b>	<b>0.7932</b>	<b>0.5185</b>	<b>0.5378</b>	<b>0.5468</b>	<b>0.6389</b>	<b>0.6511</b>	<b>0.6571</b>
	SePH	0.7216	0.7261	0.7319	0.4423	0.4562	0.4648	0.5983	0.6025	0.6109
	STMH	0.6074	0.6153	0.6217	0.3687	0.3897	0.4044	0.4471	0.4677	0.4780
	SCM	0.6939	0.7012	0.7060	0.3453	0.3410	0.3470	0.5344	0.5412	0.5484
	CMFH	0.6365	0.6399	0.6429	0.4168	0.4212	0.4277	0.5031	0.5187	0.5225
	CCA	0.5742	0.5713	0.5691	0.3493	0.3438	0.3378	0.3614	0.3494	0.3395



(a) Inter-modal invariance preserved without adversary

(b) Inter-modal invariance preserved with adversary

(c) Inter-modal invariance and intra-modal discriminative-  
ness preserved with adversary



# PROGRESSIVE GROWING OF GANs FOR IMPROVED QUALITY, STABILITY, AND VARIATION

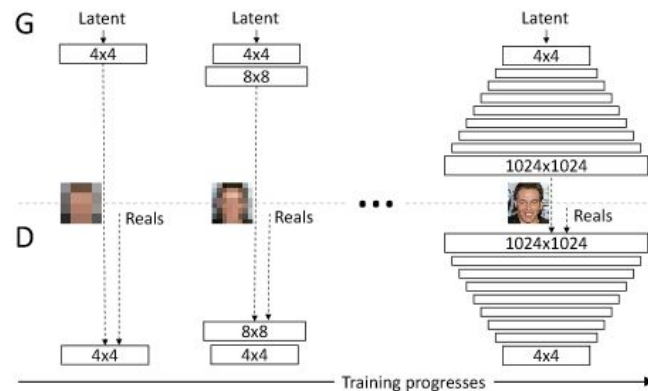
**Tero Karras**  
NVIDIA

**Timo Aila**  
NVIDIA

**Samuli Laine**  
NVIDIA

**Jaakko Lehtinen**  
NVIDIA and Aalto University

{tkarras, taila, slaine, jlehtinen}@nvidia.com



# A Style-Based Generator Architecture for Generative Adversarial Networks

Tero Karras  
NVIDIA

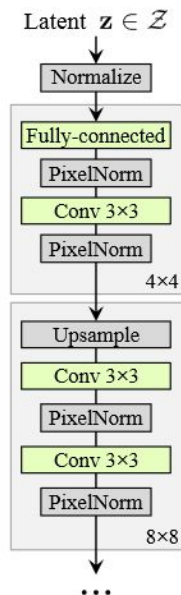
tkarras@nvidia.com

Samuli Laine  
NVIDIA

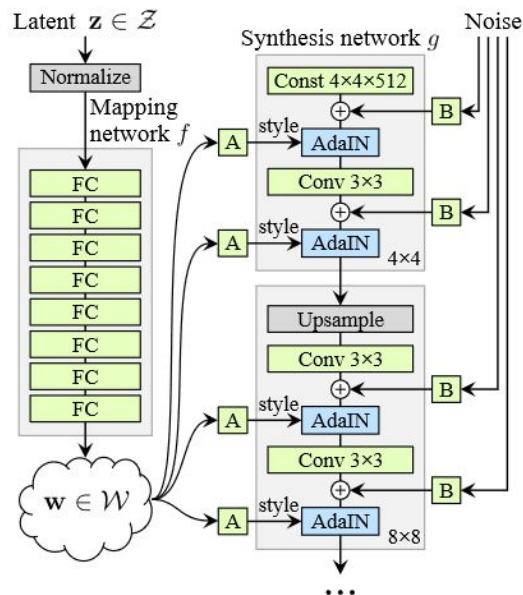
slaine@nvidia.com

Timo Aila  
NVIDIA

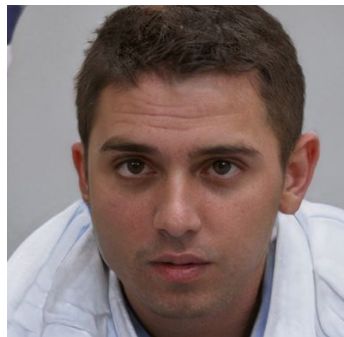
taila@nvidia.com



(a) Traditional



(b) Style-based generator



# Model generation AI





Idols created by AI

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**AI × Creative**

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## Realize a society created with AI

Until now it has been said that it is difficult for AI to do creative work.

We are developing creative AI by applying deep learning.

AI stimulates the imagination of human creators and aims to be a society where people and AI co-create.



# Final Announcements

# VDLM on Github

<https://github.com/vdlm/meetups>

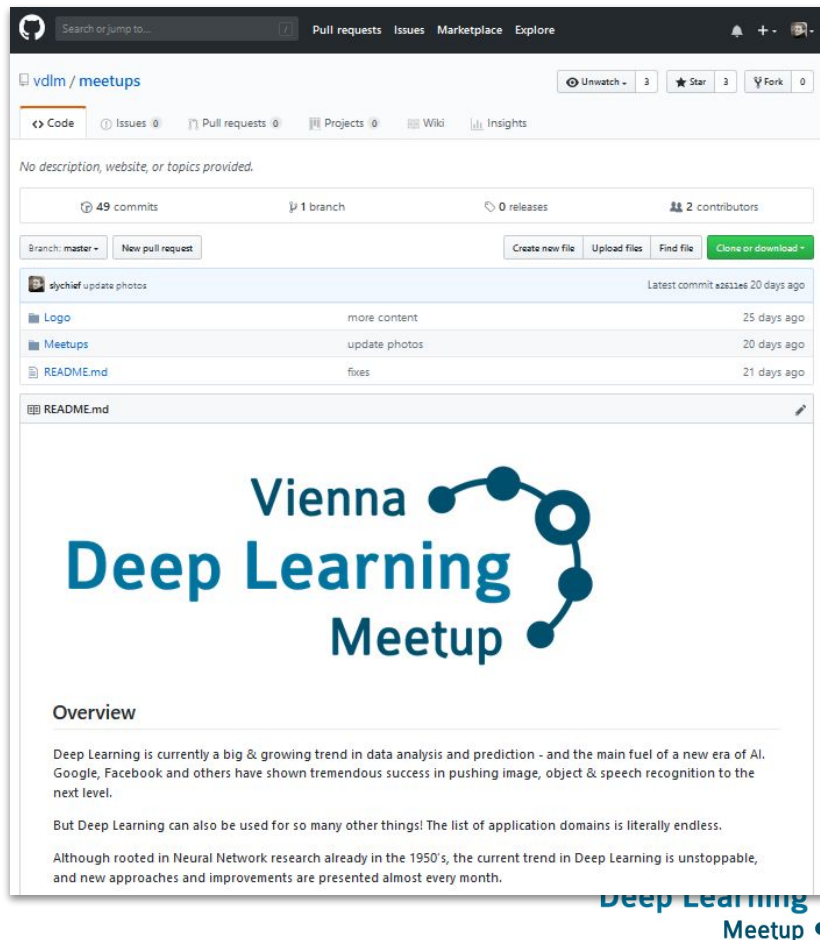
- all talks
- slides
- photos
- videos
- Wiki

## Meetups

#	Date	Place	Topic	Link	Video	Meetup.com
1	2016-04-07	Sector 5	intro	<a href="#">more</a>		<a href="#">link</a>
2	2016-05-09	Sector 5		<a href="#">more</a>		<a href="#">link</a>
3	2016-06-06	Sector 5		<a href="#">more</a>		<a href="#">link</a>
4	2016-07-07	TU Wien		<a href="#">more</a>		<a href="#">link</a>
5	2016-09-22	Automatic Software GmbH		<a href="#">more</a>		<a href="#">link</a>
6	2016-10-12	Sector 5		<a href="#">more</a>		<a href="#">link</a>
7	2016-12-01	Agentur Virtual Identity		<a href="#">more</a>		<a href="#">link</a>
8	2017-01-17	TU Wien Informatik		<a href="#">more</a>		<a href="#">link</a>
9	2017-02-21	bwin.party services (Austria) GmbH		<a href="#">more</a>		<a href="#">link</a>

## Talks

Date	MU#	Speaker	Topic	Slides
2016-04-07	1	Thomas Lidy	An overview presentation of Deep Learning	<a href="#">pdf</a>
2016-04-07	1	Jan Schlüter	History, Approaches, Applications	<a href="#">pdf</a>
2016-05-09	2	Alex Champandard	Neural Networks for Image Synthesis	
2016-05-09	2	Gregor Mitscha-Baude	Recurrent Neural Networks	<a href="#">pdf</a>
2016-06-06	3	Jan Schlüter	Open-source Deep Learning with Theano and Lasagne	<a href="#">pdf</a>
2016-09-22	5	Josef Puchinger	Deep Learning & The Future of Automation	
2016-09-22	5	Christoph Körner	Going Deeper with GoogleNet and CaffeJS	<a href="#">pdf</a>



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Logo more content 25 days ago

Meetups update photos 20 days ago

README.md fixes 21 days ago

## Vienna Deep Learning Meetup

### Overview


Deep Learning is currently a big & growing trend in data analysis and prediction - and the main fuel of a new era of AI. Google, Facebook and others have shown tremendous success in pushing image, object & speech recognition to the next level.

But Deep Learning can also be used for so many other things! The list of application domains is literally endless.

Although rooted in Neural Network research already in the 1950's, the current trend in Deep Learning is unstoppable, and new approaches and improvements are presented almost every month.

Deep Learning Meetup

# VDLM Youtube Channel




**Vienna Deep Learning Meetup**  
198 Abonnenten


VON 198 ABONNIERT

ÜBERSICHT VIDEOS PLAYLISTS KANÄLE DISKUSSION KANALINFO >


**Uploads** ALLE WIEDERGEHEN



**Ethics and Bias in Artificial Intelligence - 18th Vienna**  
2:54:03  
964 Aufrufe • vor 4 Monaten gestreamt




**Ethics and Bias in Artificial Intelligence - 18th Vienna**  
Keine Aufrufe • vor 4 Monaten




**17th Vienna Deep Learning Meetup (part 2):**  
54:49  
195 Aufrufe • vor 4 Monaten gestreamt


**BELIEBTE KANÄLE**



**Kurzgesagt – In a Nuts...**  
ABONNIEREN



**7-SEKUNDEN-RÄTSEL**  
ABONNIEREN



**Dinge Erklärt – Kurze...**  
ABONNIEREN

<https://www.youtube.com/ViennaDeepLearningMeetup>

24  
JUN

Montag, 24. Juni 2019

# 28th Deep Learning Meetup in Vienna



Veranstaltet von Alexander Schindler und 3 anderen

Von [Vienna Deep Learning Meetup](#)

Öffentliche Gruppe

Vienna  
**Deep Learning  
Meetup**



Montag, 24. Juni 2019

18:30 bis 22:30

[Zum Kalender hinzufügen](#)



Talent Garden Austria

Liechtensteinstraße 111-115 · Wien

- **Deep Learning for Electrical Biosignals and their Application in Medical Products**  
*[Franz Fürbass](#), Scientist in Biosignal Processing Group at Austrian Institute of Technology (AIT)*
- **Adversarial Machine Learning - An Introduction to Backdoor, Evasion and Inversion Attacks**  
*[Rudolf Mayer](#), Senior Researcher, SBA Research & Lector, TU Wien*

**Last before Summer !**





# Vienna Deep Learning Meetup



Next Meetup:  
24<sup>th</sup> June 2019 @ Talent Garden Austria

[www.meetup.com/Vienna-Deep-Learning-Meetup](http://www.meetup.com/Vienna-Deep-Learning-Meetup)