

# DASI: Defining Class Expectations to Achieve Increased Student Capabilities

# Adapting to GAI: Adapting Assignments

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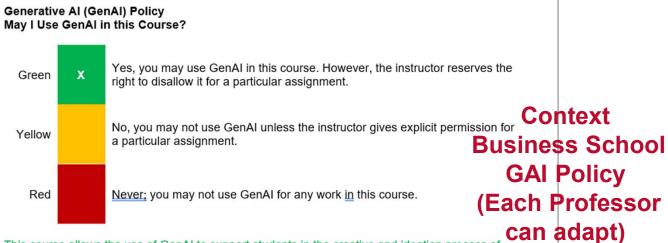
### **Motivation and Overview**

- Easy access of GAI tools
  - Less focus on syntax assume can code faster in R and Python
- Graduate Predictive Modeling with R course
  - · Use R and R Studio
  - Added more questions around presenting the results from the analysis
- Undergraduate Python Programming class
  - Used Google Colab
  - Completely embraced the use of Gemini embedded in the interface.
  - Had them present their projects to me to help me find out if they knew what their code was doing.





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This course allows the use of GenAl to support students in the creative and ideation process of assignments. However, students are expected to produce original work and not rely solely on Al to complete assignments.

#### Attribution

If students use a GenAl tool to assist in an assignment beyond the inline help in Google Colab, they must cite that work by providing the following:

- Write a reference list citation for the GenAl tool by using this <u>Grammarly Citation Generator</u>.
- Briefly describe with one or two sentences what task GenAl was used for.

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# Context – Business School GAI Policy (Each Professor can adapt)

#### **Additional Cautionary Statements**

#### GenAl can get things wrong:

GenAl tools sometimes produce outputs that are not fully accurate or are completely fabricated. Check facts and numbers for accuracy using another source. Students will be accountable for any errors produced by the tools they use.

#### Input matters:

The quality of a prompt when using GenAl impacts the quality of the output. Low-effort prompts produce low-quality results. It is important to iterate on and refine prompts.

The Dean has recently said we should just allow students to use GAI......



### **Context**

# TRY 3 BEFORE ME

### Ask Yourself

Stop and think about the directions or question. What do you know and what are you wanting to know?

# Ask the Course

Look at the syllabus, modules, class notes, and other class resources that were provided for you.

## 2 Ask Another

Ask a classmate, search the internet, ask a Generative Al tool like MS CoPilot

THEN... if you still aren't sure, ask me by clearly stating the question and your 3 attempts to answer with your reflection.



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- MSBA Students in their final Quarter
  - Adding more presenting results
  - Used R and R Studio
    - There is a GitHub I couldn't get an account ☺
      - Has Anyone got this to work?
    - https://docs.posit.co/ide/user/ide/guide/tools/copilot.html
- To use GitHub Copilot, you must have a GitHub account and an active subscription to Copilot Individuals or Copilot for Business. For more information, see <u>billing for GitHub Copilot</u> ☑. Students and faculty can use GitHub Copilot for free as part of the GitHub Education program For more information, see the GitHub Education page ☑.

# Course Topics: (Meets 2 2hr10m Classes per Week)

0	Complete DataCamp Courses: (Intro to Tidyverse/Intermediate R)
1	R/R Studio Intro/Tidyverse / Control Logic / Statistical Learning
2	Simple/Multiple Linear Regression
3	Regression Assumptions & Transformations / Logistic Regression
4	Discriminant Analysis / Cross Validation/Bootstrap
5	Ridge Regression and Lasso / Smoothers: Polynomials and Splines
6	Gen Additive Models / CART Trees
7	Bagging/Random Forest/Boosting
8	Support Vector Machines / Neural Networks
9	Deep Learning / Present Projects

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# **Advanced Predictive Modeling with R**

- Assignment Examples
  - Shared document for interpretation summaries
  - Expectation: more variety and allowed to see other examples
  - Example 1: Regression pruned/interaction
  - Example 2: Logistic and Discriminant Analysis
  - Final Project Presentation to Peers (present twice)



 Example 1: Regression pruned/ interaction Reg\_adv\_full: sales = 2.939 + 0.046(TV) + 0.189(radio) + -0.001(newspaper)

Reg\_adv\_pruned: sales = 2.921 + 0.046(TV) + 0.188(radio)

Reg\_adv\_pruned r^2: 0.897 Reg\_adv\_pruned MAE: 1.254

Reg\_adv\_interact: sales = 6.75 + 0.019(TV) + 0.029(radio) + 0.001(TV\*radio)

Reg\_adv\_interact r^2: 0.968 Reg\_adv\_interact MAE: 0.660

(Describe process)



The three models above were evaluated on their accuracy in predicting sales for every dollar spent in each mode of media. The newspaper predictor was removed due to being an insignificant predictor, resulting in the creation of the pruned model. A third model was created that multiplied the TV and radio predictors to test how forcing them to interact would affect the model's accuracy. The best performing model proved to be the interact model as it has the best r^2 and the lowest MAE.

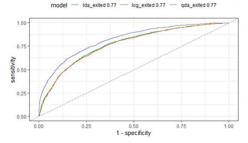
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# **Advanced Predictive Modeling with R**

- Example 2: Logistic and Discriminant Analysis
- Include in your discussion the 5 metrics and the ROC Curve as well as the changed cutoff models. Make a case for prioritizing sensitivity or precision.



 Example 2: Logistic and Discriminant Analysis



Of the three first models, with cutoff 0.5, the best performance is the QDA. Not <u>just</u> the AUC significantly improved but the accuracy, sensitivity, specificity and precision increased too.

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model	accuracy	sensitivity	specificity	precision auc
log_churn	0.811	0.215	0.964	0.604 0.77
lda_churn_full	0.808	0.233	0.955	0.568 0.77
qda_churn_full	0.836	0.387	0.951	0.668 0.82
log_churn_corr	0.700	0.700	0.700	0.374 0.77
lda_churn_corr	0.674	0.759	0.652	0.358 0.77
qda_churn_corr	0.763	0.697	0.781	0.448 0.82

Using the best cutoff for each of the models, QDA has the highest accuracy, precision, and specificity. However, the model that better predicts the 1 (higher sensitivity) is LDA. In this case, when false negatives are costly, LDA will be the best model since it is typically more important to identify as many potential churners as possible to mitigate churn.

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# So did anyone think, What is precision again?

describe precision versus sensitivity in classification

But I digress.....

Sure! Let's break down precision and sensitivity (also known as recall) in the context of classification

#### Precision

- Definition: Precision is the proportion of true positive results among all the positive results predicted by
  the model.
- Formula: \$\$\text{Precision} = \frac{\text{True Positives (TP)}}{\text{True Positives (TP)}} + \text{False Positives}
- Focus: Precision focuses on the accuracy of the positive predictions. It answers the question: Of all the instances the model predicted as positive, how many were actually positive?
- Use Case: Precision is crucial when the cost of false positives is high. For example, in spam detection, you
  want to ensure that emails marked as spam are indeed spam to avoid missing important emails.

#### Sensitivity (Recall)

- Definition: Sensitivity, or recall, is the proportion of true positive results among all the actual positive
- Formula: \$\$\text{Recall} = \frac{\text{True Positives (TP)}}{\text{True Positives (TP)}} + \text{False Negatives (FN)}}\$\$\$
- Focus: Recall focuses on the model's ability to identify all relevant instances. It answers the question: Of all
  the actual positive instances, how many did the model correctly identify?
- Use Case: Recall is important when the cost of false negatives is high. For example, in medical diagnostics, you want to ensure that all patients with a disease are correctly identified to provide necessary treatment.

#### Trade-of

- Inverse Relationship: Often, there is a trade-off between precision and recall. Improving one can lead to a
  decrease in the other. For instance, increasing the threshold for classifying an instance as positive can
  improve precision but may reduce recall, and vice versa.
- Balancing Act: The balance between precision and recall depends on the specific application and the costs associated with false positives and false negatives.

#### Example

Imagine a model designed to detect spam emails:

- High Precision: The model correctly identifies most spam emails but might miss some (high precision, lower recall).
- High Recall: The model catches almost all spam emails but also marks some legitimate emails as spam (high recall, lower precision).

UNIV\_Would you like to dive deeper into any specific aspect of these metrics?



- Final Project Present to peers in small groups
  - Present twice
  - · Second time typically faster, focuses in on results quicker
  - Should they present to me 3<sup>rd</sup> time recording to Zoom?



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# **Python Programming (and Excel VBA)**

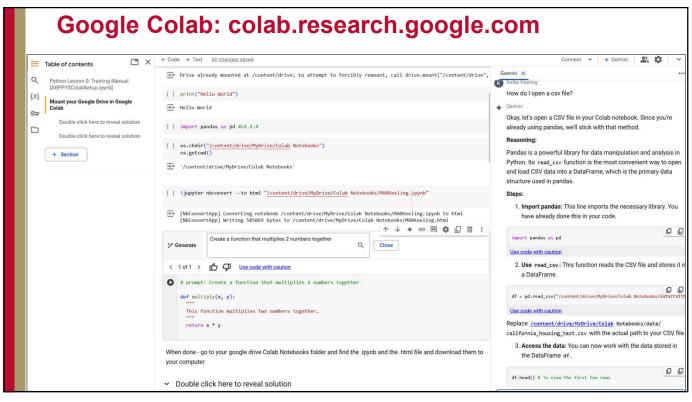
- BSBA Students (Soph, Junior, Senior)
  - First course in major/minor
  - Know some will use GAI Level Playing Field
    - Google Colab
    - <a href="https://colab.research.google.com">https://colab.research.google.com</a>



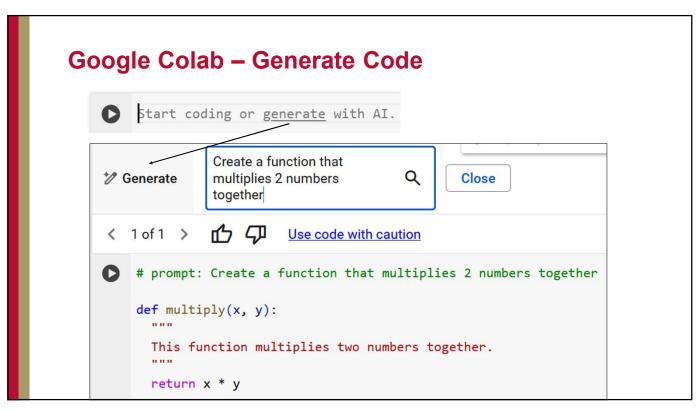
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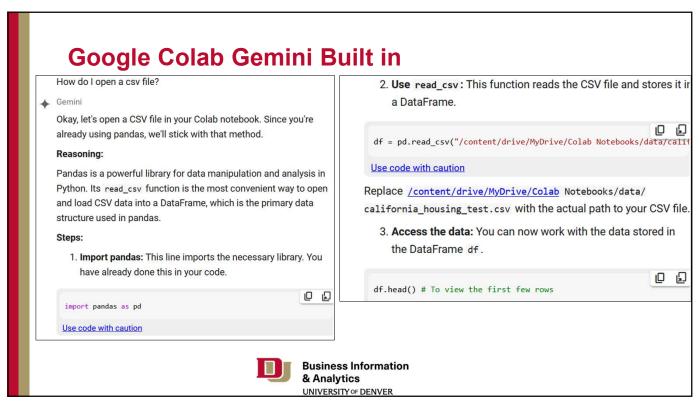
1	PY1: Intro to Python/Google Colab, Python Basics, User Input and Output
2	PY2: Basic Control Logic Using Python/Pseudocoding
3	PY3: Data Structures and Pandas Reading in Data
4	PY4:Functions, Cleaning, and Creating Summaries
5	PY5:Creating Summaries and Merging/Grouping Data
6	PY6:DataFrame Audit and Creating Graphs
7	PY7.Basic Statistics (t tests, ANOVA, Regression)
8	VBA1:Good Spreadsheets Modeling/Excel Basics & Built in Functions
9	VBA2: Structured Programming/Control Logic/Using Buttons/Message and Input Boxes/Worksheet Ranges/Charts
10	VBA3: Coding Wrap up and Excel/Python Communication
Finals	Project create Python program with menu to analyze data/Excel Workbook to create descriptive summaries and bar chart - Presentations to Me

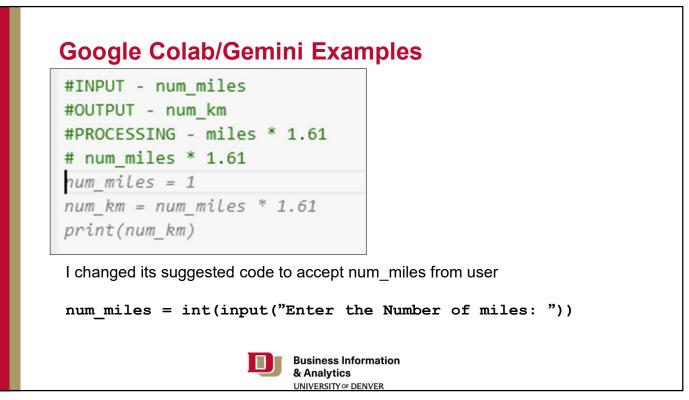
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### Google Colab/Gemini Examples

Next example, I put in INPUT and OUTPUT it scarily created the rest learning from my previous example added the #PROCESSING changed to using an input and converted it to float!!!

```
#INPUT - dog_yrs

#OUTPUT - hum_yrs

#PROCESSING - dog_yrs * 7

# dog_yrs * 7

dog_yrs = float(input("Enter the number of dog years: "))

print(dog yrs * 7)
```



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# **Python Programming Class**

Sample Problem:

#### 1.Create Pseudocode WITHOUT Generative Al

Click Settings Gear on Top Right

Choose Al Assistance, Check to Hide Generative Al Features

Create Pseudocode to solve this Prompt with INPUTS, OUTPUTS, PROCESSING, CHECKS (create 2 checks to try out your code by finding the correct answer you should get to 2 possible entered numbers)

PROMPT: **Calculate a running sum.** Ask a user to enter a number. Keep asking for numbers until a negative number is encountered. When that happens, print the running sum of previous numbers.

#### 2.Create Solution WITHOUT Generative Al

Using the Pseudocode you created and class notes/Python syntax file, solve the question in the prompt.

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# **Python Programming Class**

#### Sample Problem:

#### 3.Create Solution WITH Generative Al

Click Settings Gear on Top Right

Choose Al Assistence, UnCheck Hide Generative Al Features

Enter the Prompt into Gemini and copy it's solution below. If you provide any additional prompts to refine the answer, include those and the revised answer as well.

#### 4. Revise your solution from Part 2.

Copy your answer to the problem without GAI from Part 2 below.

Revise your answer based on what you learned from the Gemini Answer.

Reflect on how you changed what you did originally



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### **Comparing Some vs None - Help**

### **Help No Programming Exp**

**Error Explanation** 

**Advanced Code Output** 

Overall Helpfulness

Learning Aid

Effective Use

**Tool Utilized** 

### **Help Some Programming Exp**

Time Efficiency

Language Flexibility

Focus on Understanding

Limited Use in Exams

Efficiency and Guidance



### **Comparing Some vs None - Hinder**

### **Hinder No Programming Exp**

Understanding Issues
Incorrect Outputs
Over-Reliance
Learning Curve

Course-Specific Impact

### **Hinder Some Programming Exp**

Over-Reliance

Premature Assistance

Learning Impact

Syntax Memorization

**Future Concerns** 



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### **Comparing Some vs None – Future Use**

### **Future No Programming Exp**

**Professional Applications** 

Perspective and Problem-Solving

Efficiency in Basic Tasks

Limitations and Ethical

Considerations

**Programming Assistance** 

Selective Use

### **Future Programming Exp**

**Documentation and Simple Tasks** 

Workforce Experience

**Decision-Making** 

**Medical Applications** 

**Educational Benefits** 



# **Take Aways**

- Encourage to turn off GAI at times (write pseudocode first, demonstrate turning off more)
- 20 students presented projects to me this week for 12-15m (scheduled 10m). Estimated ¼ had issues explaining specific parts of code.
  - (I did pick on some things we didn't cover in class (VBA: Excel Dictionaries) or interesting features (Python: Offering list of columns for X variable that didn't include Y column chosen)
  - Wondering can this can be scaled? Or done more often?
- How else can I have them practice programming logic?



