

Abstract geometric lines in the top left corner, consisting of several thin, light brown lines that intersect and form various polygons and shapes, creating a modern, minimalist design element.

PREDICT AND MAP WHEAT PRODUCTION ACROSS THE SK, BASED  
ON A 1938–2021 DATABASE USING AN UNSUPERVISED MACHINE  
LEARNING CLUSTERING APPROACH.

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# INTRODUCTION

The crops, lands, and timing of 622 rural municipalities in Saskatchewan were used in a historical data analysis. Python and Geopanda used unsupervised machine learning clustering to analyze this large cohort.

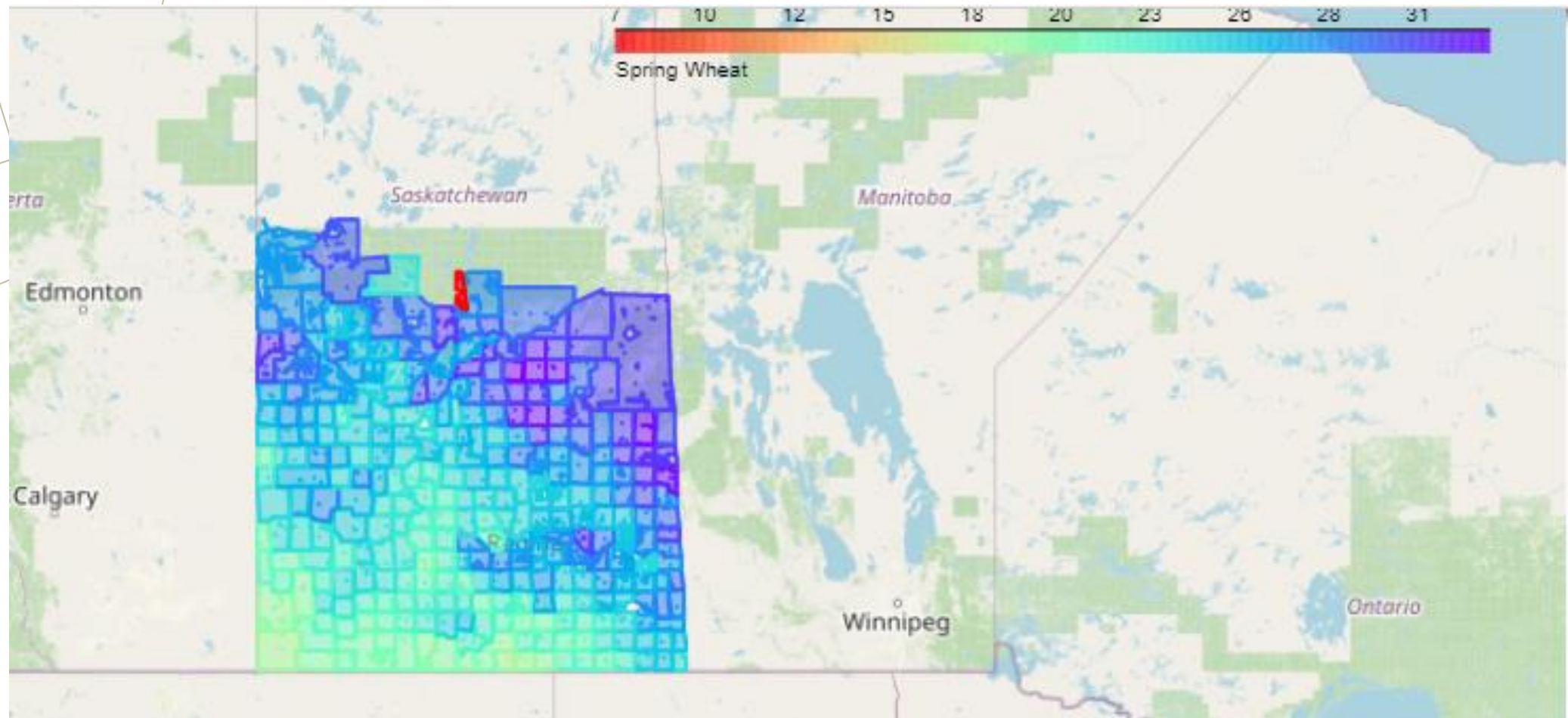
## DATA BASES

rm\_crop\_yields\_1938\_2021.csv

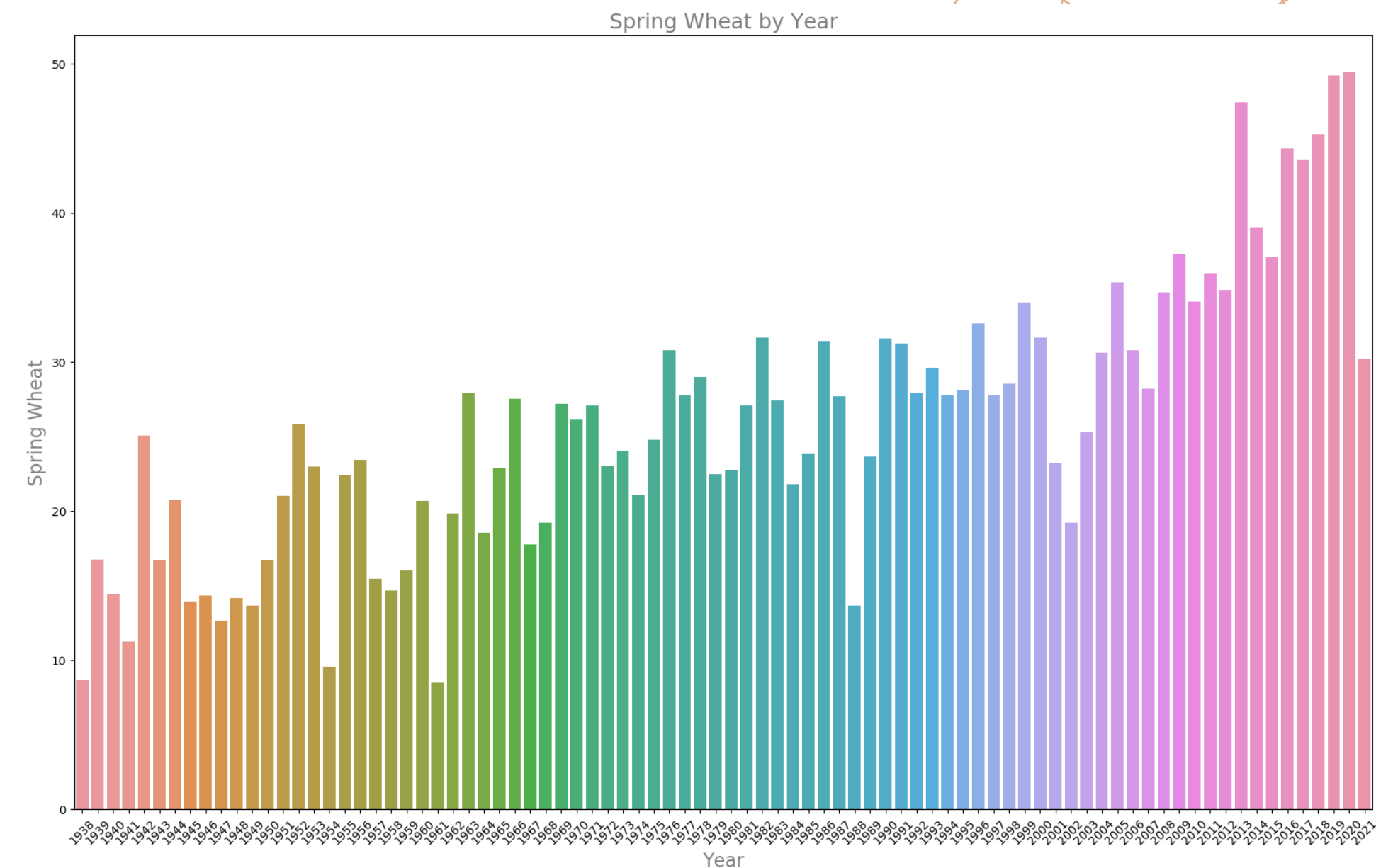
Rural Mnicipilaty.shp

# DATA ANALYSIS

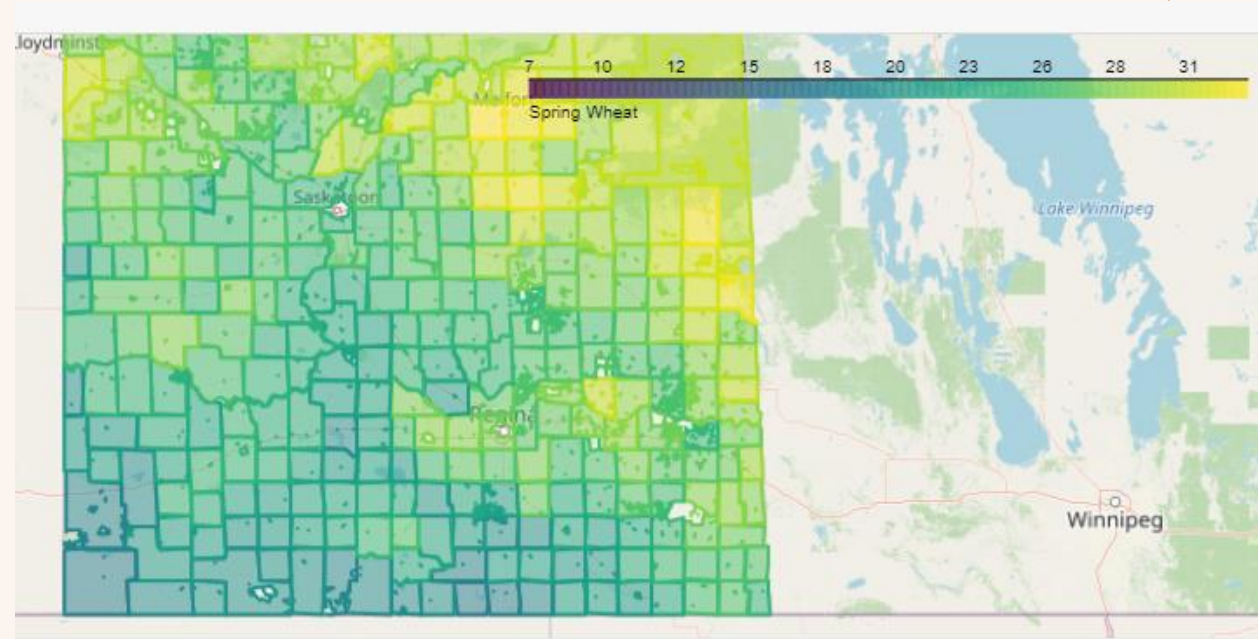
Why I chose Spring wheat data?



# WHEAT PRODUCTION ACROSS THE SK, BASED ON A 1938–2021 DATABASE

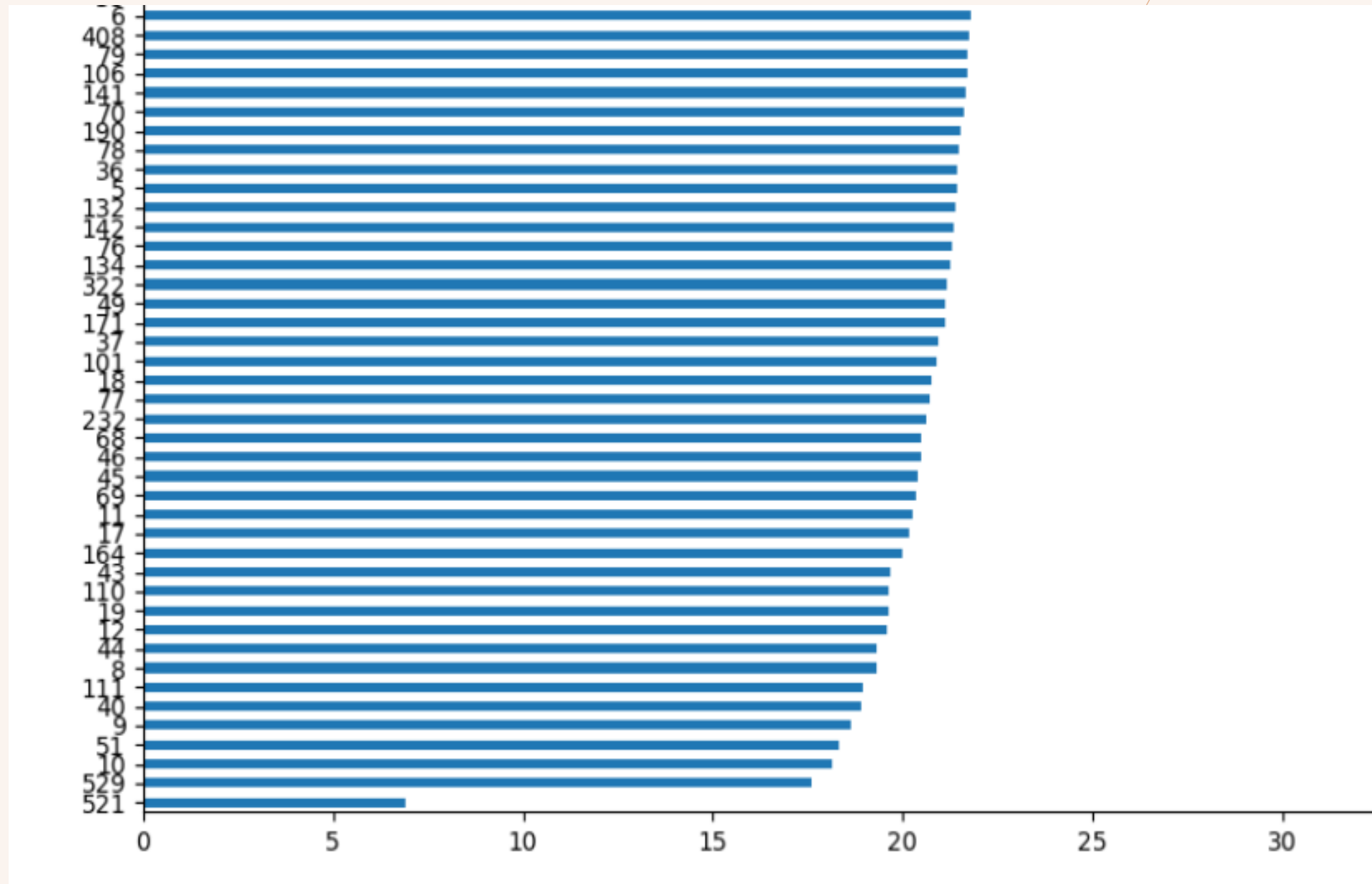


# METHODOLOGY



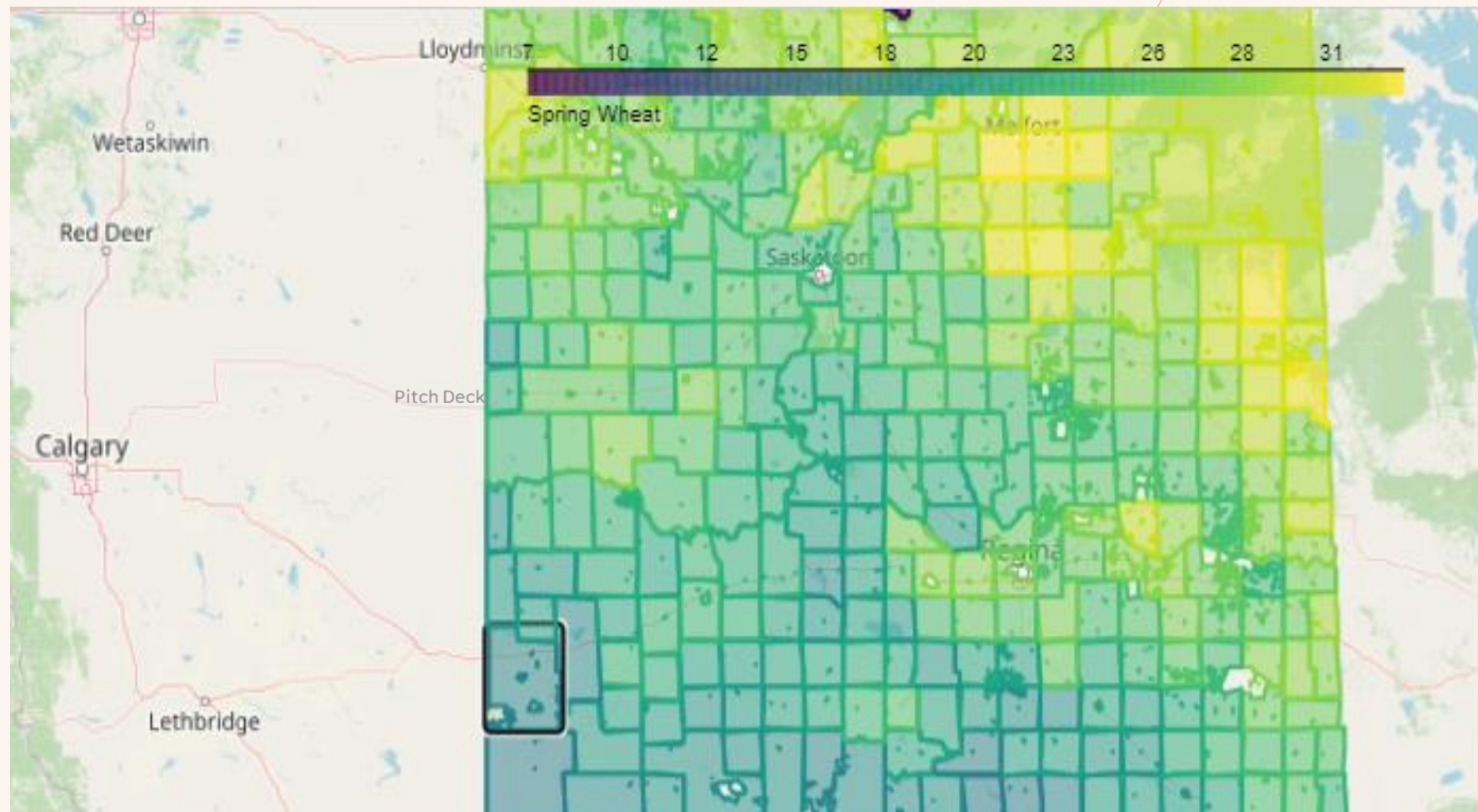
Generating Summary Statistics: Use the `describe()` method to generate summary statistics such as mean, standard deviation, quartiles, etc. Data were analyzed using the K-Means Clustering method as a technique for performing data groupings. Furthermore, the data classification procedure was based on the degree of each component's membership. This analysis was performed by using python (pandas, numpy, matplotlib, seaborn) and Geopandas.

# FINDING A RM FOR MODELING





# TIME SERIES FORECASTING RM 111 MAPLE CREEK





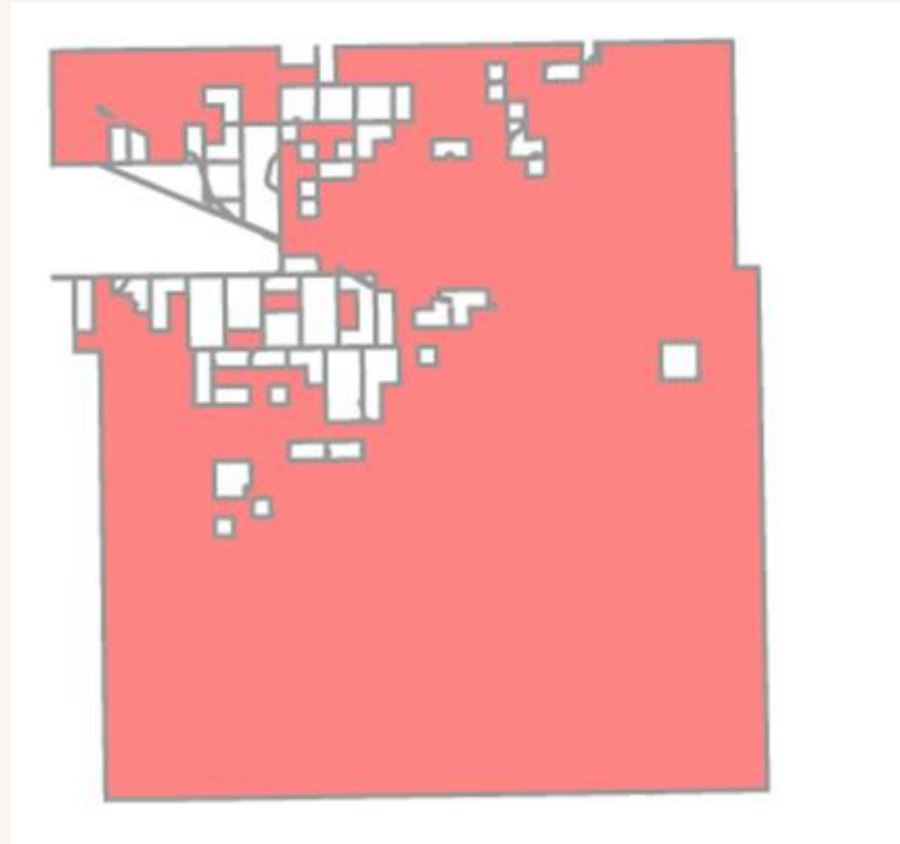
# DATA MODELING FOR RM 111, MAPLE CREEK



Unsupervised machine learning algorithms are also useful for understanding the relationships between different variables in a dataset. For example, you can use clustering algorithms such as K-Means to identify clusters of similar points in a dataset

Define the features to be used for clustering

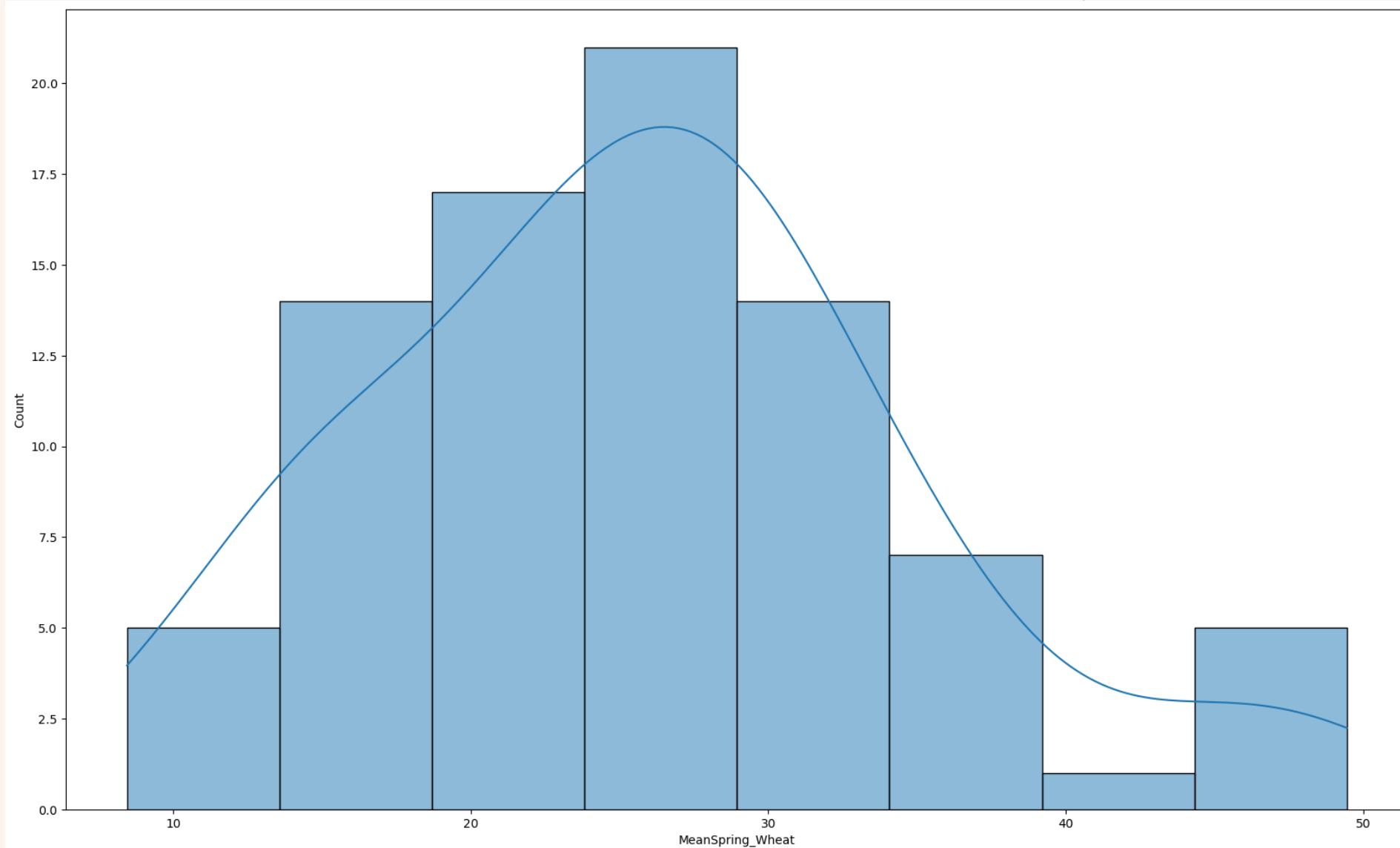
# CREATE A MODEL FOR SPRING-WEAT IN RM 111 MAPLE CREEK



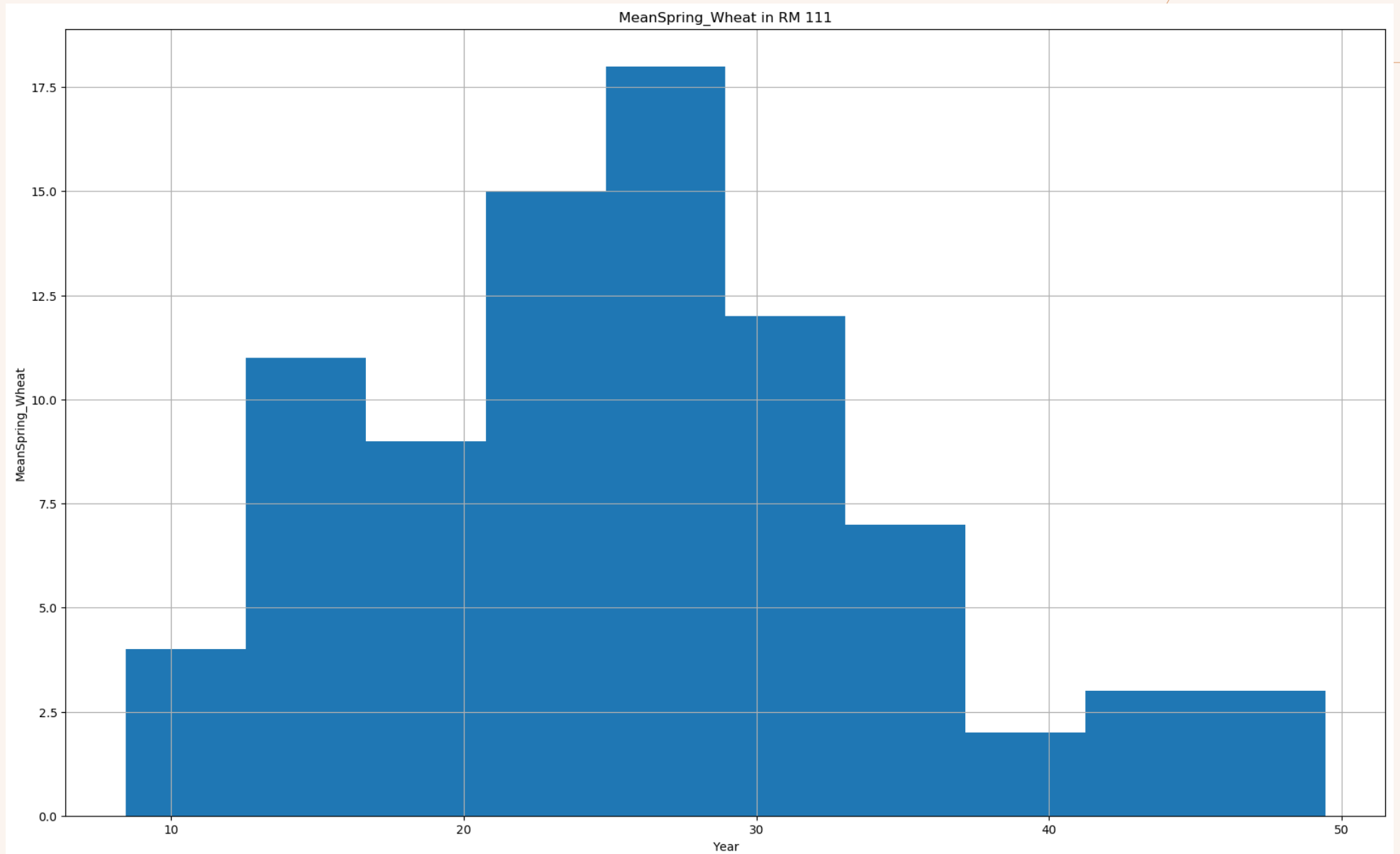
# DATA MODELING FOR SK 622 RMS -HISTOGRAM

Shape (84, 6)

slope: 0.3097599939120827  
intercept: -587.316443382177  
r: 0.8226524204739321  
p: 8.255635700786158e-22  
stderr: 0.023641040235293034

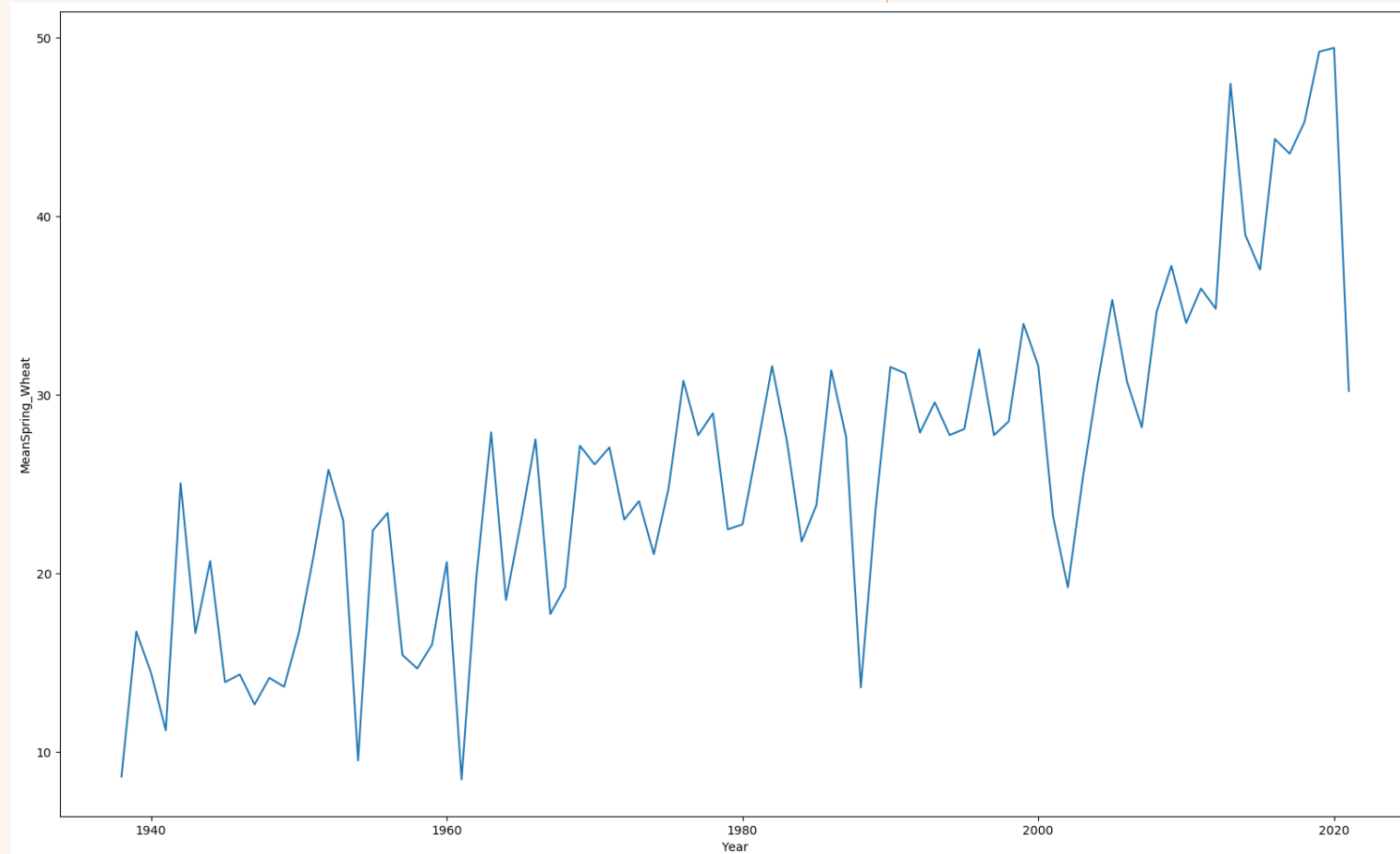


# DATA MODELING FOR SK 622 RMS – BY YEAR



# TIME SERIES FORECASTING RM 111 MAPLE CREEK

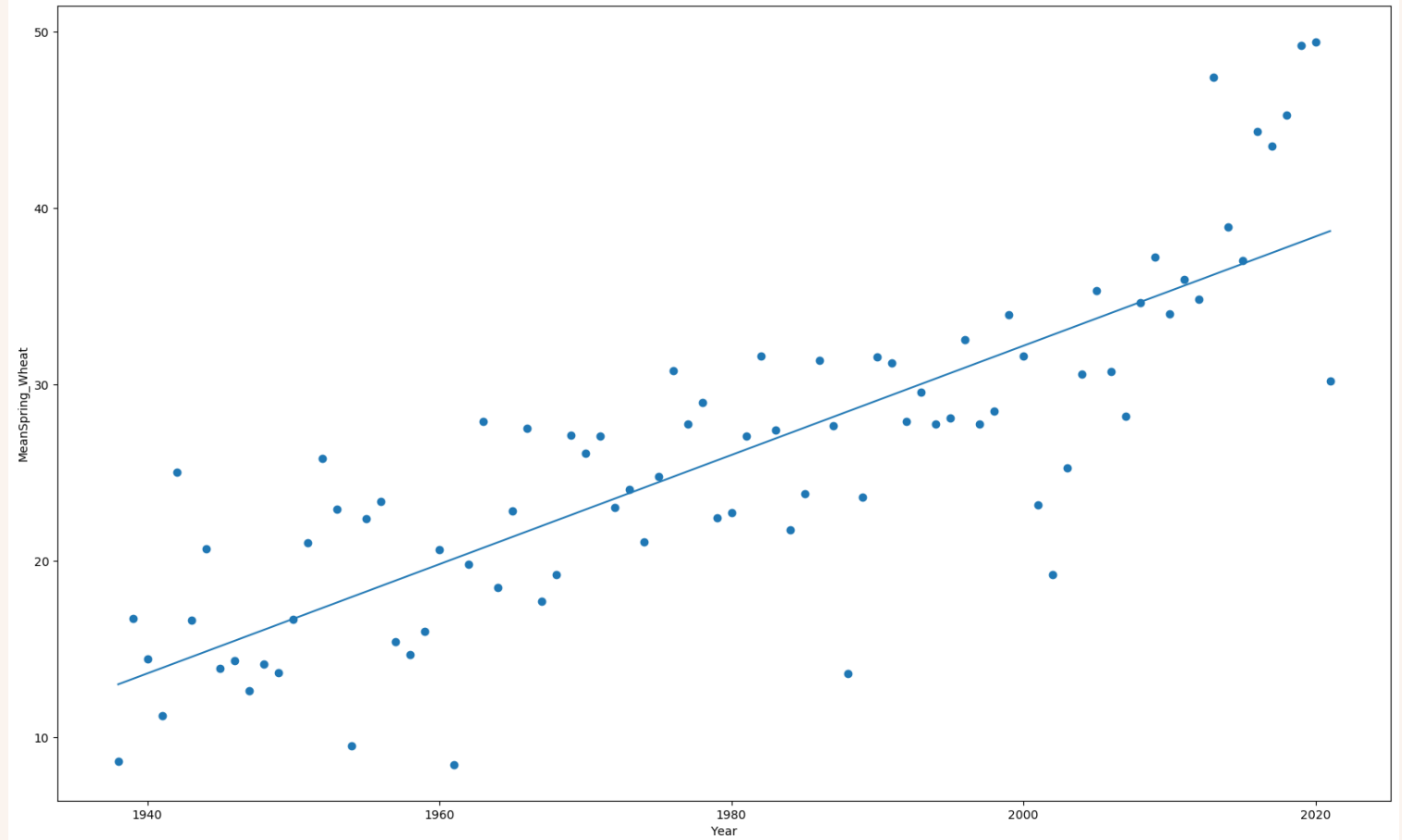
Autoregressive Integrated Moving Average (ARIMA) models facilitate demand forecasting, such as in determining future demand for spring\_weats. That is because the model provides products related to time. This forecasting can also be used to predict future yield production.



# LINEAR REGRESSION FORECASTING FOR SPRING\_WEAT IN RM 111, MAPLE CREEK

We can see that there is a positive linear relationship between amount of mean spring\_weat production and time.

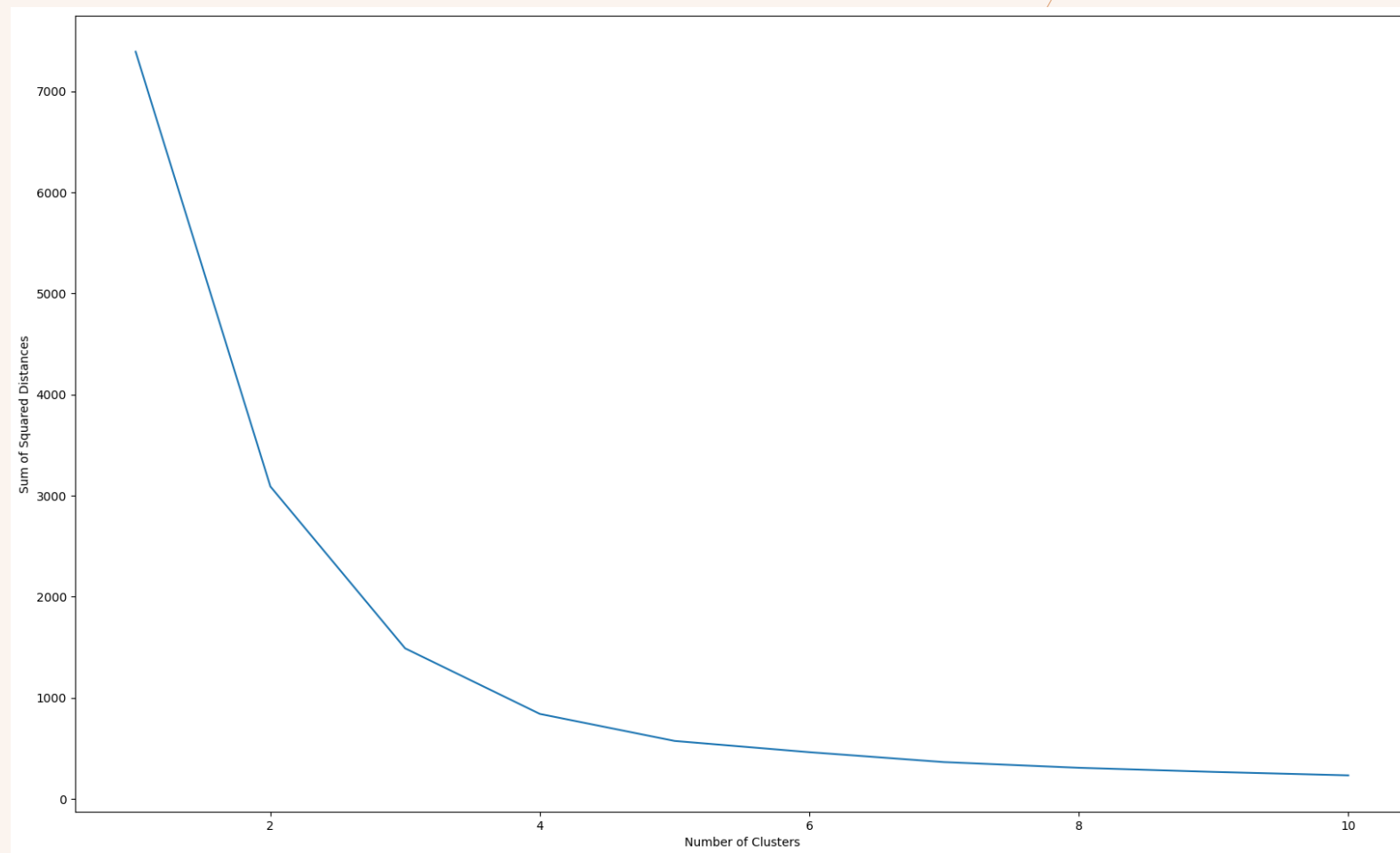
With this model, we can predict production increases with the time.



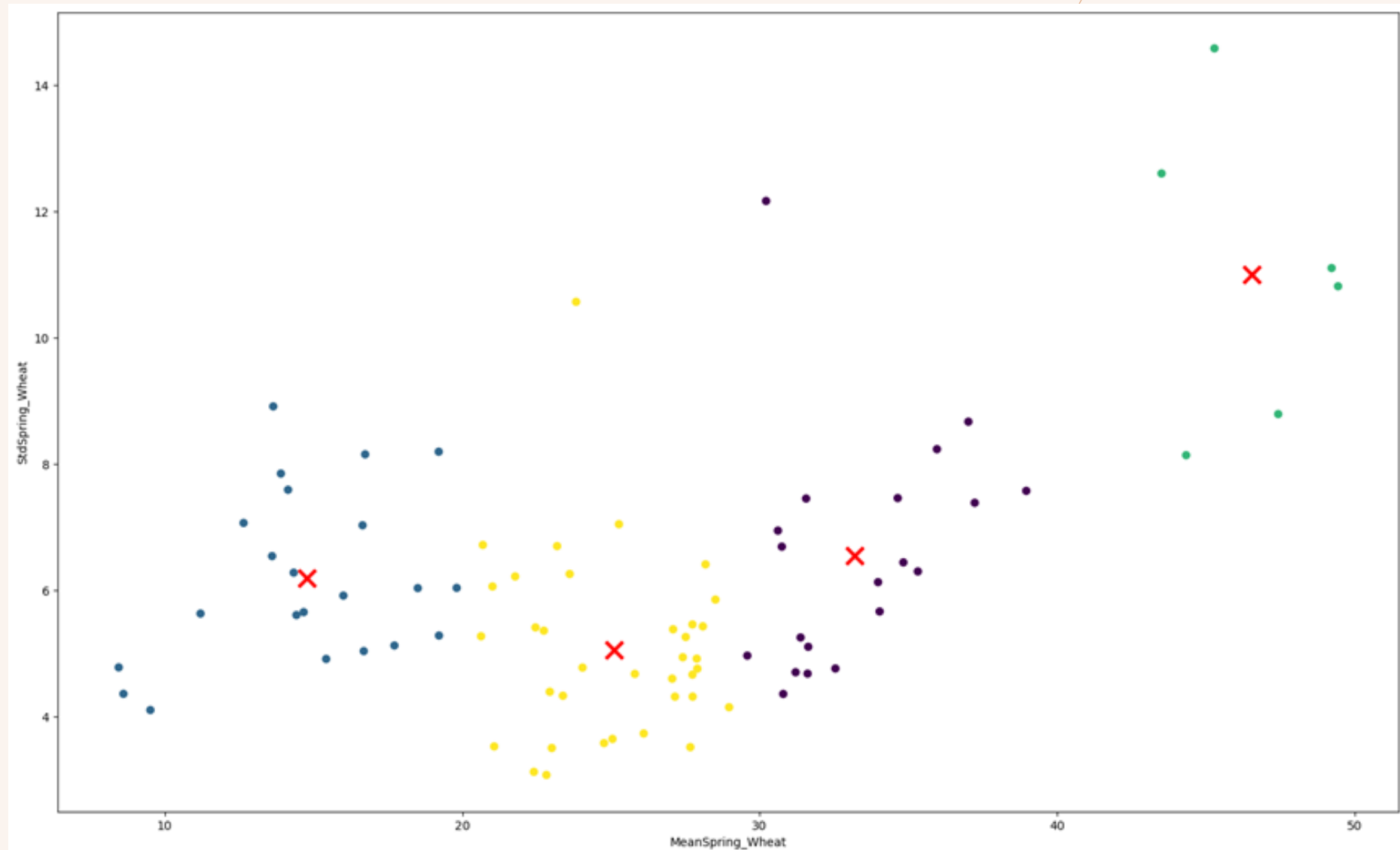


# K-MEANS CLUSTERING WITH SCIKIT-LEARN

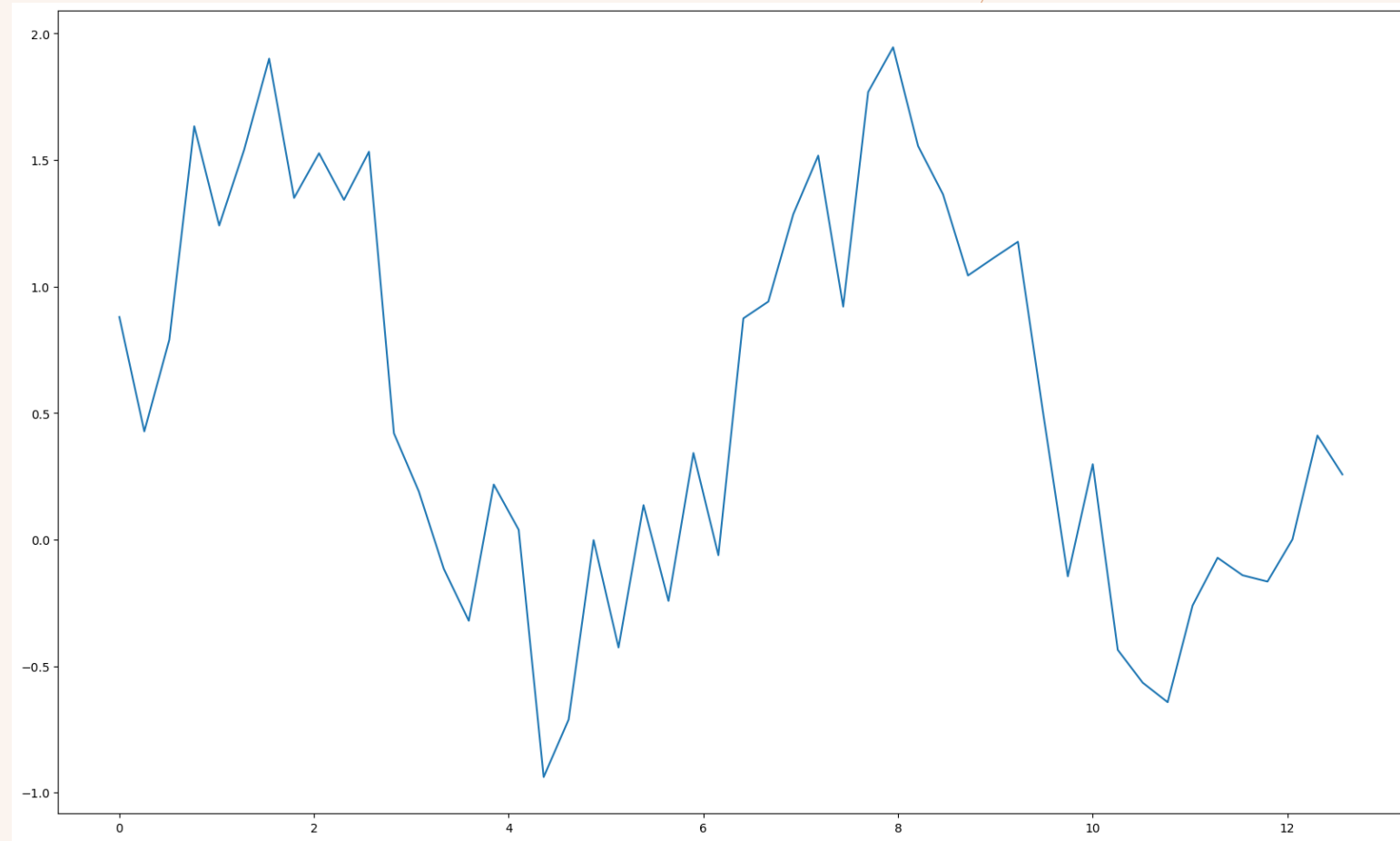
From the chart, it's clear that the “elbow” is somewhere around the k value of 4. So, let's run k-means with 4 clusters.



# K-MEANS CLUSTERS FOR SPRING\_WEAT IN RM 111, MAPLE CREEK



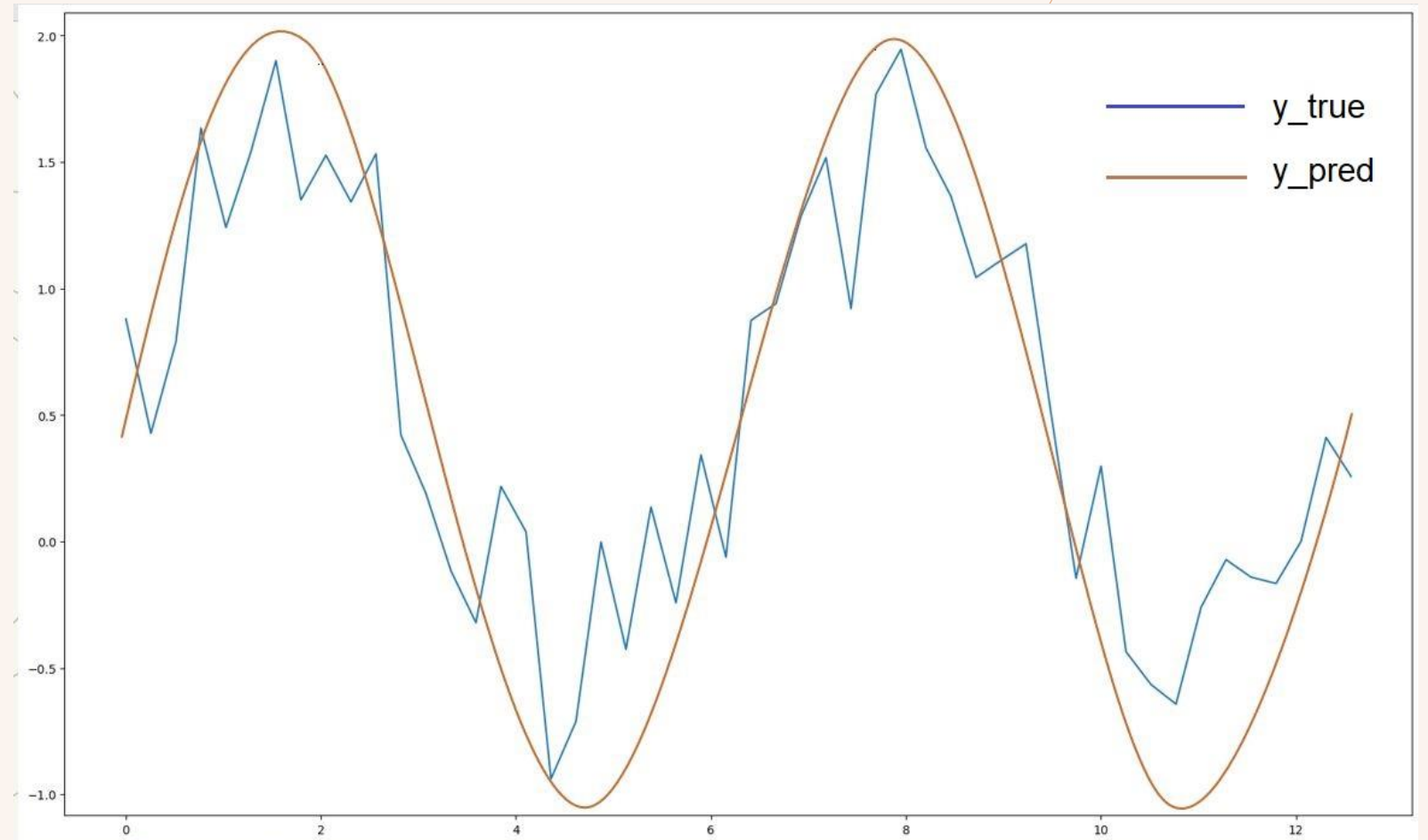
# TO FIND MEA GENERATE DATA ADDED WITH NOISE



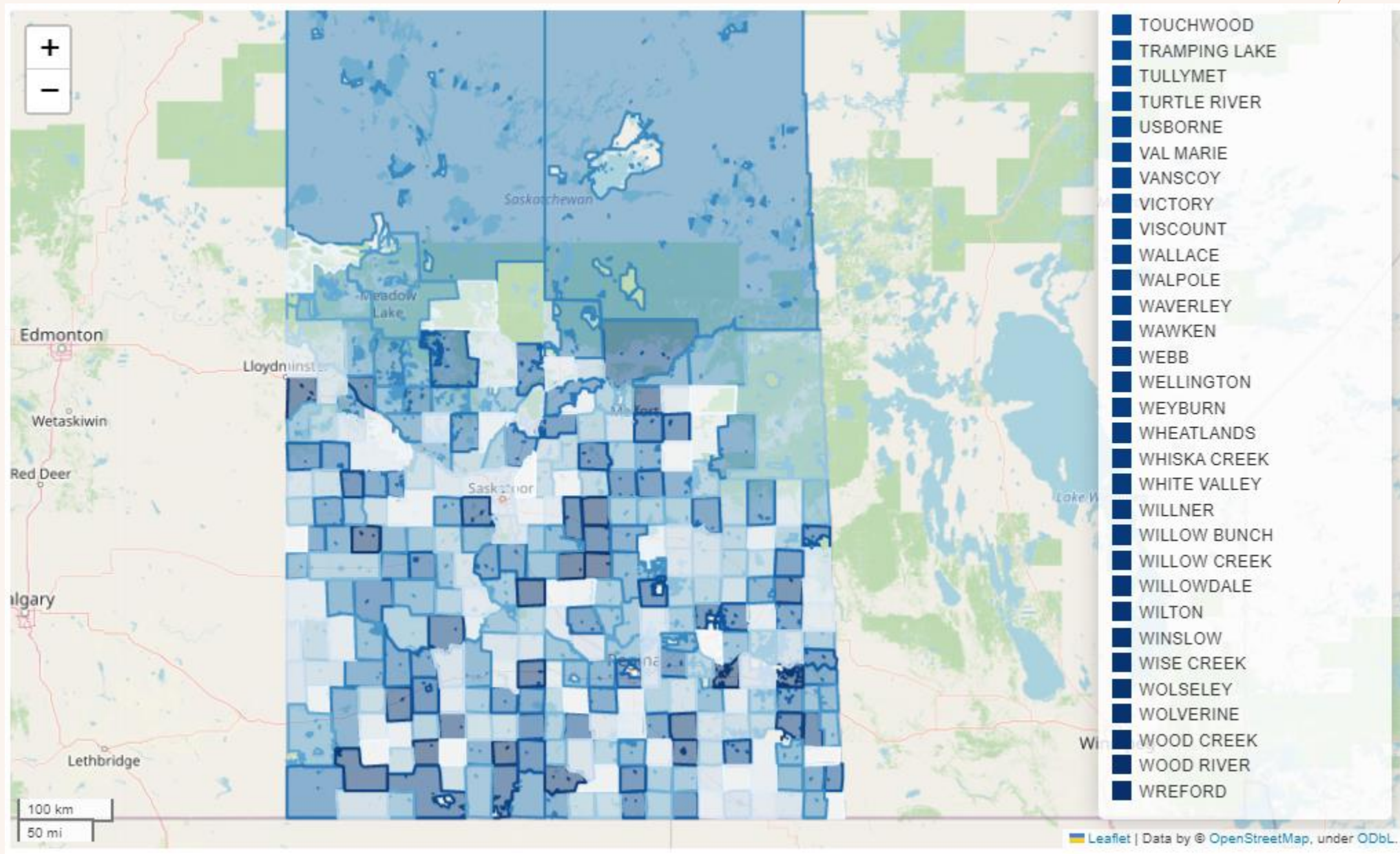
ASSUME THAT WE HAVE MADE A MODEL TO PREDICT THE Y VALUES  
FOR EACH X IN OUR DATASET.  
SINUSOIDAL DATA WITH NOISE + PREDICTION

The model output  
with data

The mean absolute error is: 0.48



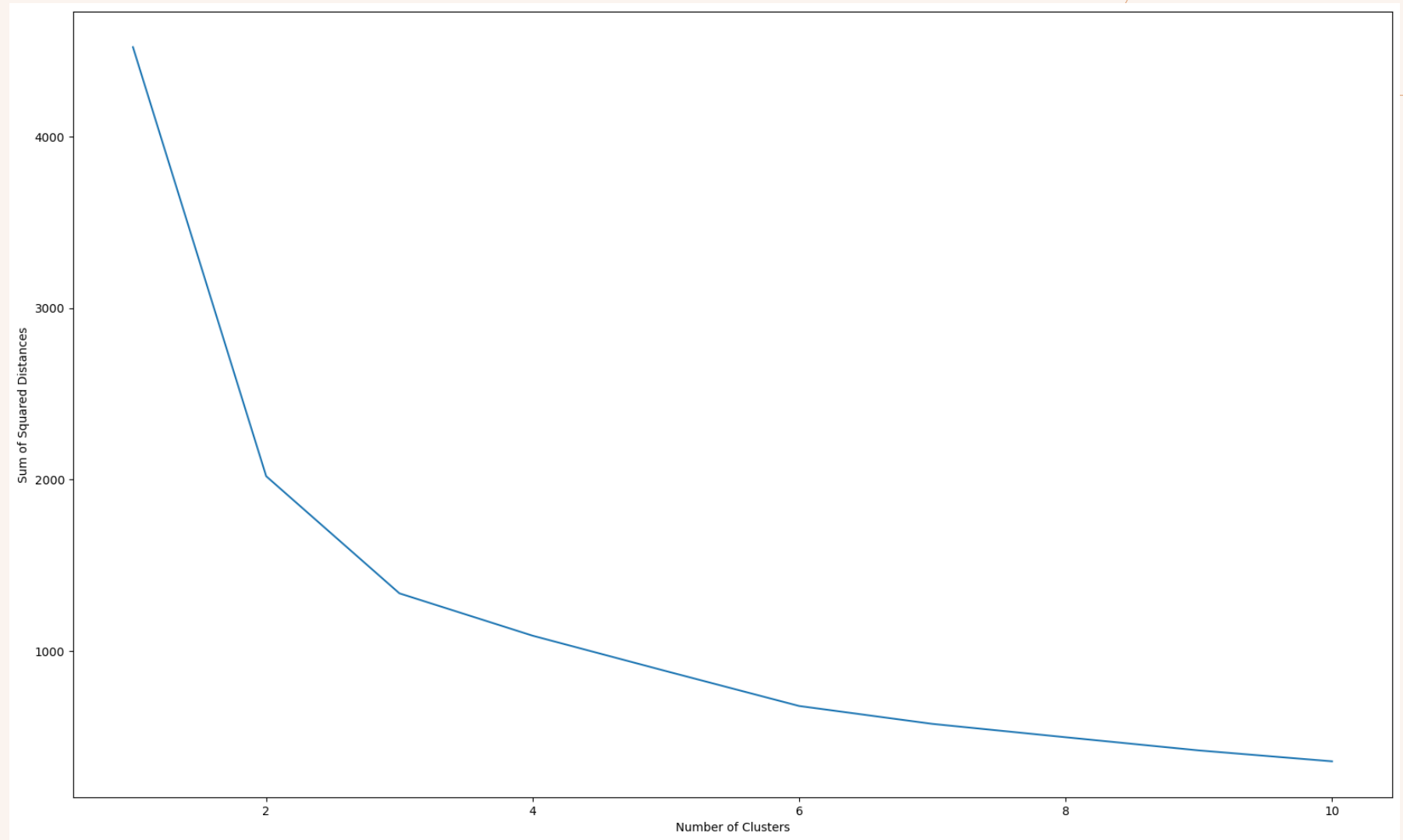
# DATA MODELING FOR SK 622 RMS



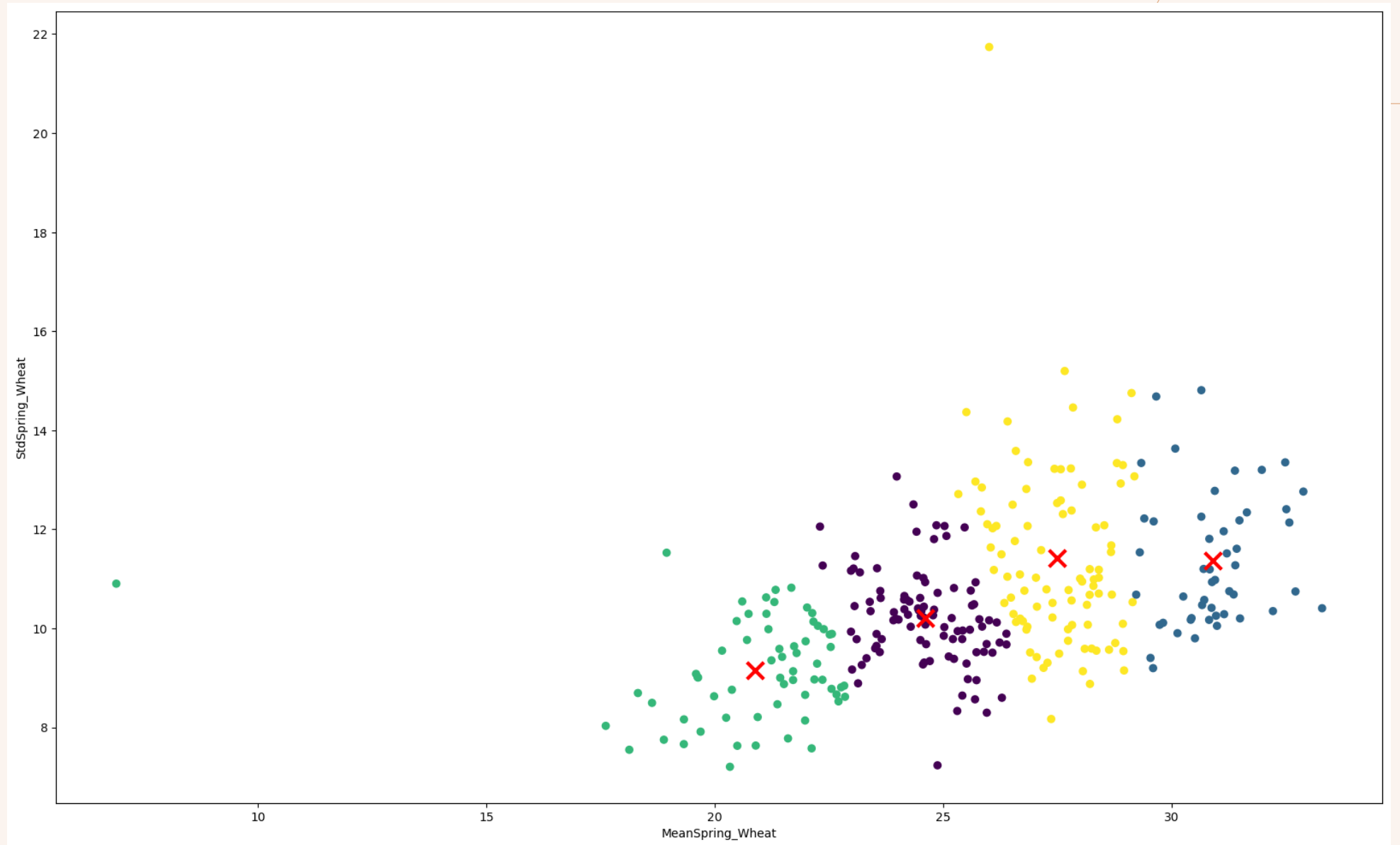


# K-MEANS CLUSTERING WITH SCIKIT-LEARN

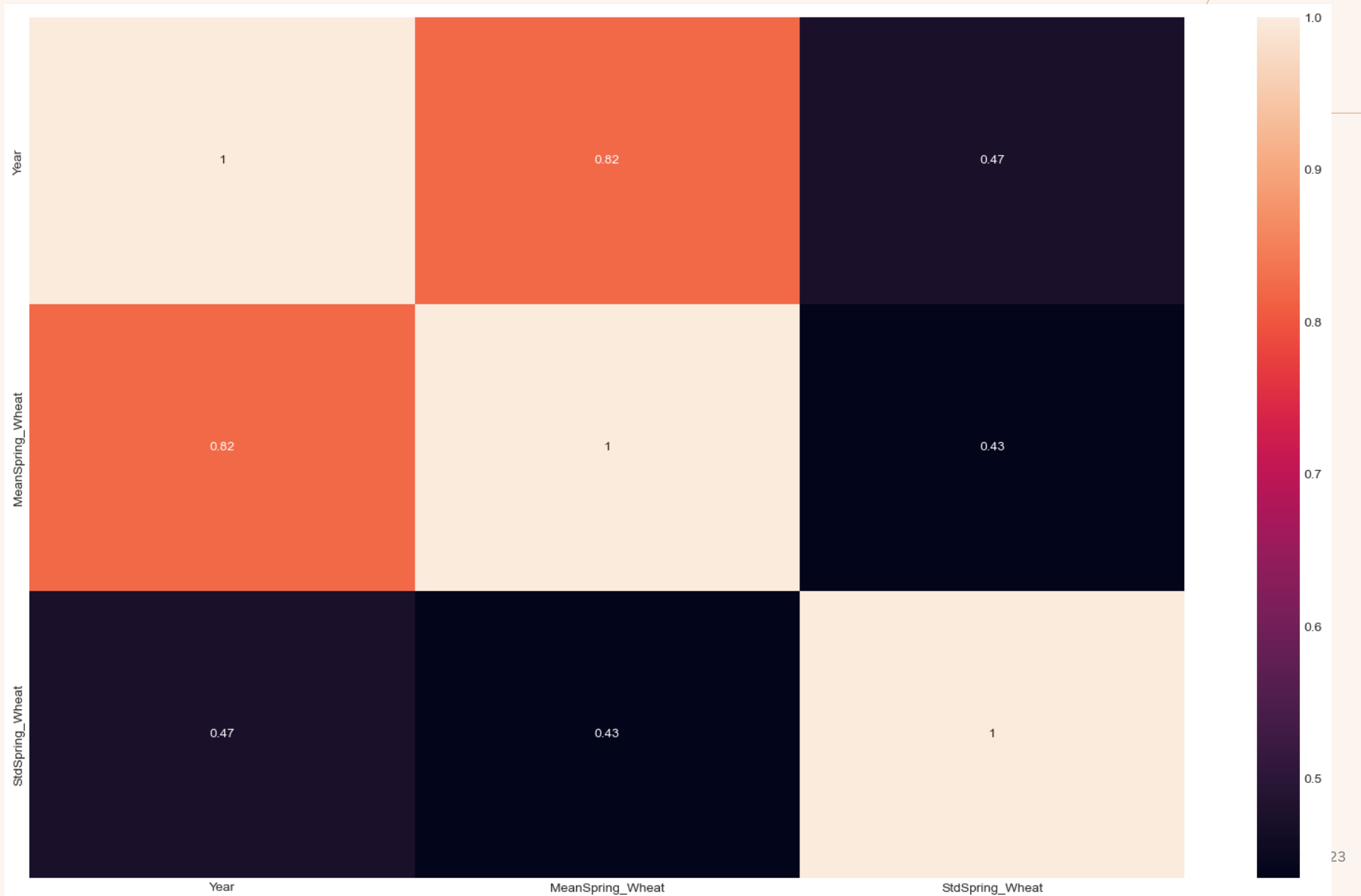
Found the “ Elbow”



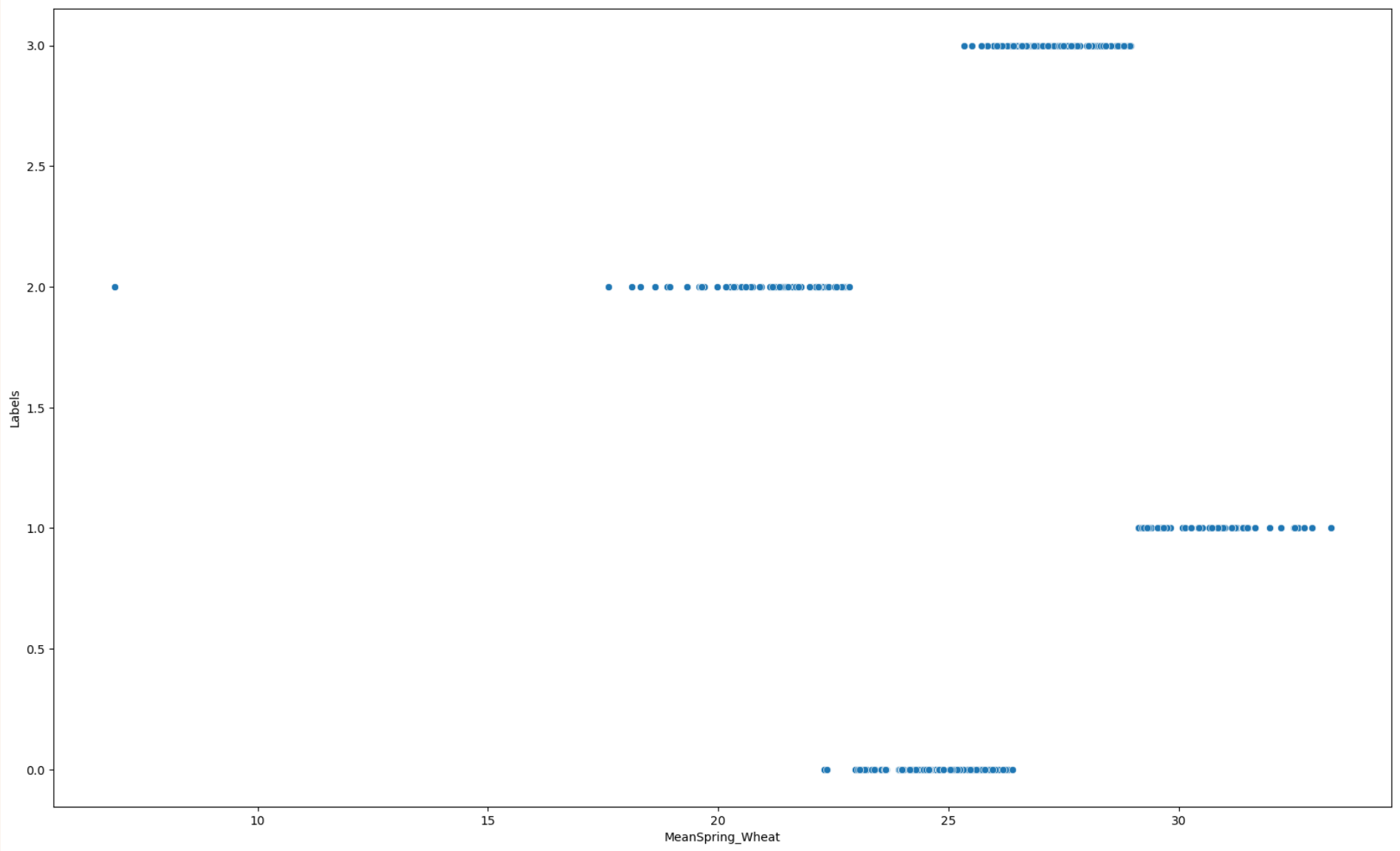
# K-MEANS CLUSTERS IN SPRING\_WHEAT FORSK 622 RMS



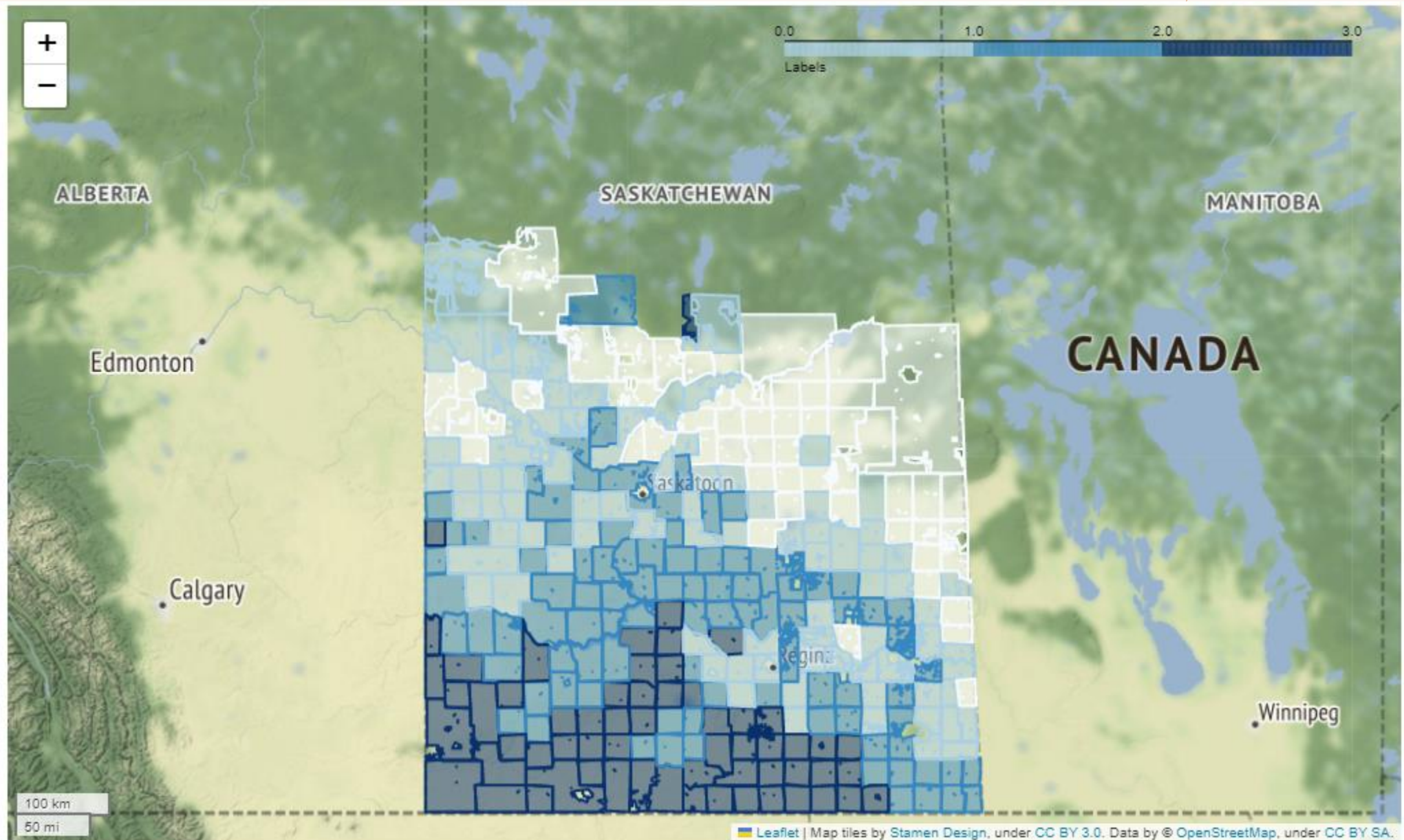
# HEATMAP FOR SPRING\_WEAT IN FOR SK 622 RMS



# SCATTER PLOT FOR SK 622 RMS(EACH CLUSTERS)



Define the  
features to be  
used for  
clustering



# RESULTS

The sizes of the four clusters are as follows: 0:  $n = 27.34$ , 1:  $n = 20.87$ ; 2:  $n = 30.66$ , 3:  $n = 24.54$ . The majority of the RMs in the south are in Cluster 3. According to the unsupervised ML, the highest value is in cluster 2. In the north SK, there is an exceptional RM, which also indicates that it belongs to class 3. However, spring wheat's high production is demonstrated in South Saskatchewan RMs.



A series of thin, light brown lines forming an abstract geometric pattern on the left side of the slide. The lines intersect to create various polygons and shapes, extending from the top left towards the center.

THANK YOU