

# Al Lesson Design & Implementation in the Biology Classroom



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## Background

Transdisciplinary summer research in the application of biologically-inspired computing systems to cancer detection provided contextual background for development of a series of 10 lessons integrating artificial intelligence (AI) and computer science content knowledge with both state-specific and NGSS high school biology standards.

#### **Real-World Problem**

- Only 40% of colorectal cancers (CRC) are detected at an early stage.
- Traditional imaging techniques during clinical exam have a high miss rate, with only 59 – 84% of cancerous lesions distinguished from non-cancerous polyps.
- Early detection of CRC is associated with a good prognosis: >90% 5-year survival rate.<sup>1</sup>





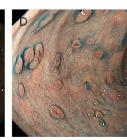


FIGURE 1 – Ignjatovic, et al, 2011. Endoscope images of colon polyps obtained using various techniques.

### Research Focus: Al-Assisted Cancer Detection

 Coupling a novel imaging technique, excitation-scanning hyperspectral imaging, with an AI model trained to distinguish normal from lesional tissue during clinical examination (endoscopy).

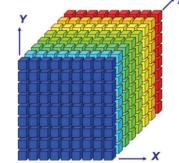


FIGURE 2 – Garini, et al, 2006. Description of spectral image dataset.

 Normal and lesional tissue samples from patients with CRC were evaluated with the novel imaging technique to create a spectral image database for use in subsequent training of an Al model.<sup>3</sup>

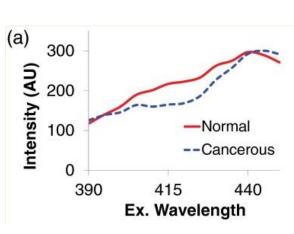


FIGURE 3 – Leavesley, et al, 2016. Fluorescence excitation scan normal tissue (solid red line) and cancerous tissue (dashed blue line) create a "spectral signature."

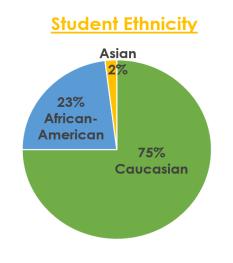
### Introduction

### **School & Course Context**

### 9<sup>th</sup> Grade Honors Biology

41 Students

15 Males 26 Female



- Large, suburban high school with 1100 students, equally distributed across grades 9 – 12.
- Title I school, with 26% economically disadvantaged students
- Follow block scheduling with semester-long courses.

### Goals of Al-Integrated Lesson Sequence

#### **Primary Goals**

### Students will:

- examine how biology has influenced computer science & how computer science/engineering have influenced the field of biology.
- 2. use projects & inquiry-based lessons to understand the emerging field of AI including learning key terminology, examining computing principles relevant to AI's functioning, and investigating biologically relevant applications of AI.
- 3. examine use of imaging in diverse contexts as well as the technology behind capturing and storing digital images.
- understand the properties of & role of light in the function of both biological systems & technological tools.

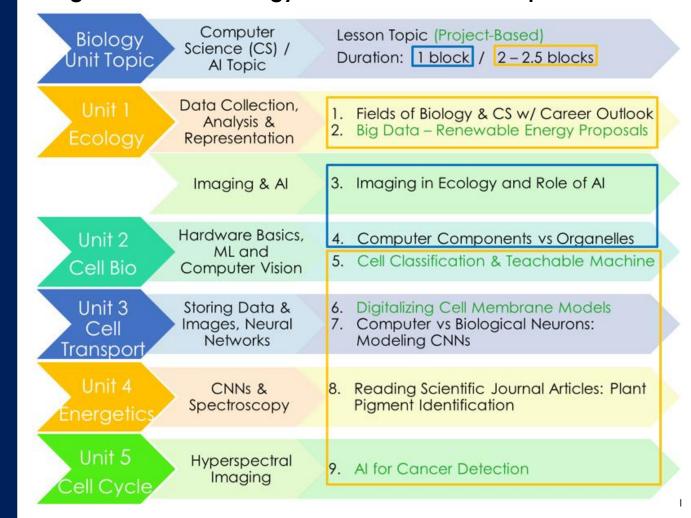
### Secondary Goals

### Students will:

- 1. strengthen basic technology skills by using a range of software applications.
- 2. use NGSS science and engineering practices.

### Methods

### Integrated AI and Biology Curriculum Lesson Sequence



- Lessons followed a 5E format.
- Implemented weekly or every 2 weeks.
- Sequence was progressive: lessons 1 4 provided foundational knowledge, lessons 5 7 were deepening lessons, and lessons 8 9 challenged and extended student learning.

| Goal #                               | 1                | 2                      | 3          | 4          |
|--------------------------------------|------------------|------------------------|------------|------------|
| Lessons Addressing<br>Primary Goal   | 1, 2,<br>3, 5, 7 | 2, 3, 5, 6,<br>7, 8, 9 | 3, 5, 6, 9 | 3, 6, 8, 9 |
| Lessons Addressing<br>Secondary Goal | 1, 2,<br>3, 5, 6 | All<br>Lessons         |            |            |

## Results

- Change of students' biology knowledge was not significant.
- Students' Al knowledge showed a statistically significant improvement at the end of the lessons.

|  | Paired Sample T-test |               | Mean   | N  | Std.<br>Deviation | t     | p      |
|--|----------------------|---------------|--------|----|-------------------|-------|--------|
|  | Pair 1               | Al Pre-Test   | 7.1207 | 29 | 1.79610           | -6.48 | <.001* |
|  |                      | Al Post-Test  | 8.8103 | 29 | 1.94300           |       |        |
|  | Pair 2               | Bio Pre-Test  | 2.2069 | 29 | 1.1626            | .424  | .675   |
|  |                      | Bio Post-Test | 2.0690 | 29 | 1.39977           |       |        |

### Conclusions

#### **Student Feedback**

26 students provided a reflection at the end of 4th quarter.

### Factors with potential negative Impact on student responses:

- Reflections were collected several weeks after the Allessons concluded.
- Timing of the reflection was at the end of the school year and right before exams.

#### Potential benefit of student reflection timing:

 Responses may provide a more accurate reflection of longer-term retention and insight into what components of the lesson sequence were most impactful for students.

### Student reflections reported the following:

- an increased understanding of how computers store information, especially image data.
- the prevalence and uses of AI, including that "AI can be used to solve problems."
- the necessity of "training" AI models and the "hierarchy" of AI, distinguishing between Machine Learning and Deep Learning.
- enjoyment in developing and sharing presentations, physically modeling an artificial neural network, comparing biological and artificial neurons, as well as training and testing Teachable Machine models.
- enjoyed all projects even though the concepts were challenging.

#### **Future Directions**

- Design and implementation of a proficiency scale for Al topics.
- Creation of student portfolio for Al projects.
- Increased teacher-student feedback on final lesson products.

### <u>Acknowledgement</u>

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### **Bibliography**

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