

# State-of-the-art analysis of the requirements in the field of AAI

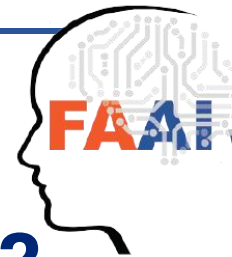
Vasyl Martsenyuk (UBB)



**FAAI:**  
The Future is In Applied Artificial Intelligence  
WP4 Teacher Training A 4.5,  
Podgorica, Montenegro, 15-19.05.2023  
(UBB Team)

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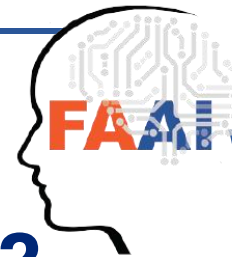


## How do skills and competencies differ?

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- **Skills** are the specific learned abilities that you need to perform a given job well. Examples, depending on the specific role, range from handling accounts and coding to welding or writing tenders.



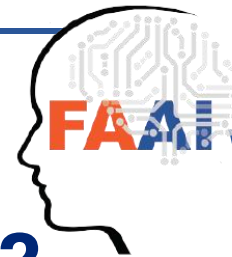


## How do skills and competencies differ?

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- There is a distinction, however, to be made between hard skills and soft skills.
- Whereas a **hard skill** is a technical and quantifiable skill that a professional may demonstrate through their specific qualifications and professional experiences, a **soft skill** is a non-technical skill that is less rooted in specific vocations.



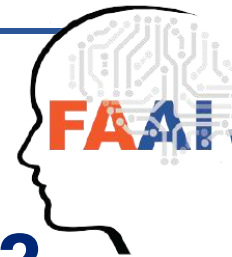


# How do skills and competencies differ?

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- An **example** of a **hard skill**,
  - then, may be **computer programming** or **proficiency in a foreign language**,
- whereas a **soft skill**
  - may be **time management** or **verbal communication**.



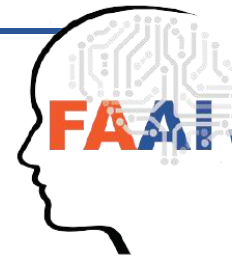


## How do skills and competencies differ?

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- **Competencies**, on the other hand, are **the person's knowledge and behaviours that lead them to be successful in a job.**
- **Examples** of competencies, then, include
  - **the improvement of business processes, strategic planning and data-based decisions.**



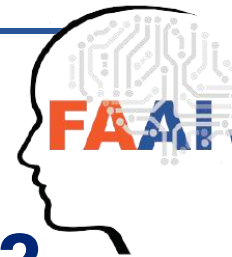


# Competence Verbs (ACM/IEEE guide)

<b>Remember – retrieve relevant knowledge from memory</b>				
Define	Duplicate	Find	Identify	Label
List	Locate	Memorize	Name	Recall
Recognize	Retrieve	Select	State	
<b>Understand – construct meaning from instructional messages</b>				
Classify	Convert	Demonstrate	Describe	Differentiate
Discuss	Exemplify	Explain	Infer	Interpret
Paraphrase	Report	Summarize	Translate	
<b>Apply – carry out or use a procedure in a given situation</b>				
Apply	Calculate	Carry out	Diagram	Edit
Execute	Illustrate	Implement	Investigate	Manipulate
Modify	Operate	Perform	Produce	Solve
Use	Write			
<b>Analyse – break material into its constituent parts and determine how the parts relate to one another and to an overall situation</b>				
Analyze	Attribute	Categorize	Compare	Contrast
Decompose	Deconstruct	Deduce	Discriminate	Distinguish
Examine	Integrate	Organize	Outline	Structure
<b>Evaluate – make judgements based on criteria and standards</b>				
Appraise	Argue	Assess	Choose	Critique
Debate	Defend	Estimate	Evaluate	Judge
Support	Test	Value	Verify	
<b>Create – put elements together to form a coherent or functional whole; reorganise elements into a new pattern or structure</b>				
Assemble	Construct	Create	Design	Develop
Devise	Formulate	Hypothesize	Invent	Make
Plan				

Table 3-1 Cognitive Processes and Associated Competence Verbs



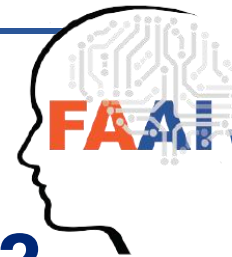


## How do skills and competencies differ?

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- Competencies effectively explain how an individual's behaviours bring about the desired results in their role.
- As with skills, there are various types of competencies – including core competencies, which are those that any successful employee requires to rise through an organisation.





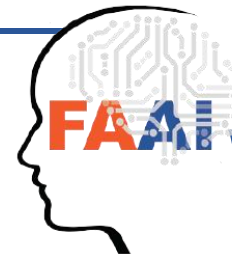
## How do skills and competencies differ?

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- In the words of marketer Aja Davis Isble, “...a core competency is something that is core to you and how you work – so it is something that could potentially set you apart from every other candidate.”





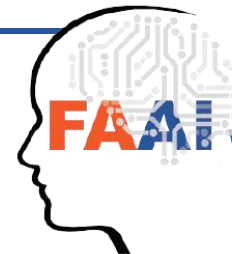


# Competency-Based Education

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- The **competency-based education** (CBE) approach allows students to advance based on their ability to master a skill or competency at their own pace **regardless of environment**.
- This method is tailored to meet different learning abilities and can lead to more efficient student outcomes.



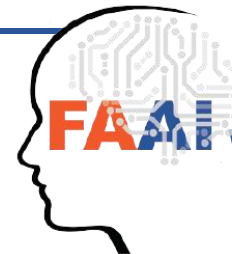


# Competency-Based Education

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- **Competency-based** or **Outcomes-based**
- Advancement based on mastery of outcomes rather than seat time or credit hours
- Mastery demonstrated through assessments



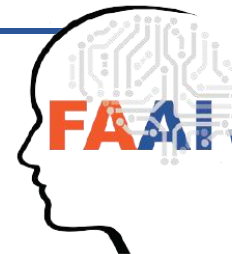


# Competency-Based Education

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- Changing paradigm:
  - Credit hour  **content mastery**
  - Focus on teaching  **focus on learning**
  - Time is constant/learning is variable  **time is variable/learning is constant**
  - **Greater focus on employer input** regarding knowledge, skill, and aptitude (KSA) needs of future employees



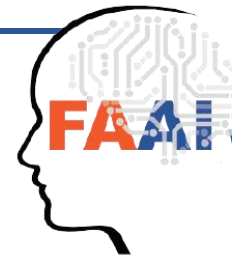


## Implementing CBE

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- **Identifying outcome competencies** is key – all supporting learning objectives must be identified and sequenced within a program and courses
- **Objectives may be used rather than outcomes** – usually require demonstration of more discrete skills or knowledge



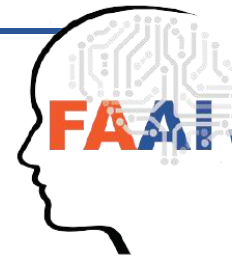


## Implementing CBE

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- Identify and **reduce or eliminate ineffective structures, practices, and expectations** that may have been historically justified but are now impeding the efficiency of student learning
  - Minimum hours & unit requirements
  - Minimum internship/externship hours requirements
  - Required skills practice regardless of current competence



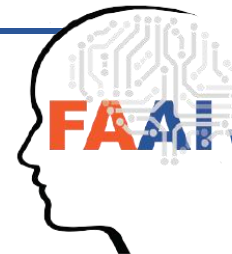


# Implementing CBE

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- Design classroom and lab experiences with preassessment opportunities
  - Based on outcomes or objectives, determine the appropriate assessment(s) and level of mastery
  - Pre-assessments may be the same as post-assessment, or the two may vary for greater demands on students
  - Pre-assessments may be integrated into early classroom activities or assignments rather than formalized as separate assessment opportunities



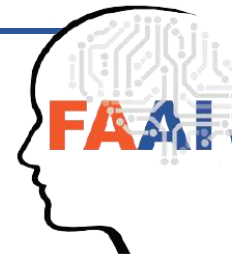


## Implementing CBE (continued)

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- Design classroom and lab experiences with preassessment opportunities
  - Instruction based on students learning, rather than faculty teaching.
  - Establish achievement benchmarks
  - Establish program exit criteria





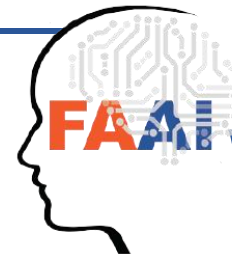
## Why CBE and why now?

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- For-credit CBE programs are a needed addition to our instructional portfolio
- CBE will help us meet our goals (Vision for Success)
- CBE is an economic mobility lever
- Career preparation is essential to the future of our state and our system
- CBE can flex with changing technologies, employer demands, and unexpected societal shifts





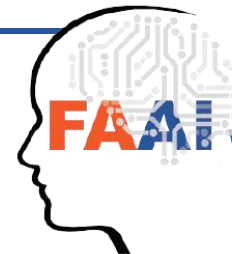


# The Equity Imperative

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- We have made substantial progress toward our goals, but gaps remain
- Noncredit serves a high number of students that are disproportionately impacted (DI)
- CBE for-credit opportunities will enhance student outcomes
- Giving students access to significant wage gains with a degree

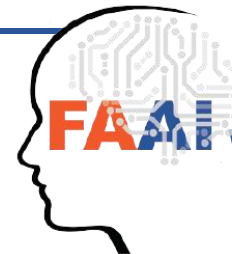




# CBE Focuses on Mastery of Competencies

- **Credit Hour Model...** based on seat time originally designed to determine faculty pensions and not as a measure of learning.
- **Credit Hour** is current basis for awarding financial aid, faculty workload & degree completion.
- **Competency Based Education...** moves beyond seat-time focusing on mastery of competencies, through learning activities and experiences that align with clearly defined programmatic outcomes.
- Hours will vary, but the **learning is fixed**.
- **Often fully** online modules.
- **Flexible** academic calendar term options.

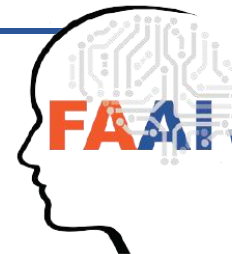




# Competency-based education is and is not....

CBE is...	CBE is not...
<b>Time is variable</b> with program offered in a flexible, self-paced approach	Learner works at course-set pace with predetermined schedule of assignments, activities, etc. with true beginning and end dates
<b>Learning is fixed</b> requiring demonstration of mastery of each competency	Learning varies with passing course grade (i.e., A, B, C, D)
Determined by rigorous <b>summative authentic assessment</b> focused on meeting core outcomes and competencies	Student evaluation varies based on accumulation of activities, exams, projects, discussion, attendance, etc.
Student <b>learning supported by faculty and staff</b> throughout learning journey	Independent study (i.e., students learn on their own and then take final exam)
Completion of program is based on mastery of <b>intentionally designed scaffold</b> of all competencies	Achievement of credential is based on passing grades and credit-hours (i.e., 60 credit hours=AA/AS)





# Computing Competencies for Undergraduate Data Science Curricula

## ACM Data Science Task Force

January 2021

- [https://www.acm.org/binaries/content/assets/education/curricula-recommendations/dstf\\_ccdsc2021.pdf](https://www.acm.org/binaries/content/assets/education/curricula-recommendations/dstf_ccdsc2021.pdf)

## Computing Competencies for Undergraduate Data Science Curricula

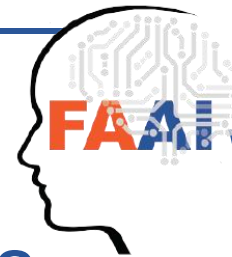
ACM Data Science Task Force

January 2021

Andrea Danyluk, Co-chair

Paul Leidig, Co-chair



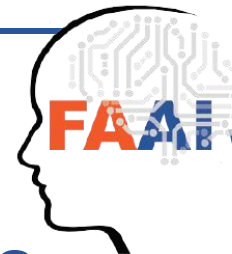


# ACM/IEEE Data Science Competencies

- Artificial Intelligence (p.51)

Scope	Competencies
<ul style="list-style-type: none"><li>• Major subfields of AI</li><li>• Representation and reasoning</li><li>• Planning and problem solving</li><li>• Ethical considerations</li></ul>	<ul style="list-style-type: none"><li>• Describe major areas of AI as well as contexts in which AI methods may be applied.</li><li>• Represent information in a logic formalism and apply relevant reasoning methods.</li><li>• Represent information in a probabilistic formalism and apply relevant reasoning methods.</li><li>• Be aware of the wide range of ethical considerations around AI systems, as well as mechanisms to mitigate problems.</li></ul>
Sub-domains	





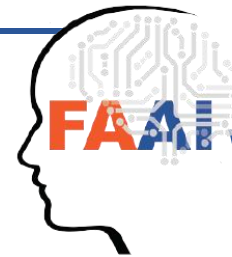
# ACM/IEEE Data Science Competencies

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- AI subdomains

AI-General – T1, T2 AI-Knowledge Representation and Reasoning (Logic-based models) – T2, E AI-Knowledge Representation and Reasoning (Probability-based models) – T1, T2, E	AI-Planning and Search Strategies – T2, E
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# Subdomain „AI-General”

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## *Knowledge*

T1:

- History of AI
- Reality of AI (what it is, what it does) versus perception
- Major subfields of AI: knowledge representation, logical and probabilistic reasoning, planning, perception, natural language processing, learning, robotics (both physical and virtual)

## *Skills*

T1:

- Explain how the origins of AI have led to the current status of AI
- Describe major branches of AI in order to recognize useful concepts and methods when needed in Data Science

T2:

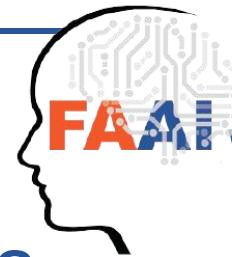
- State *what* AI systems are and that they both collect and use data to implement AI as well as collect and generate data that can be used by data scientists.
- Describe qualitatively *how* robots (physical or virtual), agents, and multi-agent systems collect and use data to embed, deliver, or implement artificial intelligence.
- Describe data collected and produced by AI systems that can be useful for data science applications.

## *Dispositions*

T1:

- Astute to, and respectful of, the fact that AI is not a new field, but rather one with a long and rich history.





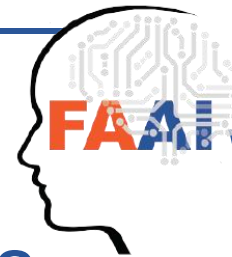
# ACM/IEEE Data Science Competencies

- Machine Learning (p.94)

Scope	Competencies
<ul style="list-style-type: none"><li>Broad categories of machine learning approaches (e.g., supervised and unsupervised).</li><li>Algorithms and tools (i.e., implementations of those algorithms) for machine learning.</li><li>Machine Learning as a set of principled algorithms (e.g., optimization algorithms), rather than as a “bag of tricks.”</li><li>Challenges (e.g., overfitting) and techniques for approaching those challenges.</li><li>Performance metrics.</li><li>Training and testing methodology.</li><li>Algorithmic and data bias, integrity of data, and professional responsibility for fielding learned models.</li></ul>	<ul style="list-style-type: none"><li>Recognize the breadth and utility of machine learning methods</li><li>Compare and contrast machine learning methods</li><li>Select appropriate (classes of) machine learning methods for specific problems.</li><li>Use appropriate training and testing methodologies when deploying machine learning algorithms.</li><li>Explain methods to mitigate the effects of overfitting and curse of dimensionality in the context of machine learning algorithms.</li><li>Identify an appropriate performance metric for evaluating machine learning algorithms/tools for a given problem.</li><li>Recognize problems related to algorithmic and data bias, as well as privacy and integrity of data.</li><li>Debate the possible effects -- both positive and negative -- of decisions arising from machine learning conclusions.</li></ul>





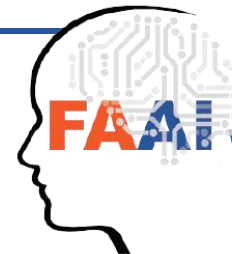


# ACM/IEEE Data Science Competencies

- ML Subdomains

Sub-domains	
ML-General – T1, T2, E ML-Supervised Learning – T1, T2, E ML-Unsupervised Learning – T1, T2, E ML-Mixed Methods – E ML-Deep Learning – T1, T2, E	Note that Reinforcement Learning appears in AI-Knowledge Representation and Reasoning (Probability-based Models)





# State-of-the-art analysis of the requirements in the field of AAI

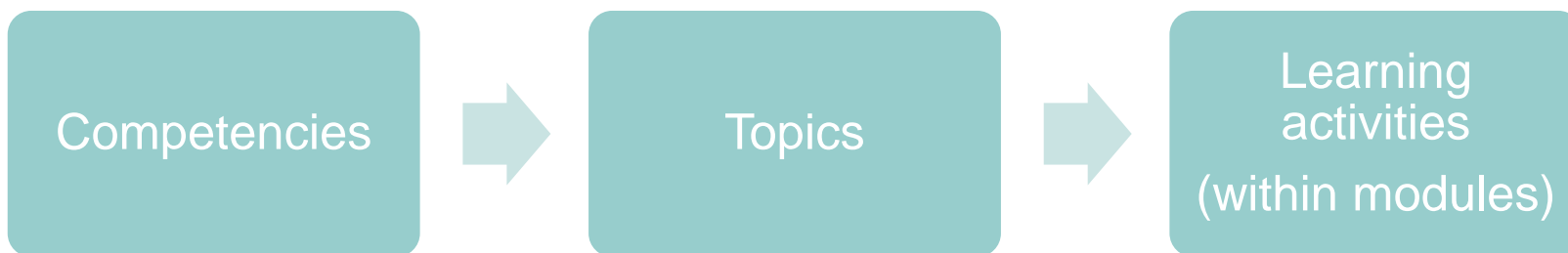
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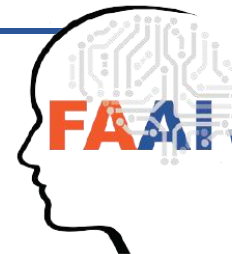
- Methodology:
  - System analysis
  - Decision making
  - Coping with uncertainties



## Studies based on WP2

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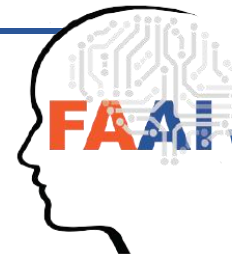


# Study of Competencies

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- Research 2: job market
- Research 4: academics
- Research 5: students
- Research 6: employers



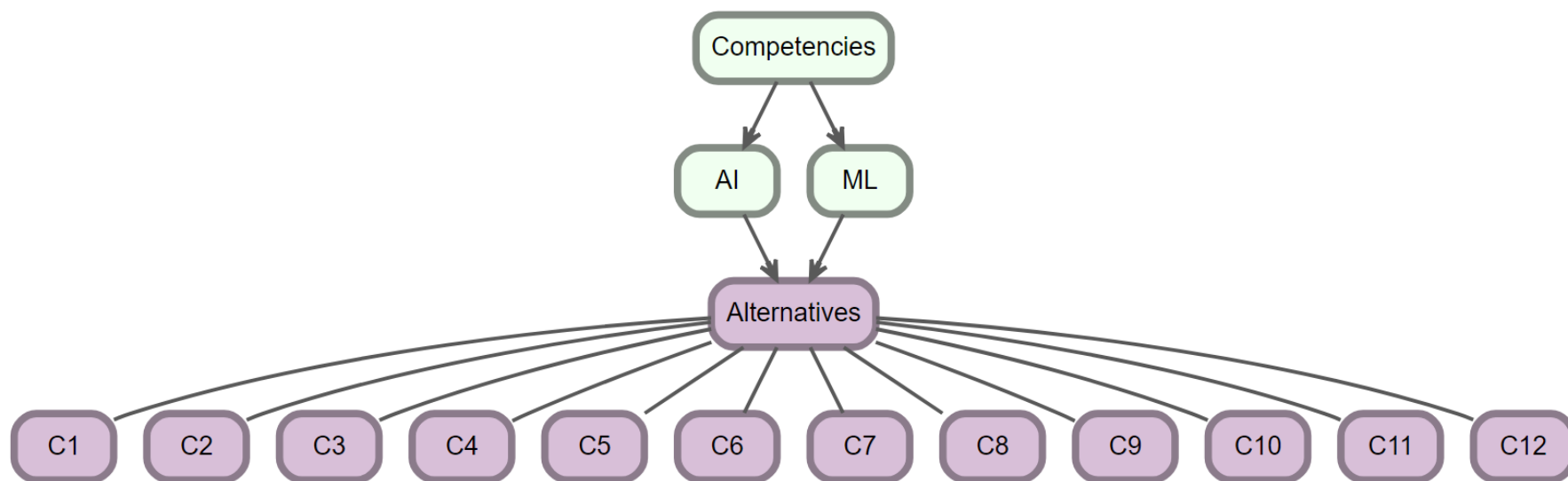


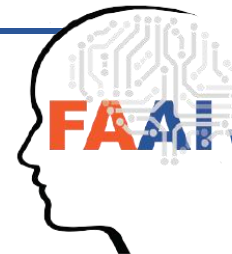
# Competencies on AI and ML from ACM/IEEE standards

- **C1.** Describe major areas of AI as well as contexts in which AI methods may be applied.
- **C2.** Represent information in a logic formalism and apply relevant reasoning methods.
- **C3.** Represent information in a probabilistic formalism and apply relevant reasoning methods.
- **C4.** Be aware of the wide range of ethical considerations around AI systems, as well as mechanisms to mitigate problems.
- **C5.** Recognize the breadth and utility of machine learning methods.
- **C6.** Compare and contrast machine learning methods.
- **C7.** Select appropriate (classes of) machine learning methods for specific problems.
- **C8.** Use appropriate training and testing methodologies when deploying machine learning algorithms.
- **C9.** Explain methods to mitigate the effects of overfitting and curse of dimensionality in the context of machine learning algorithms.
- **C10.** Identify an appropriate performance metric for evaluating machine learning algorithms/tools for a given problem.
- **C11.** Recognize problems related to algorithmic and data bias, as well as privacy and integrity of data.
- **C12.** Debate the possible effects -- both positive and negative -- of decisions arising from machine learning conclusions.



# AHP Method for AI+ML competencies



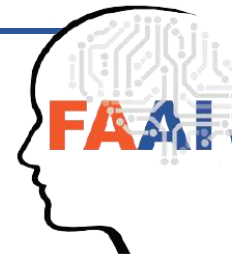


## Decision makers:

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- Job market
- Academics
- Students
- Employers





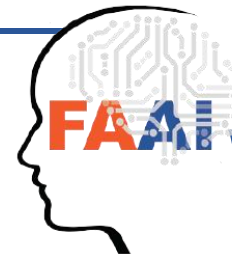
## Decision-makers:

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- #optional node, needed only if not all decision-makers have equal voting power
  - - Job market: 0.3
  - - Academics: 0.2
  - - Students: 0.2
  - - Employers: 0.3





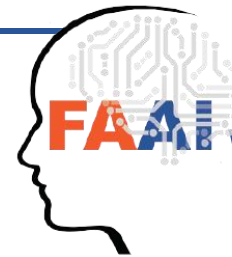


## Preferences:

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- Job market: pairwise:
  - - [AI, ML, 1/5]
- Academics: pairwise:
  - - [AI, ML, 1]
- Students: pairwise:
  - - [AI, ML, 1/3]
- Employers: pairwise:
  - - [AI, ML, 1/5]



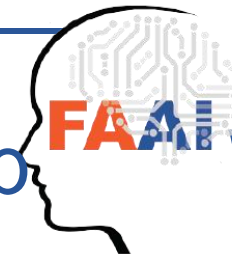


# AHP Priority Calculation Method: (eigenvalues, total contribution, all DM)

	Weight	C7	C6	C5	C8	C10	C11	C9	C12	C1	C2	C3	C4	Inconsistency
Competencies	100.0%	10.2%	10.0%	9.7%	9.6%	9.6%	9.5%	8.7%	8.3%	6.4%	6.2%	6.1%	5.7%	0.0%
ML	75.0%	9.3%	9.1%	8.7%	8.7%	8.7%	8.6%	7.8%	7.3%	1.7%	1.7%	1.7%	1.7%	0.6%
AI	25.0%	0.9%	0.9%	0.9%	0.9%	0.9%	0.9%	0.9%	0.9%	4.7%	4.6%	4.4%	4.0%	0.0%

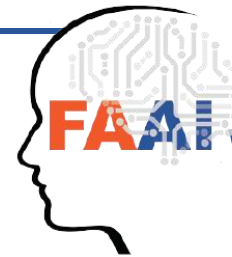


# AHP Priority Calculation Method: (eigenvalues, total contribution, DM=job market)



	Weight	C7	C6	C5	C8	C10	C11	C9	C12	C1	C2	C3	C4	Inconsistency
Competencies	100.0%	12.7%	11.5%	10.5%	10.5%	10.2%	9.1%	8.4%	8.0%	5.1%	5.1%	4.6%	4.2%	0.0%
ML	83.3%	12.0%	10.9%	9.9%	9.9%	9.6%	8.5%	7.7%	7.3%	1.9%	1.9%	1.9%	1.9%	0.6%
AI	16.7%	0.6%	0.6%	0.6%	0.6%	0.6%	0.6%	0.6%	0.6%	3.2%	3.3%	2.7%	2.4%	0.0%

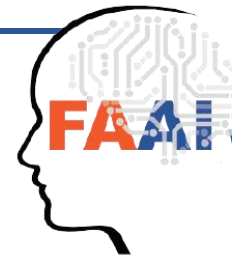




# AHP Priority Calculation Method: (eigenvalues, total contribution, DM=Academics)

	Weight	C7	C6	C5	C8	C10	C11	C9	C12	C1	C2	C3	C4	Inconsistency
Competencies	100.0%	7.6%	7.6%	7.6%	7.6%	7.4%	7.4%	7.3%	7.2%	10.4%	10.0%	10.0%	9.9%	0.0%
AI	50.0%	1.8%	1.8%	1.8%	1.8%	1.8%	1.8%	1.8%	1.8%	9.2%	8.8%	8.9%	8.8%	0.0%
ML	50.0%	5.9%	5.8%	5.8%	5.8%	5.6%	5.6%	5.5%	5.4%	1.1%	1.1%	1.1%	1.1%	0.0%

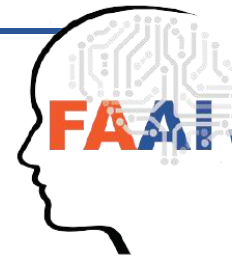




# AHP Priority Calculation Method: (eigenvalues, total contribution, DM=Students)

	Weight	C7	C6	C5	C8	C10	C11	C9	C12	C1	C2	C3	C4	Inconsistency
Competencies	100.0%	9.5%	9.4%	9.5%	9.6%	9.4%	9.5%	9.1%	9.3%	6.2%	6.2%	6.1%	6.2%	0.0%
ML	75.0%	8.6%	8.5%	8.6%	8.7%	8.5%	8.6%	8.2%	8.4%	1.7%	1.7%	1.7%	1.7%	0.0%
AI	25.0%	0.9%	0.9%	0.9%	0.9%	0.9%	0.9%	0.9%	0.9%	4.5%	4.5%	4.4%	4.5%	0.0%



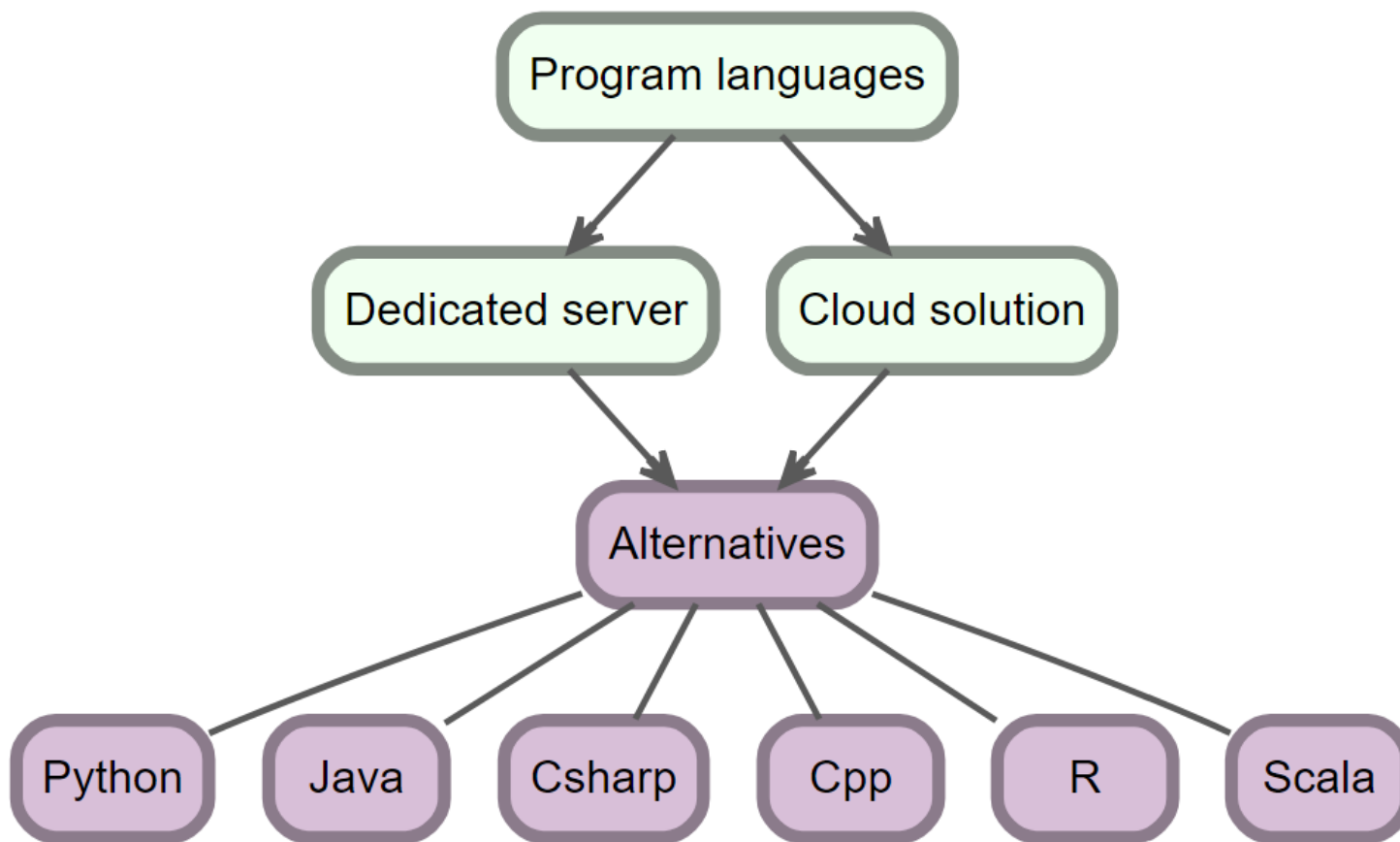


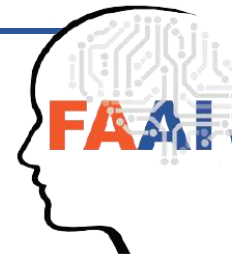
# AHP Priority Calculation Method: (eigenvalues, total contribution, DM=Employers)

	Weight	C7	C6	C5	C8	C10	C11	C9	C12	C1	C2	C3	C4	Inconsistency
Competencies	100.0%	10.1%	10.6%	10.3%	10.1%	10.6%	11.3%	9.5%	8.3%	5.1%	4.8%	5.0%	4.4%	0.0%
ML	83.3%	9.4%	10.0%	9.7%	9.4%	10.0%	10.7%	8.9%	7.6%	1.9%	1.9%	1.9%	1.9%	0.2%
AI	16.7%	0.6%	0.6%	0.6%	0.6%	0.6%	0.6%	0.6%	0.6%	3.3%	2.9%	3.1%	2.5%	0.0%



# Programming languages relative to dedicated server and cloud solution





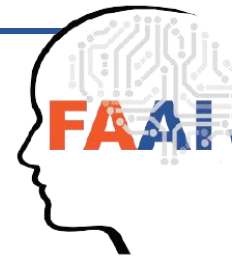
# Programming languages relative to dedicated server and cloud solution

- Job Market

	Weight	Python	Java	Csharp	Cpp	R	Scala	Inconsistency
Program languages	100.0%	44.5%	12.0%	14.2%	10.5%	11.3%	7.4%	0.0%
Dedicated server	50.0%	20.7%	4.4%	5.2%	8.0%	8.8%	2.8%	0.6%
Cloud solution	50.0%	23.8%	7.6%	9.0%	2.5%	2.5%	4.6%	4.4%





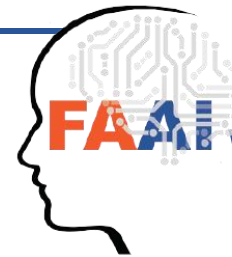


# Programming languages relative to dedicated server and cloud solution

- Employers

	Weight	Python	Java	Csharp	Cpp	R	Scala	Inconsistency
Program languages	100.0%	34.7%	21.8%	16.1%	12.0%	10.8%	4.5%	0.0%
Dedicated server	50.0%	15.8%	8.9%	5.9%	9.4%	8.1%	1.9%	1.7%
Cloud solution	50.0%	18.9%	12.9%	10.2%	2.6%	2.6%	2.6%	2.9%



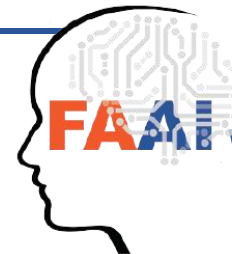


# Programming languages relative to dedicated server and cloud solution

- Good Practice

	Weight	Python	Java	Csharp	Cpp	R	Scala	Inconsistency
Program languages	100.0%	42.3%	21.0%	4.8%	10.8%	7.6%	13.5%	0.0%
Cloud solution	54.5%	23.6%	13.2%	3.0%	3.0%	3.0%	8.9%	4.0%
Dedicated server	45.5%	18.7%	7.8%	1.8%	7.8%	4.7%	4.7%	2.8%



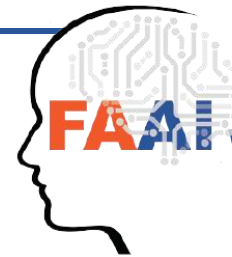


# Programming languages relative to dedicated server and cloud solution

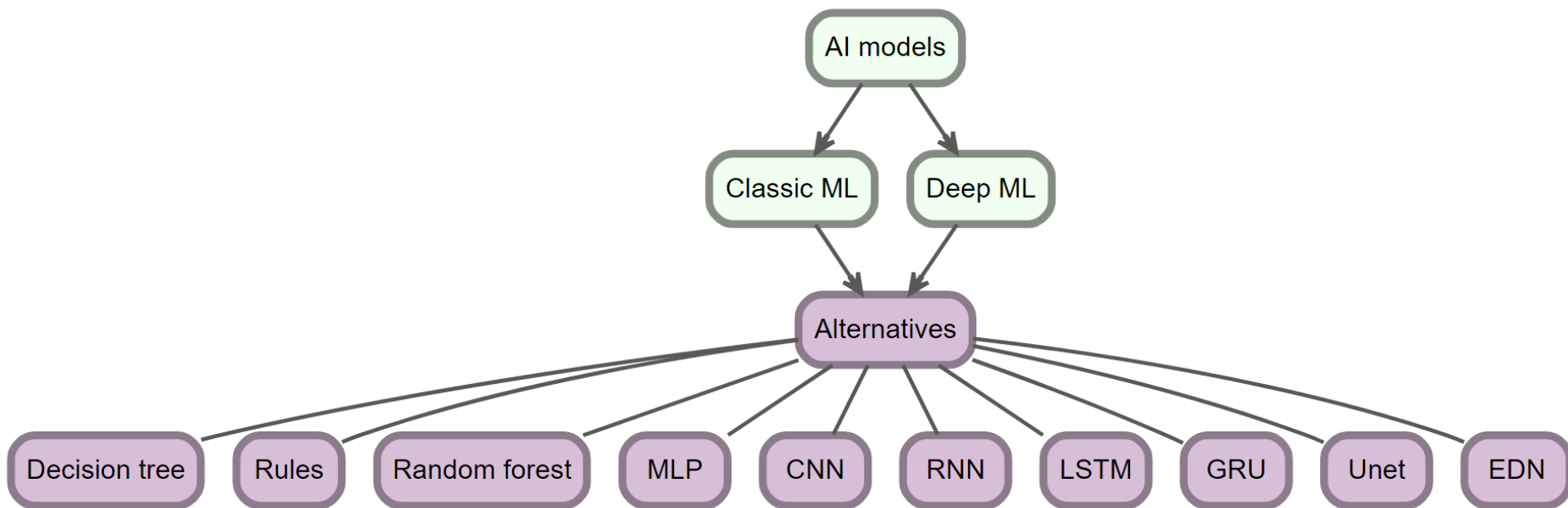
- total

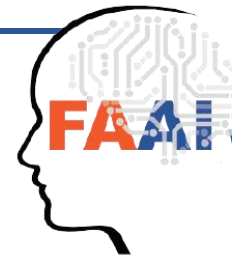
	Weight	Python	Java	Csharp	Cpp	R	Scala	Inconsistency
Program languages	100.0%	40.7%	18.6%	11.1%	11.1%	9.6%	9.0%	0.0%
Cloud solution	51.8%	22.2%	11.4%	7.1%	2.7%	2.7%	5.6%	4.4%
Dedicated server	48.2%	18.5%	7.2%	4.0%	8.3%	6.9%	3.3%	2.8%





# Artificial Intelligence models within the framework of Classic ML and Deep ML



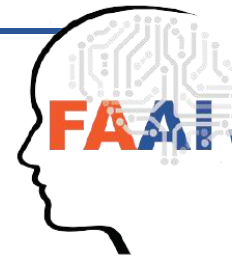


# Artificial Intelligence models within the framework of Classic ML and Deep ML

- Job market

	Weight	CNN	RNN	MLP	Decision tree	Random forest	Rules	EDN	Unet	LSTM	GRU	Inconsistency
AI models	100.0%	13.9%	12.7%	13.4%	12.1%	11.0%	12.5%	8.1%	7.0%	4.4%	5.0%	0.0%
Classic ML	54.8%	2.1%	2.1%	11.7%	10.4%	9.3%	10.8%	2.1%	2.1%	2.1%	2.1%	0.3%
Deep ML	45.2%	11.8%	10.6%	1.7%	1.7%	1.7%	1.7%	6.0%	4.9%	2.3%	2.9%	1.2%



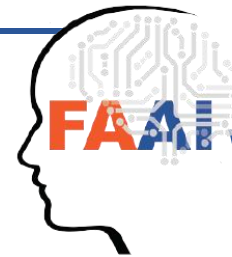


# Artificial Intelligence models within the framework of Classic ML and Deep ML

- Employers

	Weight	CNN	RNN	MLP	Decision tree	Random forest	Rules	EDN	Unet	LSTM	GRU	Inconsistency
AI models	100.0%	13.1%	10.5%	11.3%	11.7%	10.2%	11.3%	9.2%	7.1%	8.7%	7.1%	0.0%
Classic ML	50.0%	1.9%	1.9%	9.8%	10.2%	8.7%	9.8%	1.9%	1.9%	1.9%	1.9%	0.2%
Deep ML	50.0%	11.2%	8.6%	1.5%	1.5%	1.5%	1.5%	7.3%	5.2%	6.8%	5.2%	1.4%



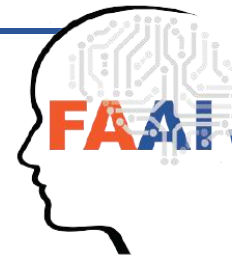


# Artificial Intelligence models within the framework of Classic ML and Deep ML

- Good Practice

	Weight	CNN	RNN	MLP	Decision tree	Random forest	Rules	EDN	Unet	LSTM	GRU	Inconsistency
AI models	100.0%	26.7%	17.2%	9.5%	9.5%	9.5%	5.5%	5.5%	5.5%	5.5%	5.5%	0.0%
Deep ML	77.8%	25.7%	16.2%	4.5%	4.5%	4.5%	4.5%	4.5%	4.5%	4.5%	4.5%	0.6%
Classic ML	22.2%	1.0%	1.0%	5.1%	5.1%	5.1%	1.0%	1.0%	1.0%	1.0%	1.0%	0.0%





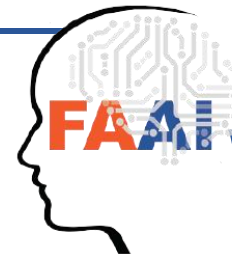
# Artificial Intelligence models within the framework of Classic ML and Deep ML

- Total

	Weight	CNN	RNN	MLP	Decision tree	Random forest	Rules	EDN	Unet	LSTM	GRU	Inconsistency
AI models	100.0%	18.2%	13.9%	11.2%	11.0%	10.4%	8.1%	8.0%	6.8%	6.4%	6.0%	0.0%
Deep ML	59.7%	16.6%	12.2%	2.6%	2.6%	2.6%	2.6%	6.3%	5.2%	4.7%	4.4%	1.4%
Classic ML	40.3%	1.7%	1.7%	8.6%	8.4%	7.8%	5.5%	1.7%	1.7%	1.7%	1.7%	0.3%





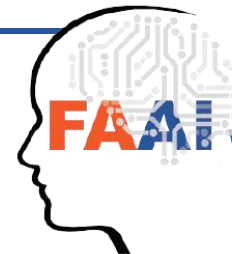


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- [https://www.acm.org/binaries/content/assets/education/curricula-recommendations/dstf\\_ccdsc2021.pdf](https://www.acm.org/binaries/content/assets/education/curricula-recommendations/dstf_ccdsc2021.pdf)





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Thank you for attention!  
Question??

