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# **Prediction of Fatty Acid Changes in Corn under Different Storage Conditions Based on SAMformer**

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### **INTRODUCTION**

- As one of the major staple crops in the world, the demand for corn has been climbing year by year due to global population growth and rising living standards. In 2023, global demand for corn reached 140 million tons, a 5% increase compared to the previous year. In different environments, the change in the fatty acid value of corn is very noticeable, and corn stored at high temperatures is more prone to acidification, resulting in an increase in fatty acid value. This phenomenon leads to a deterioration in corn quality and can cause severe losses for grain storage and food processing enterprises.
- Based on this, the current study utilizes the SAM former method, taking into account various environmental conditions to establish a model

# **ANALYSIS OF RESULTS**

#### Evaluation metrics for predictive models

When evaluating the performance and fit of prediction models, commonly used metrics include: MAE (Mean Absolute Error), MSE (Mean Squared Error), RMSE (Root Mean Squared Error), MAPE (Mean Absolute Percentage Error), and SMAPE (Symmetric Mean Absolute Percentage Error). Both MAE and MSE measure the error between predicted values and actual values, but MSE applies a heavier penalty for larger errors because it is the square of the errors. RMSE is the square root of MSE, providing a standard deviation of the errors. Compared to MSE, RMSE is more intuitive in scale and easier to interpret. MAPE is based on percentage errors, thus they help in assessing the size of prediction errors relative to actual values. Similar to

predicting fatty acid changes during corn storage, including three key factors: time, initial moisture, and temperature. The experimental results indicate that this model has the best predictive performance compared to other models.

### **METHODS**



Figure 1. SAMformer model architecture.

• In this study, we use the Python programming language to train the SA Mformer model to predict fatty acid values in corn. During the training process, SAMformer learns the trends in fatty acid data in order to pre dict fatty acid values at future time points. To determine the optimal pa rameter configuration, we conducted multiple experiments by adjustin g parameters and observed the changes in losses throughout the proces

MAPE, SMAPE symmetrically handles the differences between predicted and actual values, and its calculation method can reduce error magnification caused by very small or large values in the dataset. The formula is as follows:

$$MAPE = \frac{100\%}{n} \left| \sum_{i=1}^{n} \left| \frac{\mathbf{p}_i - y_i}{y_i} \right| \right|$$
$$MSE = \frac{1}{n} \sum_{i=1}^{n} (\mathbf{p}_i - y_i)^2$$

$$RMSE = \sqrt{\frac{1}{n} \sum_{i=1}^{n} (\mathfrak{p}_i - y_i)^2}$$

$$MAE = \frac{1}{n} \sum_{i=1}^{n} |\mathbf{p}_i - \mathbf{y}_i|$$

$$SMAPE = \frac{100\%}{n} \sum_{i=1}^{n} \frac{|\mathbf{p}_i - y_i|}{(|\mathbf{p}_i| - |y_i|)/2}$$

#### **Result**

• To effectively evaluate the performance of the SAMformer in predicting changes in fatty acids during the storage of corn, this paper selected LSTM and Transformer for comparative experiments, with each model having the same proportion of training, testing, and validation sets. To objectively assess and describe the performance of these models, we used five evaluation metrics to calculate the error values for each model: MSE, RMSE, MAP, MSPE, and MAE. Compared to other models, the SAMformer model exhibited the smallest prediction error, as shown in Table 5.

s until the best parameters were determined, as shown in Table 4.

Table 4. Parameter setting for the SAMformer model.

| Parameters    | <b>Parameter Setting</b> |  |  |
|---------------|--------------------------|--|--|
| Dropout       | 0.1                      |  |  |
| loss          | MSE                      |  |  |
| Optimizer     | Adam                     |  |  |
| Epochs        | 1000                     |  |  |
| Batch_size    | 256                      |  |  |
| Learning rate | 0.001                    |  |  |

Table 5. Experimental results on the comparative performance of SAMformer-based corn quality prediction models.

| Model       | MAE  | MSE   | RMSE | MAPE | MSPE |
|-------------|------|-------|------|------|------|
| LSTM        | 0.26 | 0.08  | 0.28 | 0.55 | 2.29 |
| Transformer | 0.16 | 0.04  | 0.20 | 0.35 | 0.21 |
| SAMformer   | 0.05 | 0.004 | 0.06 | 0.18 | 0.07 |

# CONCLUSIONS

This experiment established a corn fatty acid prediction model based on SAMformer, using time, moisture content, and temperature as independent variables, and fatty acids as the dependent variable. It predicts the fatty acid values of corn with low moisture, safe storage moisture, and high moisture under different storage conditions. This model achieves the smallest error in predicting the quality changes during corn storage. The results show that our model can reasonably predict the changes in corn quality based on the trends of quality indicators under different environmental conditions. The methods used in this paper can help regulatory authorities take timely measures to delay the deterioration of stored corn, as well as reduce environmental pollution and save resources.



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References

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