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**Fundamental Concepts of Generative Machine Learning**

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Graduate School of Informatics, METU, Ankara

# Meet the Instructor

## • Education

- B.Sc. EEE METU, Turkey (1997-2001)  
*B.Sc. Minor Program, Sociology METU (1998-2001)*
- M.Sc. EEE, METU, Turkey  
*on "3D Modelling and Graphics" (2001-2004)*
- Ph.D. EEE, METU, Turkey  
*on "3D Object/Surface Representation" (2004-2011)*

## • Professional

- Research and Teaching Assistant, EEE METU, (2001 - 2008)
- Visiting Phd Student, the Univ. of York (2008 - 2009)
- Applied Scientist, ASELSAN, Inc, (2009 - 2016)
- Research Associate, the Univ. of York, (2016)
- Assistant Professor, EEE Çankaya Uni. (2018 - 2021)
- Research Associate, the Univ. of York, (2022-2023)
- Associate Professor, MMI II METU (2021 - ...)

## Dr. Erdem Akagündüz

Graduate School of Informatics, METU

HOME PUBLICATIONS

### about me

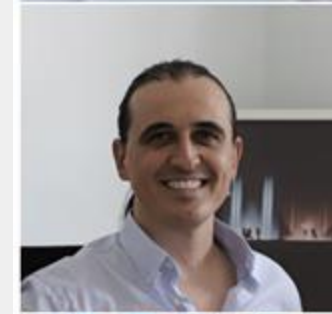


I am currently an associate professor with the [Graduate School of Informatics, METU](#). I am professionally interested in **computer vision, deep learning, pattern recognition, image processing, machine learning, object tracking, and 3D modelling**. I have several journals, conference papers, and international patents on these subjects. If you are interested, please find my [resume here](#) or check my [Google Scholar](#), [Google Patents](#), [ResearchGate](#), or [LinkedIn](#) pages for more information.

After finishing my Ph.D. at METU EEE and working as a visiting researcher at the University of York, I started working as a Computer Vision Scientist at ASELSAN. I was a part of the Image

# Our Institute

- Associate Professor, MMI II METU
  - MSc, PhD Advisorships,
  - Research Project (TUBITAK) Administrations
  - Industrial Consultancies
  - Teaching
    - DI504 Foundations of Deep Learning
    - MMI711 Sequence Models in Multimedia
    - MMI714 Generative Models in Multimedia
    - IS566 Image Processing Algorithms
    - MMI704 Human Motion Capture, Analysis and Synthesis



**Erdem Akagündüz, Assoc. Prof. Dr.**

Multimedia Informatics

**Room:** B-216, **Phone:** 7886, **Email:** akaerdem[at]metu.edu.tr

**Fields of Interest:**

Computer Vision, Deep Learning, Pattern Recognition



**Yeşim Aydın Son, Assoc. Prof. Dr. (Head of Health Informatics)**

Health Informatics

**Room:** B-207, **Phone:** 7708, **Email:** yesim[at]metu.edu.tr

**Fields of Interest:**

bioinformatics; computational biology; genomics; GWAS microarray research, personalized medicine, medical informatics, genomics, next generation sequencing, neurogenetics, molecular genetics



**Nazife Baykal, Prof. Dr.**

Information Systems

**Room:** A-206, **Phone:** 7701, **Email:** baykal[at]metu.edu.tr

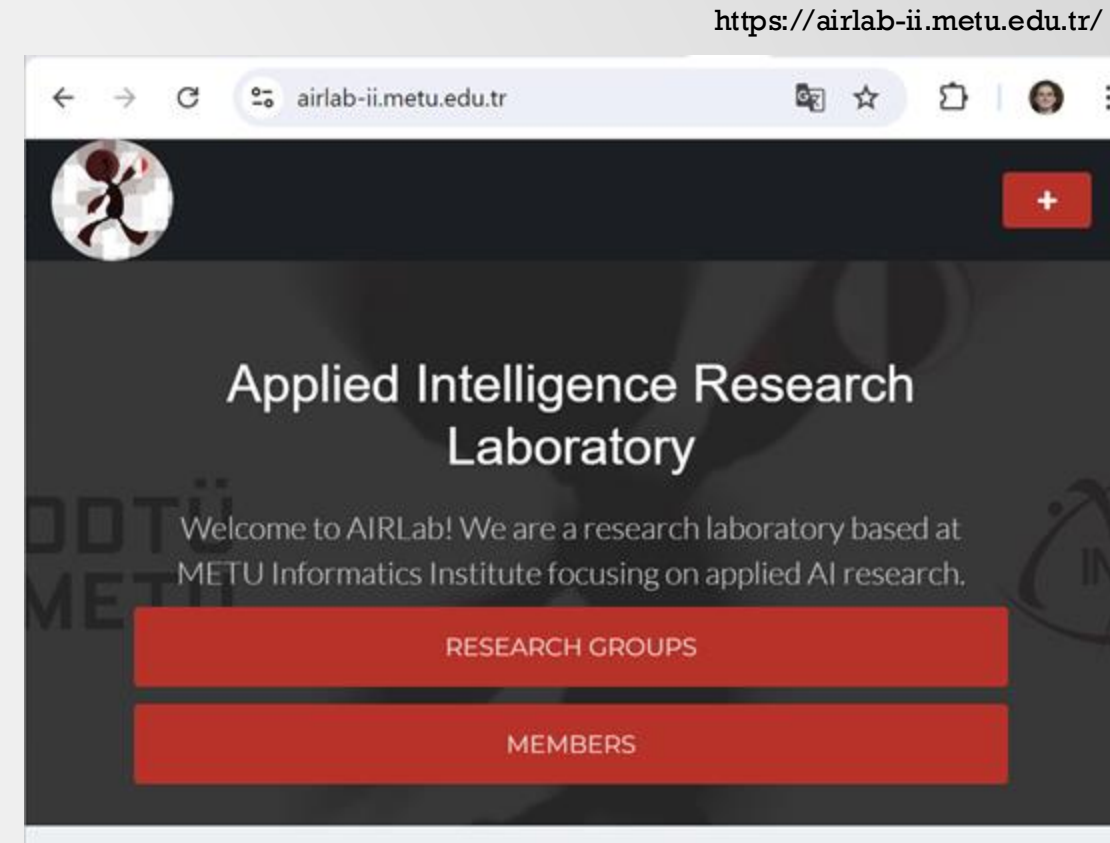
**Fields of Interest:**



**Cem Bozsahin, Prof. Dr.**

# Our Research Laboratory

- **AIRLab** is home to different research groups focusing on applying AI problems across diverse domains including, but not limited to, computer vision, geoinformatics, earthquake engineering, and decision systems.
- For comprehensive insights into specific projects, publications, and team members, please navigate to the individual pages of each research group.



# Preknowledge/Prerequisite(s)

(Must)

- Familiarity with Probability Theory
- Familiarity with Fundamental Machine Learning Concepts

(Optional/Recommended)

- Experience with Deep Neural Networks
- Experience with Information Theory

# Preknowledge/Prerequisite(s)

This course is designed for a broad audience,

- including students with basic knowledge of probability theory and machine learning,
- as well as those who are more experienced in deep learning and information theory.



# What will you learn?

## Fundamental Concepts of Generative Machine Learning

- **Foundational Concepts:** Gain a solid understanding of the core principles of generative modeling, including the necessity of modeling data as distributions within latent spaces.
- **Mathematical and Theoretical Tools:** Learn about essential mathematical concepts, evaluation techniques, and introductory information theory, such as entropy and divergence.
- **Latent Space and Generative Models:** Explore the properties of latent spaces, deep feature spaces, and their roles in self-supervised auto-encoder systems.

# Why this course?

## Fundamental Concepts of Generative Machine Learning

- Many practitioners in machine learning and deep learning often focus on model training and prediction without fully understanding the concept of generation and its connection to information theory.
- This short course is designed to bridge that gap, providing a dive into the principles of generative modeling and its critical role in understanding and shaping data distributions within the broader context of AI.



## PART I: Mathematical Background

- Generation vs. Discrimination in Machine Learning
- Data Distribution, Sampling, Inference and Generation
- Expectation and Likelihood
- Evaluation for Generative Models, Distribution Distances, Divergence and Entropy

# Course Outline



## PART II: Latent Spaces

- (Curse of) Dimensionality
- Deep Features vs. Latent Spaces
- Latent Space properties, Continuity, Entanglement, etc

## **PART III: Auto-Encoding**

- Autoencoders and Dimensionality Reduction
- Variational Inference and VAEs
- Conclusions

# What this course is

- **A Conceptual Foundation:** This course provides a deep understanding of the principles behind generative modeling, focusing on the importance of data generation through distributions.
- **Exploring Latent Spaces:** Gain insights into why modeling data within a latent space is crucial for effective generation and how it underpins various generative models.
- **Connecting with Information Theory:** Understand the relationship between generative modeling and key concepts in information theory, emphasizing why a simple distribution is often necessary.

# What this course isn't

- **Not Just Another Course on GANs, VAEs, or Diffusion Models:** This course does not focus on teaching specific generative model architectures or their technical details.
- **No Architecture-Specific Tutorials:** You won't find step-by-step guides on how to build GANs or diffusion models here.
- **Not a Tool-Centric Approach:** The course emphasizes the underlying ideas and theory behind generation, rather than the application of specific tools or techniques.

# Course Objectives

- **Understand the Core Principles:** Develop a solid grasp of the fundamental concepts in generative modeling, including the role of distributions in generating data.
- **Explore Latent Spaces:** Learn why latent spaces are essential for modeling and generation, and how they differ from deep feature spaces.
- **Evaluate Generative Models:** Understand the evaluation techniques used to assess the performance and accuracy of generative models.
- **Bridge Theory and Practice:** Connect theoretical concepts with practical applications, understanding how they underpin the generative models used in modern AI.



# Set Up/Configure/Install .....

- **No Installation Required:** This course focuses on the theoretical foundations of generative modeling, so you won't need to install any software or tools.
- **Conceptual Understanding:** Our emphasis is on understanding the underlying principles and ideas, rather than hands-on coding or model implementation.

# Thanks



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