Overview

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Key Concepts of the Newsvendor Problem	10	
Contextual Features, Linear Regression, and Time Series Models	10	
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Total:	60	

- (a) (2 Points) What is the primary goal of the Newsvendor Problem" discussed in the session?
 - A. To minimize the total number of orders
 - B. To balance underage and overage costs
 - C. To predict future market trends
 - D. To increase customer satisfaction with faster delivery times
- (b) (2 Points) In the ZenBite case example, what is the purpose of splitting the data into a training and test dataset?
 - A. To ensure that both datasets have the same demand values
 - B. To evaluate the model's performance on unseen data
 - C. To ensure a random distribution of data
 - D. To separate high-demand and low-demand products
- (c) (2 Points) What is the main advantage of using the average demand for planning in the ZenBite case study?
 - A. It provides a simple, easy-to-implement baseline strategy
 - B. It guarantees zero underage and overage costs
 - C. It ensures the lowest possible cost for every product
 - D. It adjusts dynamically to fluctuations in demand
- (d) (2 Points) What is the main idea behind the Sample Average Approximation (SAA) method?
 - A. To calculate the optimal order quantity based on a normal distribution of demand
 - B. To maximize the fulfillment of daily demand using the highest observed demand in the data
 - C. To defrost the same quantity each day based on the median demand of past data
 - D. To minimize average costs by estimating the optimal order quantity based on historical demand data
- (e) (2 Points) Briefly explain how to calculate the optimal service level α and what it represents?

Solution: α is the ratio of underage and overage costs, it represents how often we want to fulfill the total demand

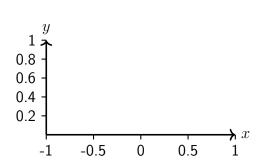
$$\alpha = \frac{cu}{cu + co}$$

- Question 2 Contextual Features, Linear Regression, and Time Series Models 10 Points
 - (a) (2 Points) What is a contextual feature in the Newsvendor Problem?
 - A. Additional information available at the time of decision-making that may influence demand
 - B. A feature used to directly calculate the demand based on past values
 - C. A specific type of data that includes only numeric variables like sales figures or temperature
 - D. Historical data that is only used to validate the accuracy of the model
 - (b) (2 Points) How does the linear regression Newsvendor model differ from ordinary linear regression?
 - A. Linear regression Newsvendor minimizes the sum of squared errors between actual and predicted demand
 - B. Linear regression Newsvendor uses a loss function that penalizes underordering different from overordering
 - C. Linear regression Newsvendor only applies to time series data
 - D. There is no significant difference between the two
 - (c) (2 Points) What is overfitting in the context of machine learning?
 - A. Fitting the model to only the first half of the dataset
 - B. When a model performs well on training data but poorly on test data due to being too complex
 - C. Underestimating future demand consistently
 - D. Using too few contextual features in the model
 - (d) (2 Points) Briefly explain what autocorrelation in a time series means?

Solution: Autocorrelation in a time series measures how current values are related to past values at different time lags.

- (e) (2 Points) What is an example for incorporating seasonality into a time series model?
 - A. By adding a lag feature for the last day
 - B. By including the month of the year as a feature
 - C. By ignoring trend components in the data
 - D. By ensuring the model only considers weekdays

- Question 3 Deep Learning and Neural Networks in the Newsvendor Problem 10 Points
 - (a) (2 Points) What is the primary purpose of using deep learning in the Newsvendor Problem?
 - A. To replace all previous models with a more complex algorithm
 - B. To reduce the number of features needed for accurate demand prediction
 - C. To account for non-linear relationships between features and demand, and capture interactions between features
 - D. To eliminate the need for training data
 - (b) (2 Points) Why do we train deep learning models in an iterative process over multiple epochs?
 - A. To gradually adjust the model's parameters and improve its predictions with each pass through the data
 - B. To ensure the model reaches a global minimum of the loss function on the first pass
 - C. To avoid using too much data at once, which could cause the model to overfit
 - D. To ensure that the model changes weights randomly with each iteration
 - (c) (2 Points) Why is the learning rate important when training deep learning models?
 - A. It controls how much the model's parameters are adjusted at each step during training, impacting how quickly or slowly the model learns
 - B. It determines how many layers the model will have
 - C. It ensures that the training data is normalized before being used in the model
 - D. It prevents the model from overfitting by limiting the number of parameters learned
 - (d) (2 Points) What is the role of hyperparameters in training a deep learning model?
 - A. Hyperparameters control the structure and training process of the neural network, such as the number of layers, learning rate, and batch size
 - B. Hyperparameters are automatically optimized by the model and do not require manual tuning
 - C. Hyperparameters determine the weights of the connections between neurons
 - D. Hyperparameters are used to adjust the features being input into the model
 - (e) (2 Points) Draw the ReLU activation function in the coordinate system below.



- (a) (2 Points) What is the role of a Decision Tree when solving the Newsvendor Problem?
 - A. To provide a single numerical prediction for demand based on all features combined
 - B. To improve the accuracy of linear models by reducing error rates
 - C. To avoid the need for optimization steps by providing pre-calculated solutions
 - D. To split data into smaller subsets based on the most informative features
- (b) (2 Points) In the context of Decision Trees, what are leaf nodes?
 - A. The final nodes in the tree that contain data points grouped by similar characteristics, used to make predictions
 - B. The initial nodes that contain all the data before any splits are made
 - C. Intermediate nodes that contain predictions for each feature
 - D. Nodes that perform optimization on the predicted data
- (c) (2 Points) Why are Random Forests more robust than individual Decision Trees?
 - A. Random Forests use deep learning techniques to enhance model performance
 - B. Random Forests are trained on more features than individual decision trees
 - C. Random Forests combine multiple trees, to reduce variance and improve accuracy of the predictions
 - D. Random forests eliminate the need for feature engineering
- (d) (2 Points) How does the use of multiple decision trees in a random forest help avoid overfitting?
 - A. Each tree is trained on a random subset of the data, reducing the reliance on specific patterns and improving generalization
 - B. Random forests apply pruning techniques to limit the depth of each tree
 - C. Random forests automatically drop irrelevant features during training
 - D. The use of a single strong learner in random forests helps minimize overfitting
- (e) (2 Points) What does it imply if your trained Decicion Tree has exactly one observation in each leaf node?

Solution: It implies that the model overfitted.

```
dlnv = DeepLearningNewsvendor(
                    cu=15,
                    co=5,
                    neurons=[25, 12],
                    activations=['ReLU', 'ReLU'],
                    optimizer='adam',
                    epochs=250,
                    batch_size=64,
                    learning_rate=0.003,
                    drop_prob=0.15,
10
                    random_state=None,
11
                    verbose=1,
12
                    print_freq=10
13
   )
14
15
   dlnv.fit(
16
           X_train,
17
           demand.to_numpy()[train],
18
           validation_data = (X_val, demand.to_numpy()[val]))
19
20
   q_dlnv = dlnv.predict(X_val)
21
22
   avg_cost_dlnv = average_costs(demand.to_numpy()[val], q_dlnv, cu=cu, co=co)
23
24
   print(f"The final validation performance is: {avg_cost_dlnv:.2f}\)
```

(a) (2 Points) What do you have to change to train the model to have a higher service level than 75%?

Solution: Increase cu or decrease co

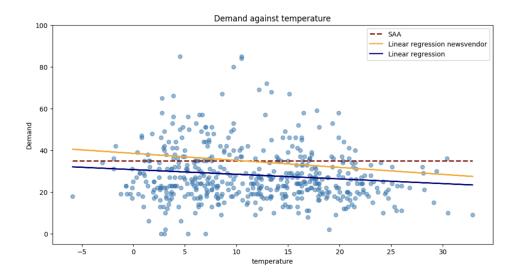
(b) (2 Points) What do you have to change in the code to use a three-layer network with 16 neurons in each layer.

Solution: Change the code in row 4 to neurons=[16,16,16] and change the code in row 5 to activations=['ReLU', 'ReLU', 'ReLU']

(c) (2 Points) You suspect that your model overfits. Name two changes in the code above that would help to react to this problem.

Solution: Increase drop_prob or decrease the number of neurons

```
from ddop2.newsvendor import LinearRegressionNewsvendor
   from sklearn.linear_model import LinearRegression
   from ddop2.newsvendor import SampleAverageApproximationNewsvendor
   # Define the cost structure
   cu = 15
   co = 5
   # Create the models
   lr = LinearRegression()
   lrnw = LinearRegressionNewsvendor(cu, co)
11
   SAA=SampleAverageApproximationNewsvendor(cu, co)
12
13
   # Fit the models
14
   lr.fit(features.loc[train][['temperature']], demand[train])
   lrnw.fit(features.loc[train][['temperature']], demand[train])
   SAA.fit(demand[train])
```



Please answer the following questions:

(a) (2 Points) How does the red line (SAA) change if we increase cu?

Solution: Moves up because it is more expensive to understock

(b) (2 Points) How does the blue line (Linear Regression) change if we increase co?

Solution: Unchanged because Linear Regression minimizes MSE

(c) (2 Points) How does the yellow line (Linear Regression Newsvendor) change if we decrease cu?

Solution: Moves down because the costs of overstocking decrease

(a) (6 Points) In the table below you find the results of a prediction using a Random Forest for the next day. Use the output to calculate the optimal order quantity given underage costs (cu) of 15 and overage costs (co) of 5.

Sample	Weekend	Temperature [in °C]	Demand	# Occurrences in leaf nodes across trees	Weight
1	0	19	27	2	2/10
5	0	24	20	5	5/10
6	0	22	23	3	3/10
Total:				10	1

Tabelle 1: Sample Table with Occurrences and Weight

Solution: Step 1 - Calculate service level:

$$\alpha = \frac{15}{15+5} = 75\% \tag{1}$$

Step 2 - Find the 0.75 quantile:

$$d^{ordered} = [20, 23, 27] \tag{2}$$

$$(1 - p^{weighted}) = [0.5, 0.8, 1] \tag{3}$$

We get an optimal order quantity of 23

(b) (2 Points) We used the following code to train the model:

```
from ddop2.newsvendor import RandomForestWeightedNewsvendor

# Set-up the model including the cost structure
frow = RandomForestWeightedNewsvendor(cu=cu,co=co,
criterion="squared_error",
n_estimators=100,
min_samples_split=5,
min_samples_leaf=3,
min_weight_fraction_leaf=0.,
max_features=0.9,
max_leaf_nodes=None,
)
```

What happens if we set min_samples_leaf to 5?

- A. The new table will have exactly 5 rows
- B. The new table will have at most 5 rows
- C. The new table will have a minimum of 5 rows
- D. The new table will have the same number of rows