

A Message to CourseKata

Edmond Niu, Sarah Tran, Amaris Man, Sherry He



The Problem

- ❖ **Most students have an accurate confidence level in what they learned in the previous chapter**
 - Their confidence level reflects how well they did in the previous End of Chapter (EOC) review
 - Students who reported higher confidence (4/5) tend to perform better on EOC (>0.65 accuracy)
 - Students who reported lower confidence (0/1) tend to perform worse on EOC (<0.6 accuracy)
- ❖ **There is a subset of individuals that overestimated their confidence and performance on the EOC**
 - 56% percent of students estimated that they would perform better than they did
 - Dunning-Kruger Effect

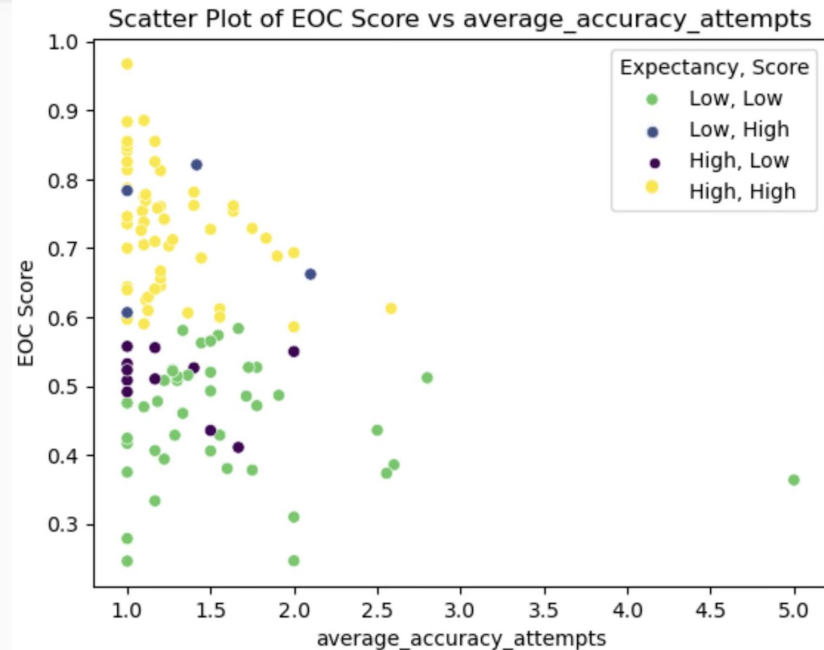
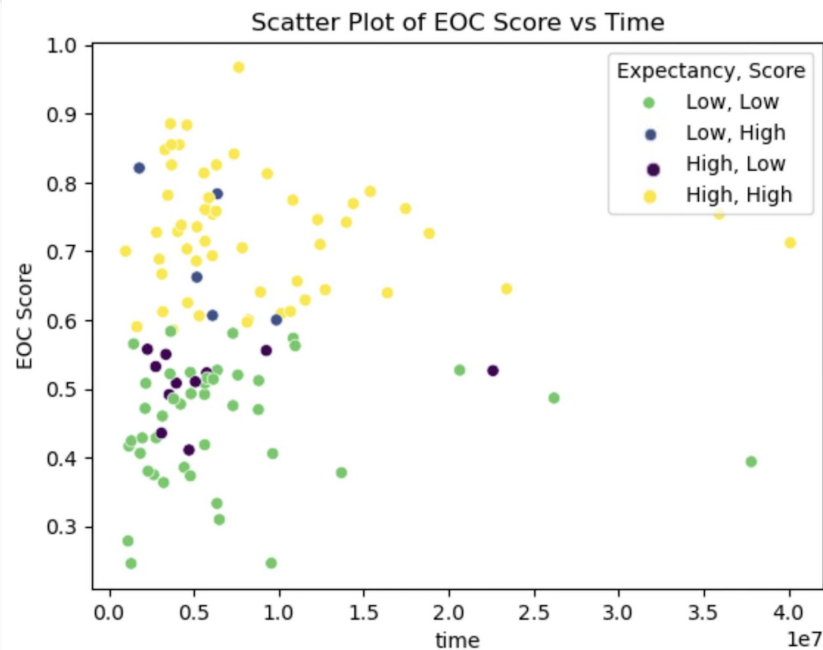
The Dunning-Kruger Effect

***The Dunning-Kruger effect** is a cognitive bias in which people with limited competence in a particular domain overestimate their abilities. Some researchers also include the opposite effect for high performers: their tendency to underestimate their skills.*

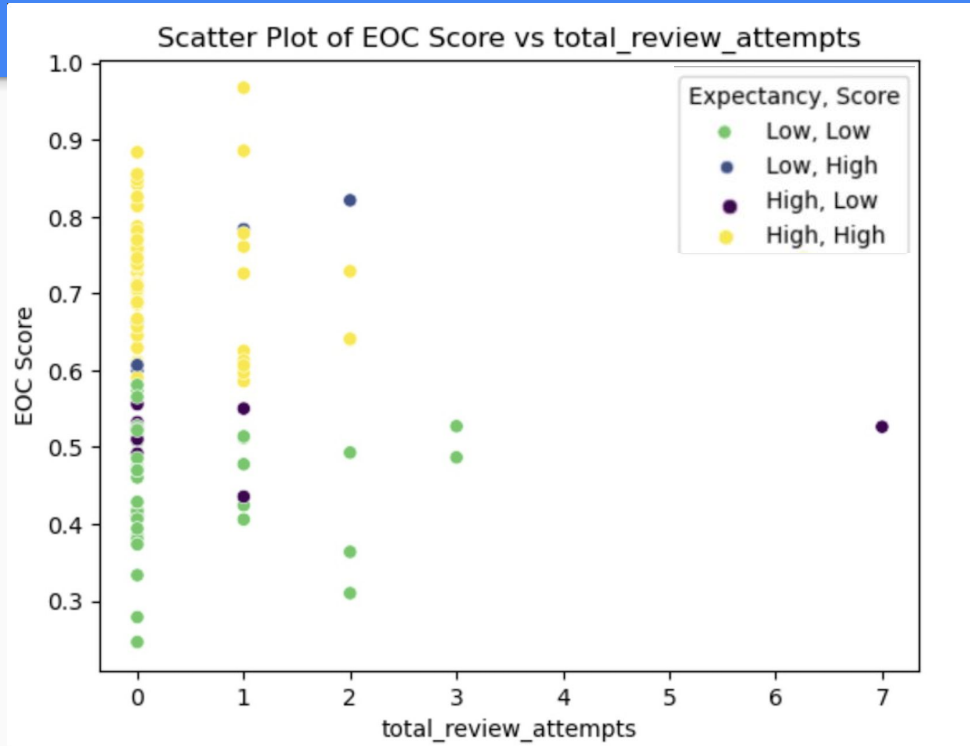
Study patterns of high expectancy low performance students

- ❖ **Their scores correlate to...**
 - ❖ The accuracy of their solutions
 - ❖ Time spent engaged on material
 - ❖ Attempts taken reviewing previous material

Study patterns of high expectancy low performance students



Study patterns of high expectancy low performance students



- ❖ Data shows that these students spend less time studying and reviewing, even compared to their low-performance counterparts that also had a low expectancy of themselves.
- ❖ This suggests wasted potential.

Potential features for helping these individuals

- ❖ **Discussion Forums:** encouraging knowledge sharing between users and providing a channel to see the insights/responses of other users
- ❖ **Reflective Prompts:** integrate more opportunities to reflect on the learning experiences
- ❖ **Case Studies & Real World Examples:** stimulations to apply knowledge to real-world scenarios
- ❖ **Peer Comparison:** allows user to compare their quiz scores/progress/time spent with other users anonymously (Alicke & Govorun, 2005)
- ❖ **Reward based on accurate self-assessment:** copes with the lack of incentive to give accurate self-assessments (Dunning, 2011)
- ❖ **Model predicting EOS scores** (example on subsequent slides)

EOC-Predict: Re-evaluation of Confidence

- ❖ **A way for AI to predict students' EOC performance based on chapter engagement**
- ❖ **Especially helpful for those that are not prepared, but still confident**
 - May induce them to study more and do some more reviewing
 - Can provide personal recommendations / areas of weaknesses / improvement

Our Model

- ❖ Predicts EOC Scores with accuracy of ~75%
- ❖ Convolutional Neural Network (Deep Learning Model)
- ❖ Pytorch Machine Learning Framework Library
- ❖ Trained on 3 different dimensions:
 - **Num_passed**: Total number of code chunks passed (per chapter per student)
 - **Engaged**: Total time spent engaged (per chapter per student)
 - **Length**: Total word count of all summary questions (per chapter per student)
- ❖ Predicts **EOC** accuracy:
 - percent questions correct for the end of chapter quiz

Our Model (Continued)

- 80/20 split: training and testing data
- Normalization: Input features are normalized to have mean 0 & variance 1
- Two fully connected layers
 - First layer (fc1) 3 input features \rightarrow 50 features (ReLU)
 - Second layer (fc2) 50 features \rightarrow 1 output value (EOC accuracy)
- Loss Function: Mean Squared Error Loss
- Adam optimizer with learning rate of 0.01
- Trained for 1000 epochs

Training Data

Feature Inputs (3)

Label

| Unnamed: 0 | student_id | chapter_number | num_passed | engaged | length | EOC |
|------------|--------------------------------------|----------------|------------|------------|--------|----------|
| 0 | 001824fb-a2fd-431d-aef6-7a1250d97a62 | 1.0 | 24 | 5663330.0 | 131 | 0.921053 |
| 1 | 001824fb-a2fd-431d-aef6-7a1250d97a62 | 2.0 | 26 | 30902762.0 | 91 | 0.839080 |
| 2 | 001824fb-a2fd-431d-aef6-7a1250d97a62 | 3.0 | 29 | 34127489.0 | 65 | 0.769231 |
| 3 | 001824fb-a2fd-431d-aef6-7a1250d97a62 | 5.0 | 12 | 15971898.0 | 71 | 0.777778 |
| 4 | 001824fb-a2fd-431d-aef6-7a1250d97a62 | 6.0 | 19 | 17611781.0 | 73 | 0.697917 |
| ... | ... | ... | ... | ... | ... | ... |
| 5995 | a3ac1e2d-7dc6-4ff1-a8df-bb6ff0f52e67 | 12.0 | 14 | 6230812.0 | 40 | 0.560000 |
| 5996 | a41d13e7-0e8b-402c-be3c-26313c7d28ce | 1.0 | 24 | 5341326.0 | 47 | 0.945946 |
| 5997 | a41d13e7-0e8b-402c-be3c-26313c7d28ce | 2.0 | 26 | 11551124.0 | 88 | 0.879518 |
| 5998 | a41d13e7-0e8b-402c-be3c-26313c7d28ce | 3.0 | 29 | 10541477.0 | 115 | 0.816327 |
| 5999 | a41d13e7-0e8b-402c-be3c-26313c7d28ce | 5.0 | 12 | 5999197.0 | 61 | 0.714286 |

6000 rows x 7 columns

Test Data

Feature Inputs (3)

| Unnamed: 0 | student_id | chapter_number | num_passed | engaged | length |
|------------|--------------------------------------|----------------|------------|------------|--------|
| 6001 | a41d13e7-0e8b-402c-be3c-26313c7d28ce | 9.0 | 13 | 5099303.0 | 53 |
| 6002 | a41d13e7-0e8b-402c-be3c-26313c7d28ce | 10.0 | 12 | 4668429.0 | 34 |
| 6003 | a41d13e7-0e8b-402c-be3c-26313c7d28ce | 11.0 | 16 | 5589897.0 | 29 |
| 6004 | a41d13e7-0e8b-402c-be3c-26313c7d28ce | 12.0 | 14 | 4784469.0 | 23 |
| 6005 | a446e23a-ace7-4cb4-b630-9f50a52248fe | 1.0 | 24 | 3550720.0 | 75 |
| ... | ... | ... | ... | ... | ... |
| 9233 | ff6ec9fe-de1d-4b45-8136-59465d9c85ab | 5.0 | 12 | 9989833.0 | 48 |
| 9234 | ff6ec9fe-de1d-4b45-8136-59465d9c85ab | 6.0 | 19 | 16650582.0 | 56 |
| 9235 | ff6ec9fe-de1d-4b45-8136-59465d9c85ab | 9.0 | 13 | 14206097.0 | 91 |
| 9236 | ff6ec9fe-de1d-4b45-8136-59465d9c85ab | 10.0 | 12 | 8793743.0 | 40 |
| 9237 | ff6ec9fe-de1d-4b45-8136-59465d9c85ab | 11.0 | 16 | 10646661.0 | 54 |

3237 rows x 6 columns

Results

```
[29]: df['perc_error'].mean()
```

```
[29]: 0.25203619419468953
```

Feature Inputs (3)

| num_passed | engaged | length |
|------------|------------|--------|
| 13 | 5099303.0 | 53 |
| 12 | 4668429.0 | 34 |
| 16 | 5589897.0 | 29 |
| 14 | 4784469.0 | 23 |
| 24 | 3550720.0 | 75 |
| ... | ... | ... |
| 12 | 9989833.0 | 48 |
| 19 | 16650582.0 | 56 |
| 13 | 14206097.0 | 91 |
| 12 | 8793743.0 | 40 |
| 16 | 10646661.0 | 54 |

Actual

| EOC |
|----------|
| 0.596154 |
| 0.721311 |
| 0.640000 |
| 0.560000 |
| 0.755556 |
| ... |
| 0.641791 |
| 0.500000 |
| 0.611650 |
| 0.469697 |
| 0.518519 |

Predicted

| Pred_EOC |
|----------|
| 0.630910 |
| 0.611050 |
| 0.528446 |
| 0.576863 |
| 0.857072 |
| ... |
| 0.644638 |
| 0.685258 |
| 0.739731 |
| 0.628444 |
| 0.647440 |

Error

| perc_error |
|------------|
| 0.058300 |
| 0.152862 |
| 0.174303 |
| 0.030113 |
| 0.134360 |
| ... |
| 0.004437 |
| 0.370516 |
| 0.209402 |
| 0.337978 |
| 0.248635 |

Conclusion

- ❖ We identify the problem of confident students over-estimating their scores, leading to them reviewing less and spending less time.
- ❖ Using a CNN machine learning model, we predict students' end of chapter scores based on their number of code chunks passed, time spent reviewing, and word density of typed responses.
- ❖ We use this information to offer students personalized advice on how they can improve their scores.