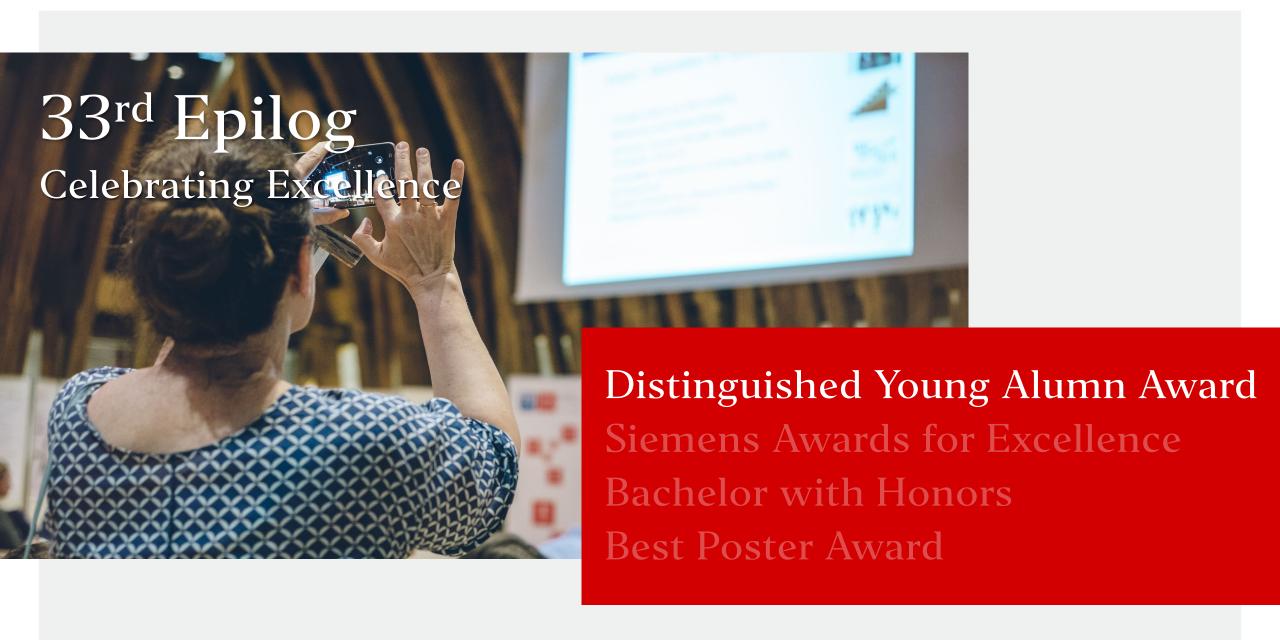


Welcome Address

- Sabine Seidler Rector, TU Wien
- Gerti Kappel Dean, TU Wien Informatics
- Host: Stefan Woltran Vice Dean of Academic Affairs, TU Wien Informatics







Axel Brunnbauer

Model-Based Reinforcement Learning for Autonomous Racing

Distinguished Young Alumn Candidate

Axel Brunnbauer

Supervisor: Radu Grosu

191-01 Cyber-Physical Systems



Tobias Geibinger

Investigating Constraint
Programming and Hybrid Answer-set
Solving for Industrial Test Laboratory
Scheduling

Distinguished Young Alumn Candidate

Tobias Geibinger

Supervisor: Nysret Musliu

192-02 Databases and Artificial

Intelligence



Viktoriia Korchemna

The Complexity of Bayesian Network Learning: Revisiting the Superstructure

Distinguished Young Alumn Candidate

Viktoriia Korchemna

Supervisor: Robert Ganian

192-01 Algorithms and

Complexity



Sebastian Mazza

Homomorphic-Encrypted Volume Rendering

Distinguished Young Alumn Candidate

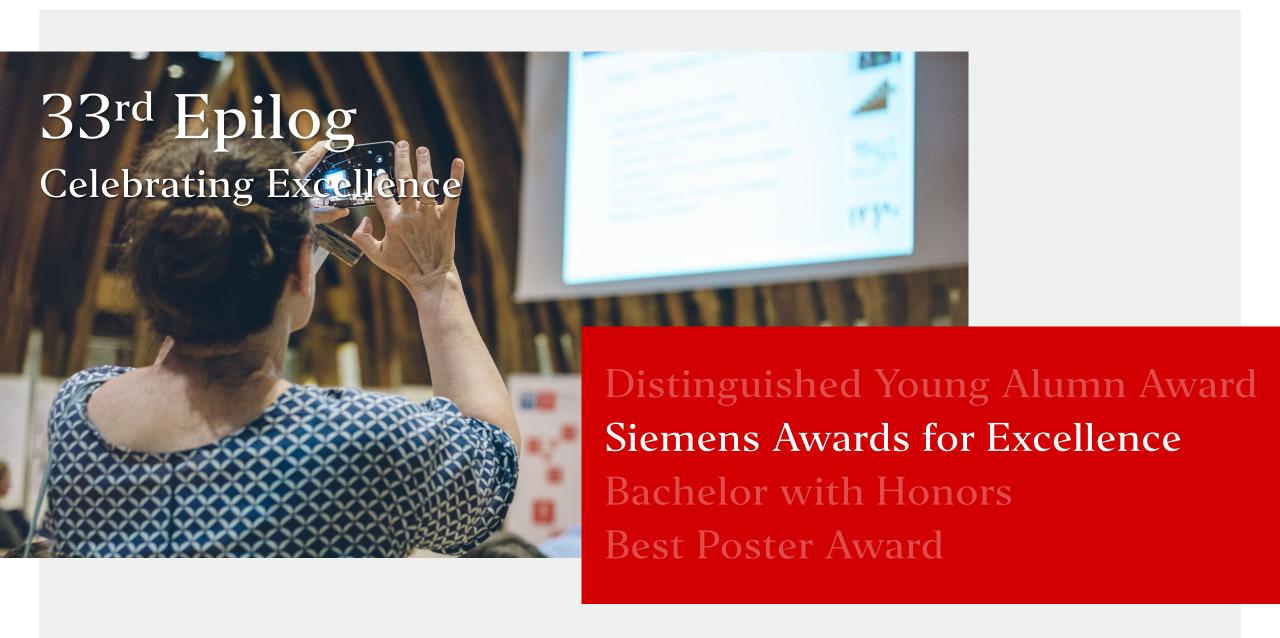
Sebastian Mazza

Supervisor: Ivan Viola

193-02 Computer Graphics









Siemens Awards for Excellence

Together with Siemens Austria we honor the best female students of the year for outstanding achievements in their studies.



Siemens Awards for Excellence

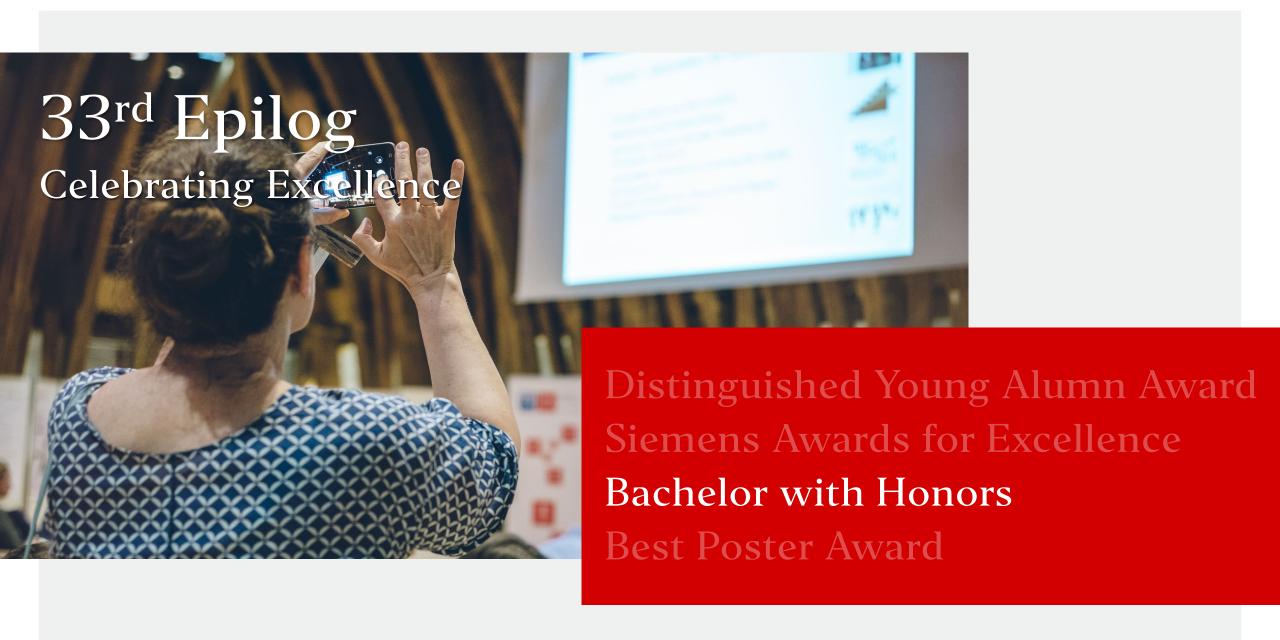
Bachelor Students

Master Students

- Lisa Fürst
- Sonja Gerstenmayer
- Sophia Schober
- Christina Tüchler
 Timea Toth

- Jasmine Haider
- Theresa Müller
- Raffaela Pichler







Bachelor with Honors

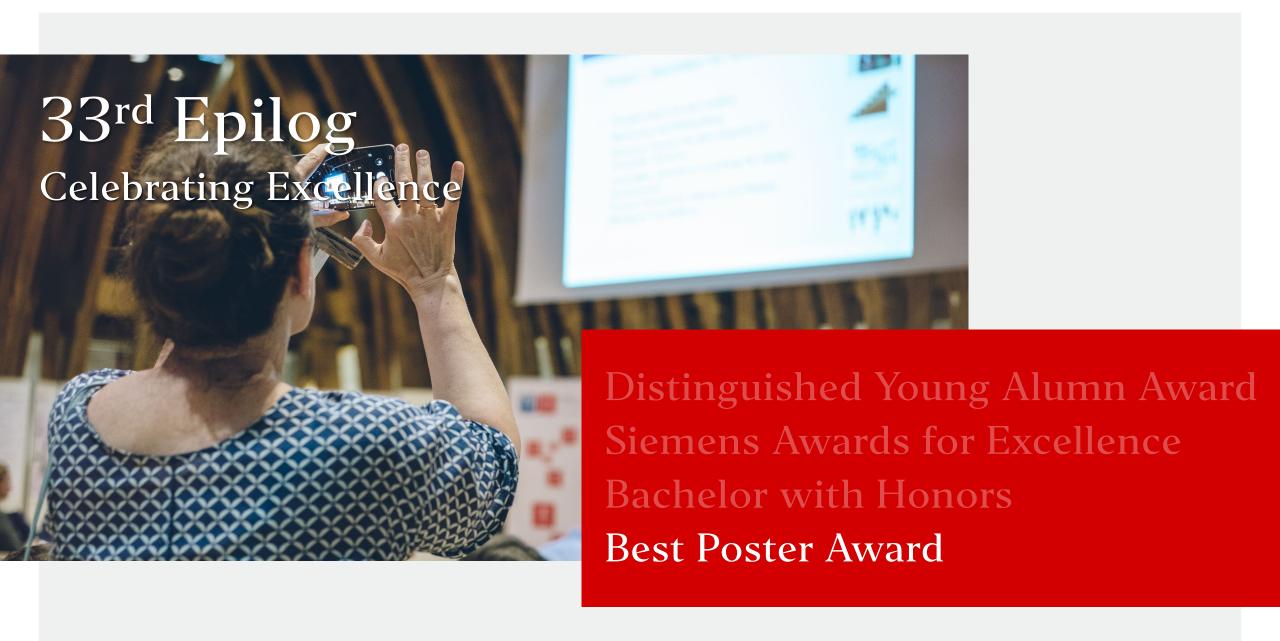
With our excellence program, we enable you to prove your potential in individually designed challenges, nurture your talents, and engage you in scientific research.



Bachelor with Honors The Graduates

- Andreas Auer
 Software and Information Engineering
- Adrian Chmurovic
 Software and Information Engineering
- Alexander Falzberger
 Software and Information Engineering
- Christoph Ullrich Günther
 Software and Information Engineering











Representation Learning for Variable-Sized Multiple Sequence Alignments

Tamara Drucks Logic and Computation

TU Wien Informatics

Institute of Logic and Computation
Databases and Artificial Intelligence Group
Supervisor: Priv.-Doz. Dr. Nysret Musliu
Co-Supervisor: Univ.-Prof. Dr. Arndt von Haeseler
Contact: tamara.drucks@aon.at

MULTIPLE WHAT?

A multiple sequence alignment¹ refers to the alignment of three or more molecular sequences (DNA, RNA or protein), aligned such that the similarity between the sequences is maximized.

human CATGGCATGTTACA...
dog CTTCGCAACTTAAA...
mouse GTTCGCAGGTACCA...
oppossum CATGGCAACTTGCT...

Multiple sequence alignments form the basis to address numerous fundamental questions arising in biology. Many such questions stem from the research field of phylogenetics'.

PHYLOGENETICS

Phylogenetics studies the evolutionary history among biological entities. The evolutionary relationships between a set of entities are typically depicted in a **phylogenetic tree**.

| human | dog | mouse | oppossum |



The reconstruction of a phylogenetic tree is based on some observable heritable traits of the given entities, such as molecular sequence data (i.e. an alignment of related sequences).

WHY LEARN REPRESENTATIONS?

Inferring phylogenies from multiple sequence alignments is hard*-3, and remains inefficient, despite heuristics, for larger alignments. Data-driven learning approaches give hope to speed up this process. Most machine learning algorithms have some input size constraint. The computation of fixed-size representations constitutes a crucial first step.

PROBLEM STATEMENT

Can we devise a framework that is able to produce semantically meaningful representations of **fixed size**, suitable as input for a task in phylogenetics, for **variable-sized** multiple sequence alignments?

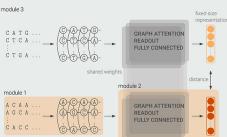
WHAT?

We devised a framework which has the ability to:

- (a) Handle alignments of variable sizes
- (b) Learn semantically meaningful fixed-size representations
- (c) Maximize the amount of extracted information

We define semantics for alignments implicitly by means of similarity, i.e. similar alignments should be embedded close to each other, while dissimilar alignments should be embedded distant from each other.

REPRESENTATION LEARNING FRAMEWORK



HOW?

The framework comprises three main modules, which implement conditions (a) – (c):

- (1) Graph transformation module
- (2) Embedding module
- (3) Training module

(1) transforms a given alignment into a graph. (2) computes a fixed-size representation using graph attention layers with average pooling as graph readout, followed by fully-connected layers. (3) implements the training procedure using a siamese neural network with contrastive loss function. Training is done with pairs.

EXPERIMENTAL EVALUATION & BENCHMARK STUDY

We conducted systematic experiments to assess the representation learning framework. The phylogenetic task of the selection of the model of sequence evolution' (mSE) serves as learning objective for a proof of concept. We simulated alignments evolving under up to four different mSEs (JC, K2P, F81 and GTR).

WHAT DID WE LEARN?

The learned representations are semantically meaningful, given our notion of semantics.

Performance generally improved with the size of the alignment, the framework thus seems to be able to maximize the extracted information.

t-SNE of alignments evolved under two (above) and three (below) mSEs show that similar alignments are close, while dissimilar alignments are distant to each other in the embedding space.



HOW DO WE DO?

We compared our classifier siamSE with the established methods¹ for model selection AIC, AICc and BIC.

mSE	siamSE	AIC	AICc	BIC
JC	94.0	71.7	79.0	98.7
K2P	95.7	95.7	97.3	99.7
F81	92.3	90.0	92.0	93.3
GTR	96.0	100.0	100.0	98.0
mean	94.5	89.3	92.0	97.4

siamSE ranks second best on average with 94.5% accuracy.

IT WORKS, WHAT'S NEXT?

The results of our empirical evaluation are promising. What are possible next steps?

- Include more complex models of sequence
 well-time.
- Provide more formal notions and guarantees
- Learn representations for a different phylogenetic task, such as e.g. tree topology
- 4. Explore other approaches for this problem
- 5. Compile benchmark dataset

 M. Steel. The phylogenetic handbook: a practical approach to phylogenetic analysis and hypothesis testing (2010).
 W.H.E. Day, D.S., Johnson, and D. Sankoff. The computational complexit of inferring rooted phylogenies by parsimory (1986) and 3. B. Chor and T. Tuller. Maximum likelihood of evolutionary trees is hard (2005).

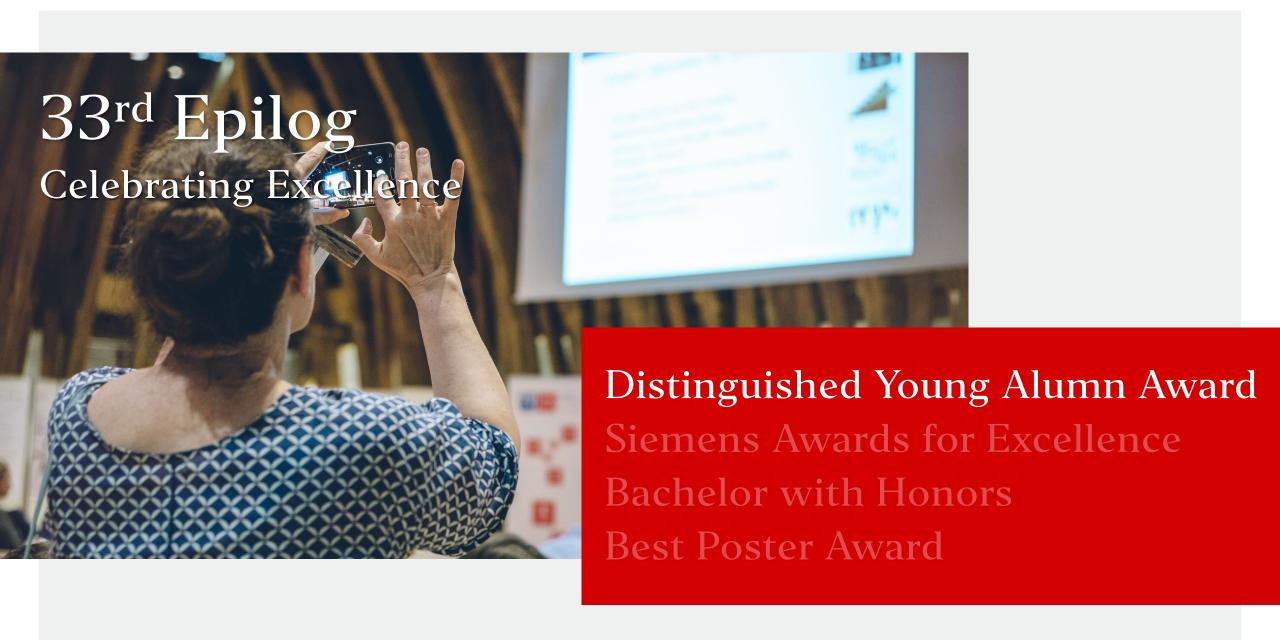
33rd Epilog

Best Poster Award Winner

Tamara Drucks

Supervisor: Nysret Musliu 192-02 Research Unit of Databases and Artificial Intelligence







Distinguished Young Alumn Award Candidates

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- Tobias Geibinger
- Viktoriia Korchemna
- Sebastian Mazza



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