

# PRML Assignment 1 (Team 33 )

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## Results for Assignment 1

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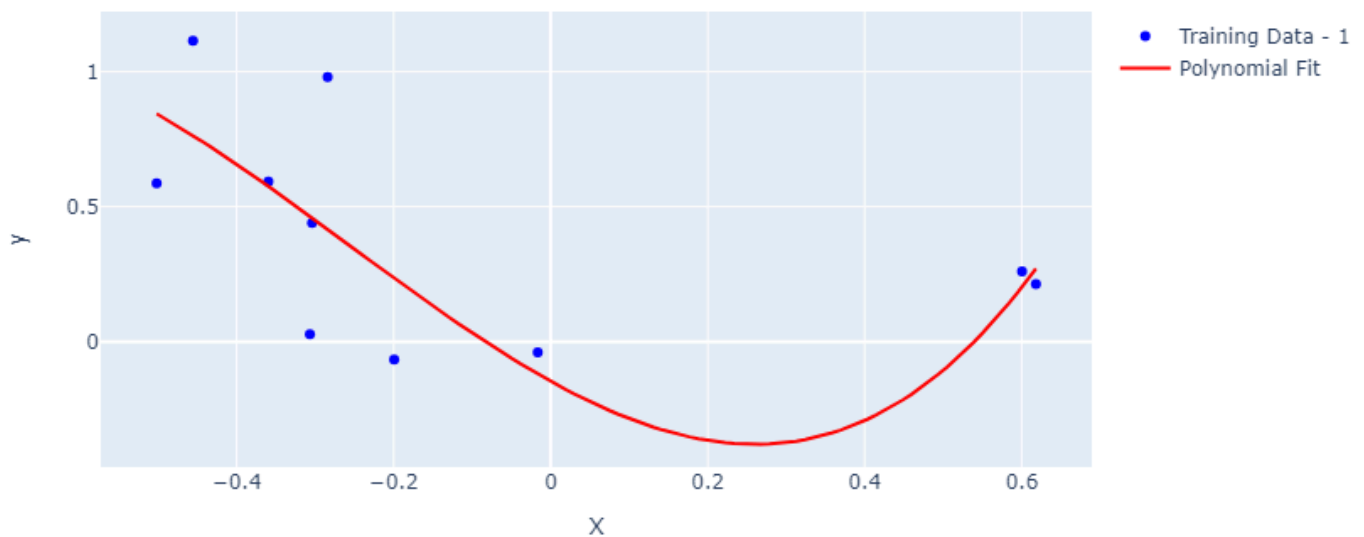
- **Regression Model:** Linear model for regression using polynomial basis functions
- **Regularization method:** Quadratic regularization

### Training Dataset 1(a): 10 examples

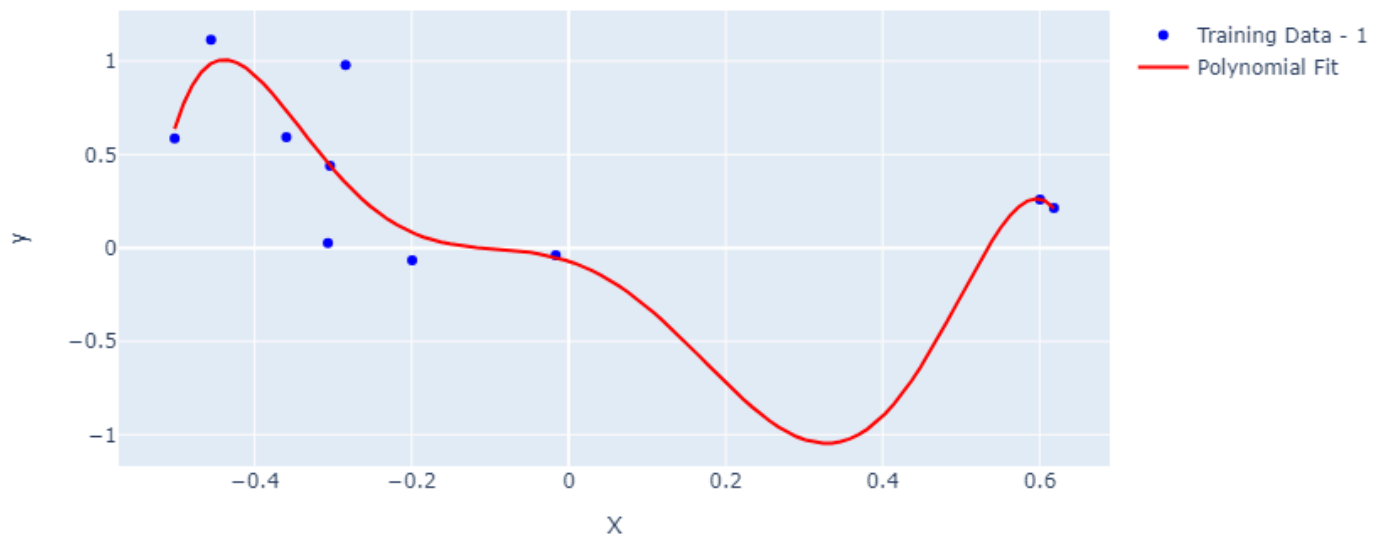
1-dimensional (Univariate) input data

Without Regularization

Polynomial Regression (degree=3, lambda=0.0) on Training Data - 1



Polynomial Regression (degree=6, lambda=0.0) on Training Data - 1

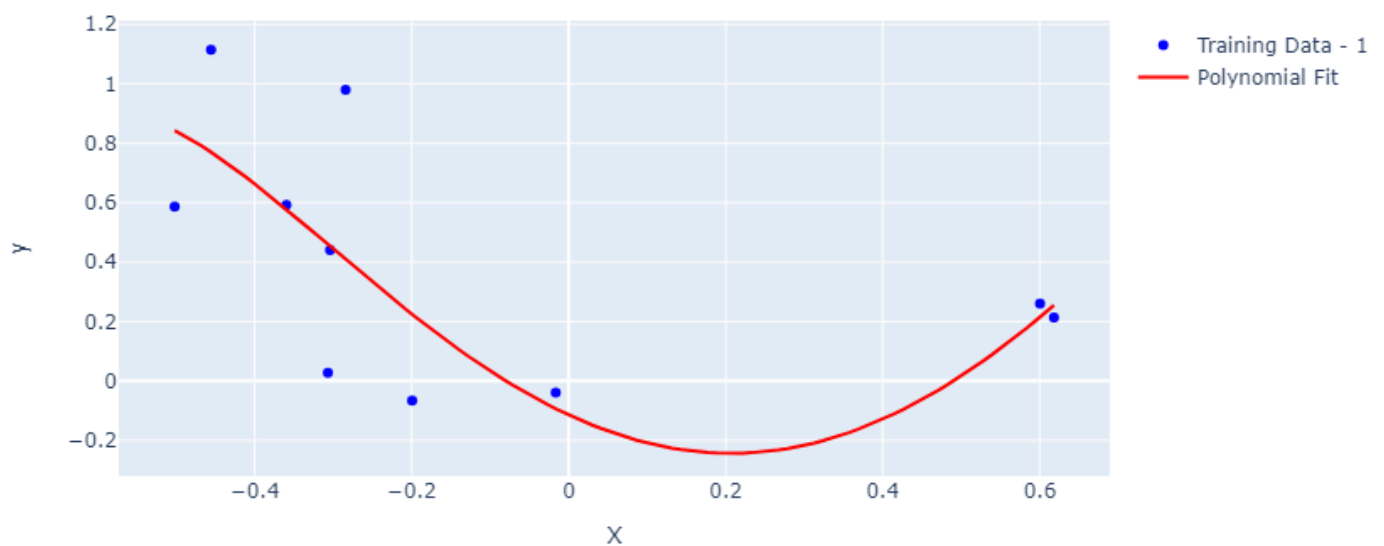


#### Observations

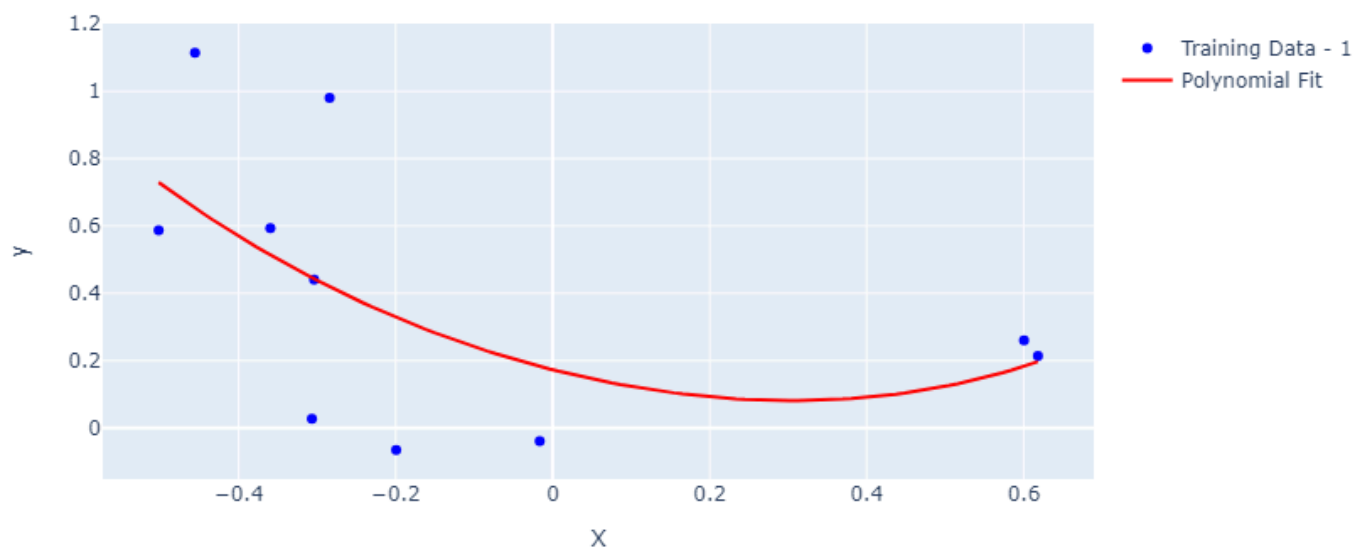
1. **Degree 3 without Regularization:** Comparable performance with regularization, slightly better in test MSE.
2. **Degree 6 without Regularization:** Severe overfitting with extremely high MSE values in validation and test.

#### With Regularization

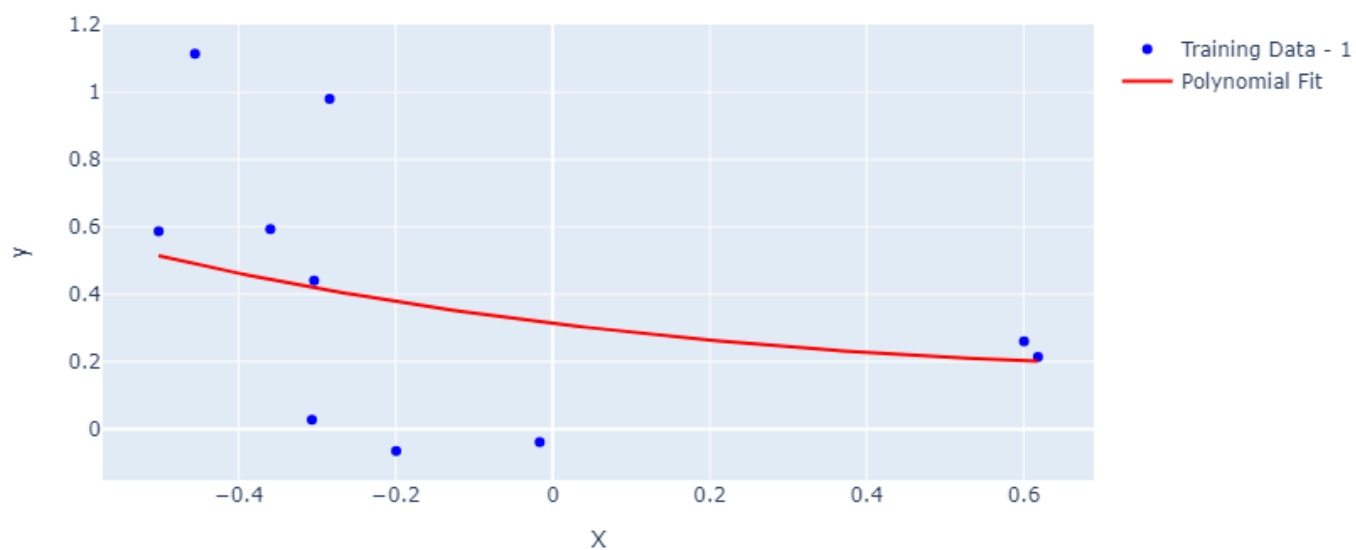
Polynomial Regression (degree=9, lambda=0.001) on Training Data - 1



Polynomial Regression (degree=9, lambda=0.1) on Training Data - 1



Polynomial Regression (degree=9, lambda=1) on Training Data - 1



#### Observations

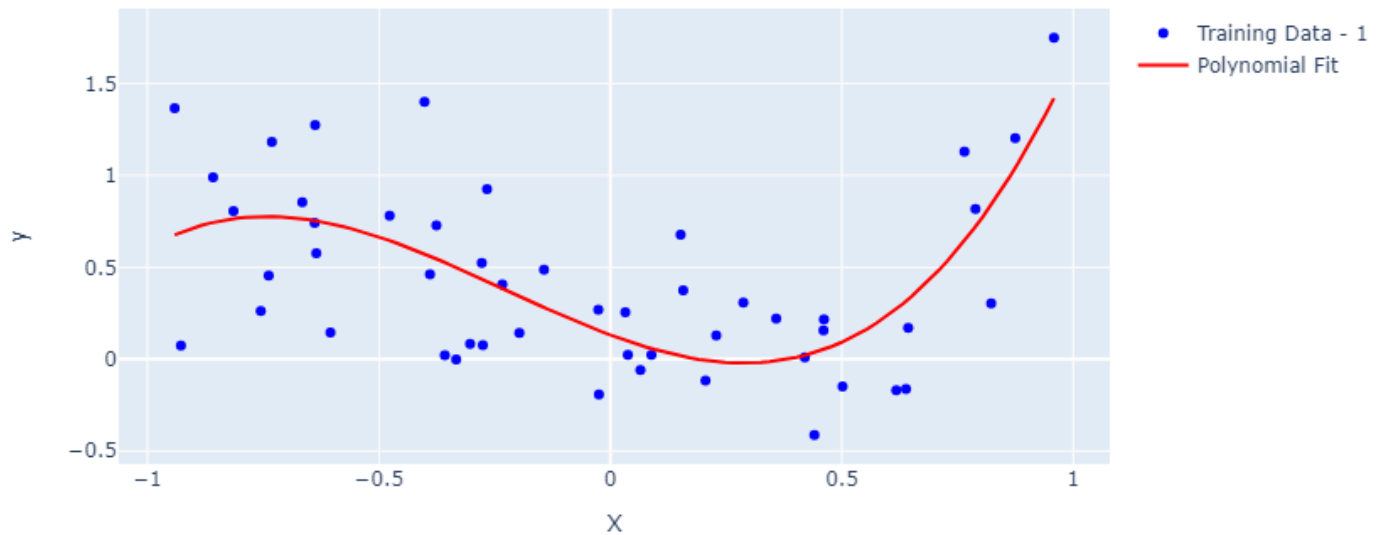
1. **Degree 9 with Regularization ( $\lambda = 0.001$ ):** Lower MSE in training, but higher in test, suggesting some overfitting.
2. **Degree 9 with Regularization ( $\lambda = 0.1$  and 1):** Increasing lambda decreases the test MSE, improving generalization.

#### Training Dataset 1(b): 50 examples

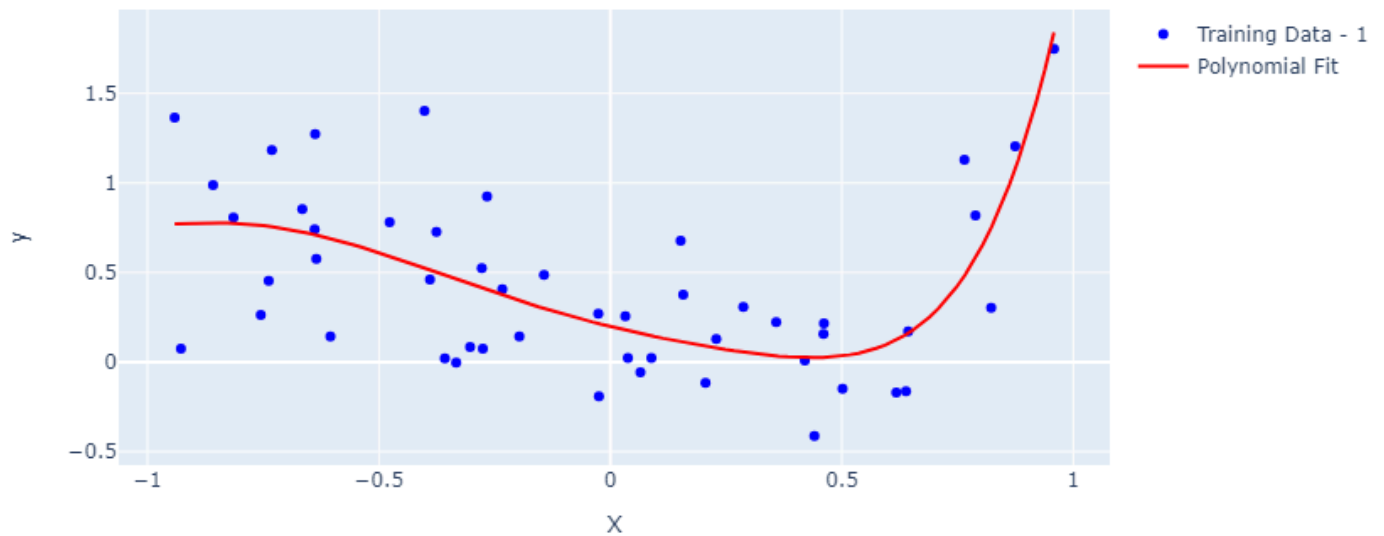
1-dimensional (Univariate) input data

## Without Regularization

Polynomial Regression (degree=3, lambda=0.0) on Training Data - 1



Polynomial Regression (degree=6, lambda=0.0) on Training Data - 1

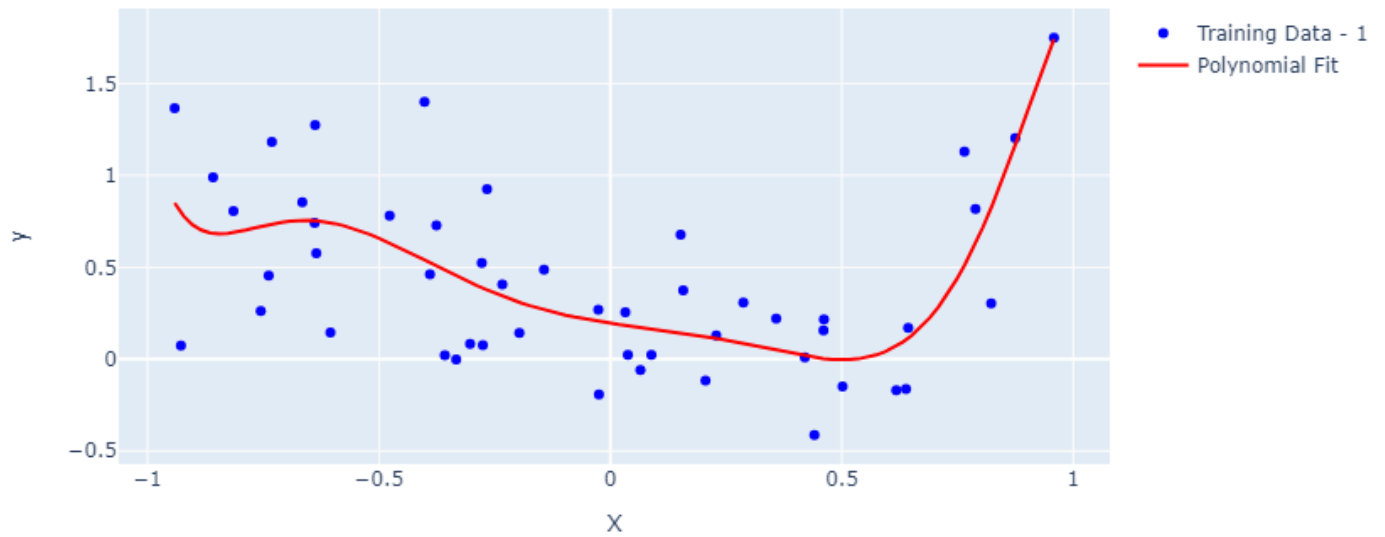


## Observations

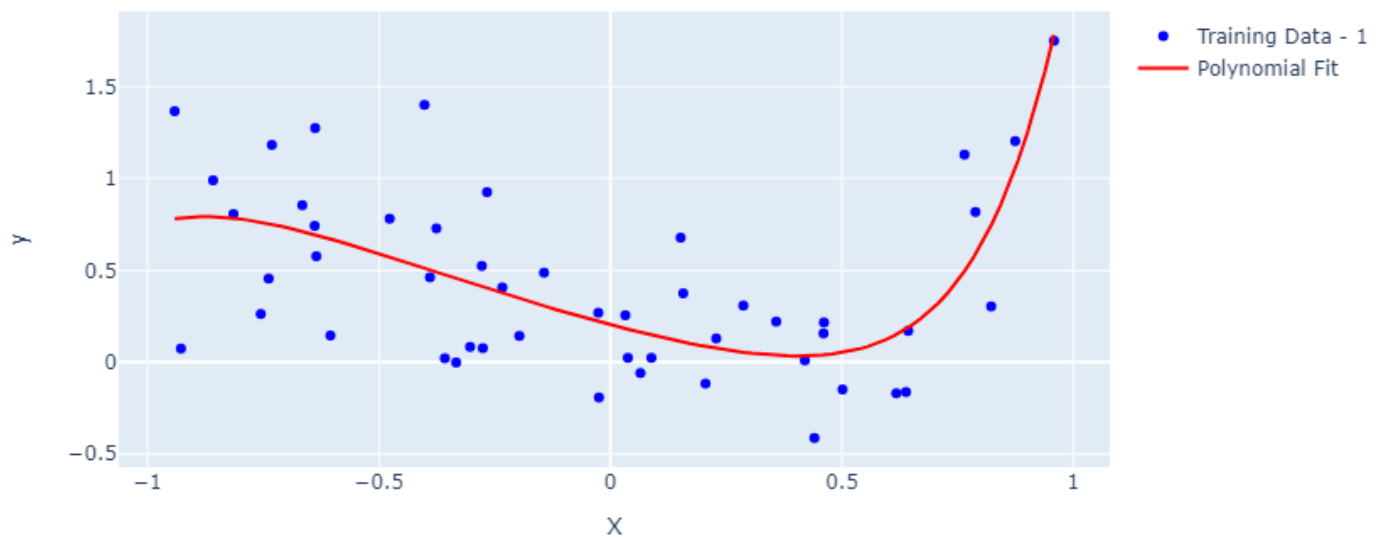
1. **Degree 3 without Regularization:** Slightly worse test performance, but better than some regularized cases.
2. **Degree 6 without Regularization:** Similar performance to Degree 9 with regularization, but slightly less effective.

## With Regularization

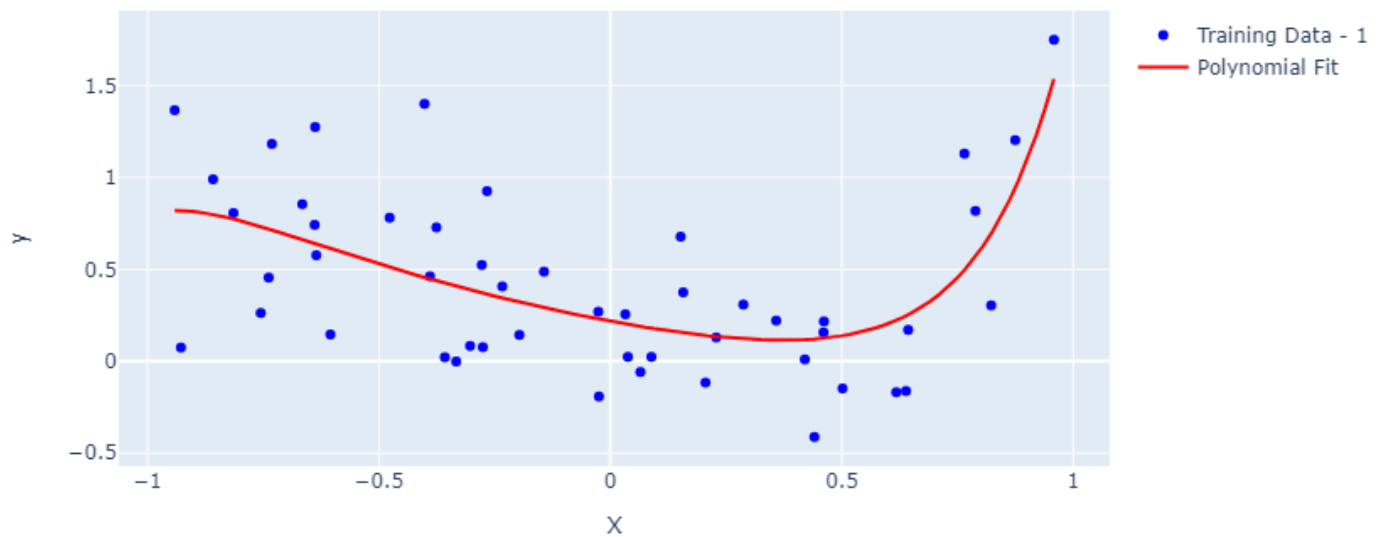
Polynomial Regression (degree=9, lambda=0.001) on Training Data - 1



Polynomial Regression (degree=9, lambda=0.1) on Training Data - 1



## Polynomial Regression (degree=9, lambda=1) on Training Data - 1



### Observations

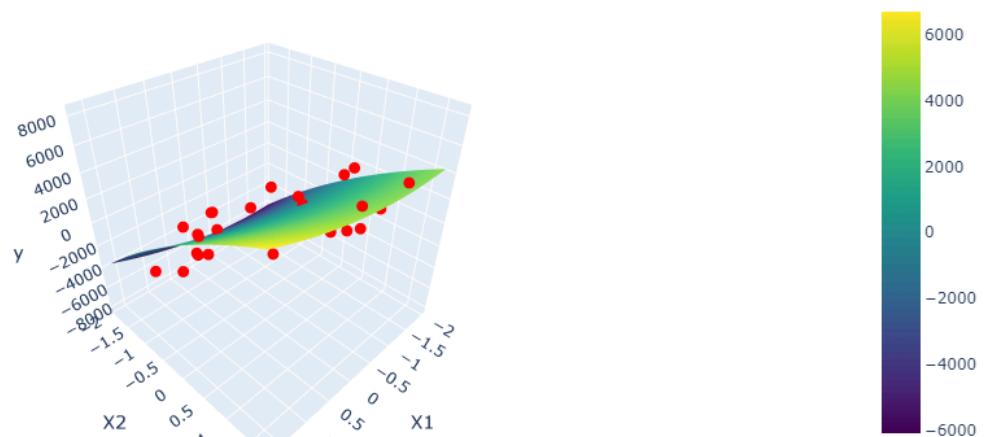
1. Degree 9 with Regularization ( $\lambda = 0.001$ ): Good generalization, with slightly better MSE in validation and test than 1a.
2. Degree 9 with Regularization ( $\lambda = 0.1$  and 1): Increasing lambda results in marginal increases in test MSE.

## Training Dataset 2(a): 25 examples

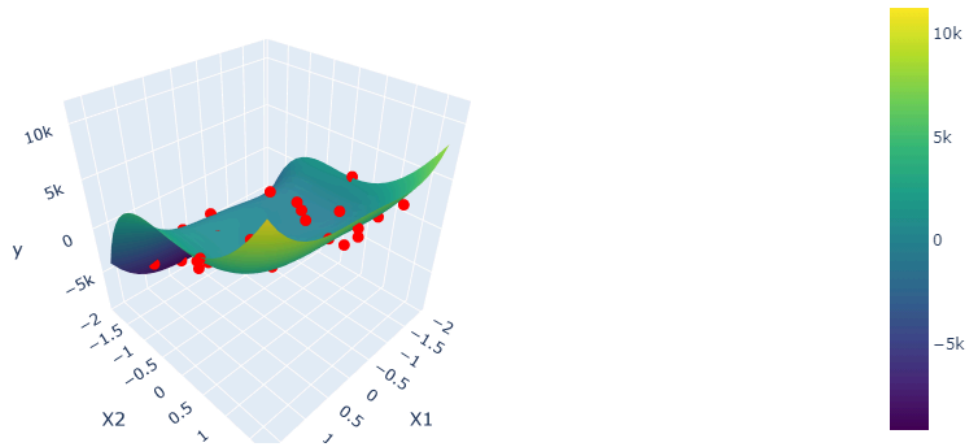
2-dimensional (Bivariate) input data

### Without Regularization

## Polynomial Regression (degree=2, lambda=0.0) on Training Data - 2a



Polynomial Regression (degree=4, lambda=0.0) on Training Data - 2a

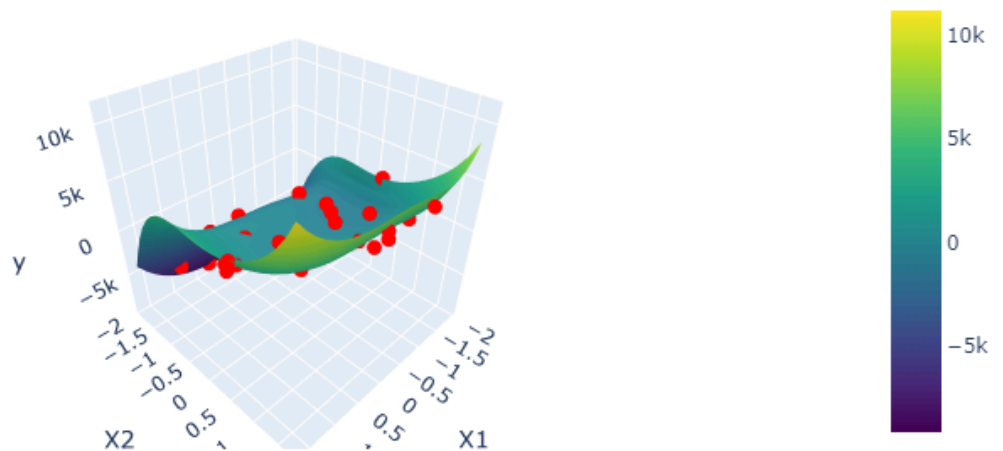


Observations

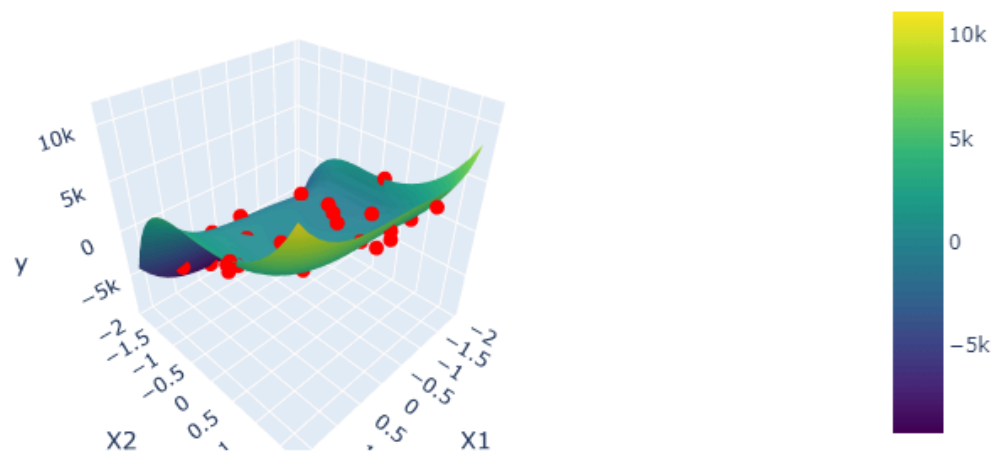
1. Degree 2 and 4 without Regularization: High MSE, showing inability to generalize.

With Regularization

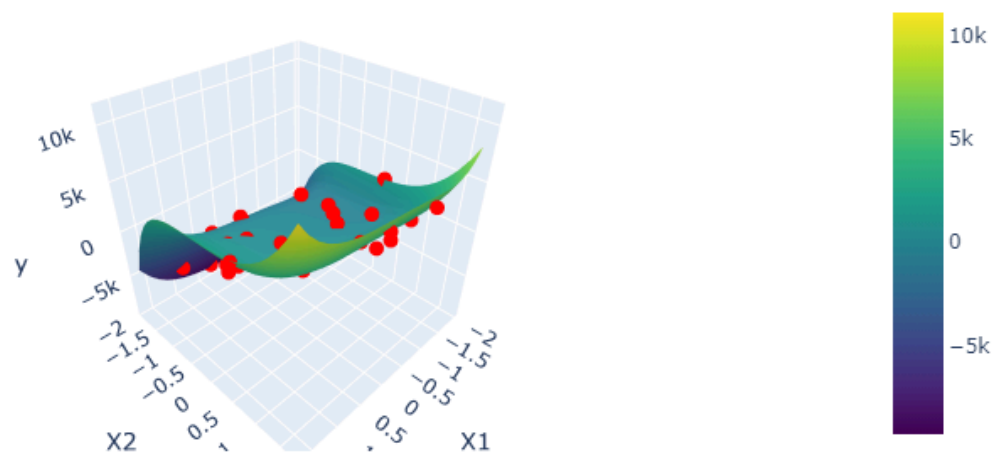
Polynomial Regression (degree=6, lambda=0.001) on Training Data - 2



## Polynomial Regression (degree=6, lambda=0.1) on Training Data - 2



## Polynomial Regression (degree=6, lambda=1) on Training Data - 2



### Observations

1. **Degree 6 with Regularization:** Extremely high MSE values across the board, suggesting poor fit.

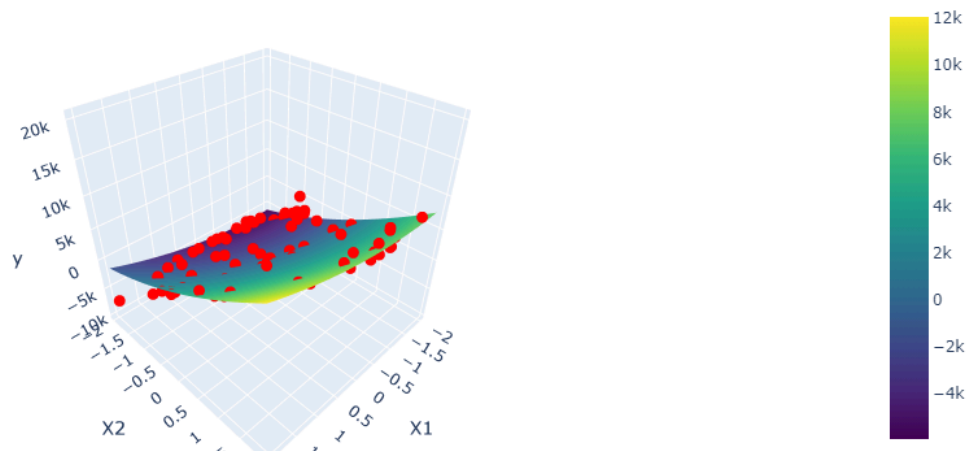
### Training Dataset 2(b): 100 examples

2-dimensional (Bivariate) input data

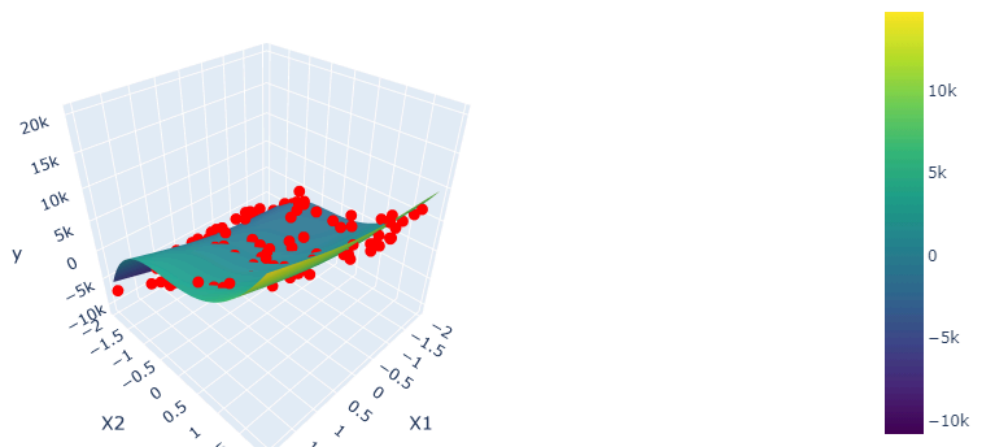


## Without Regularization

Polynomial Regression (degree=2, lambda=0.0) on Training Data - 2b



Polynomial Regression (degree=4, lambda=0.0) on Training Data - 2b

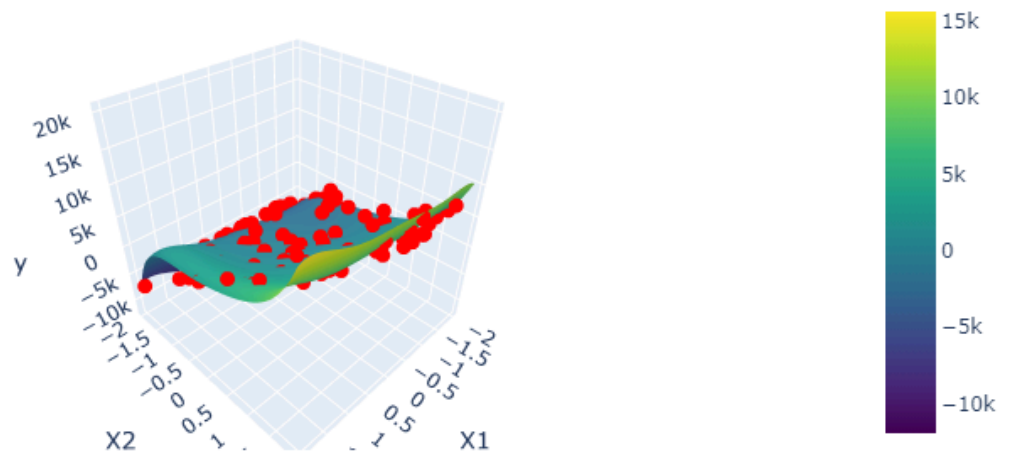


## Observations

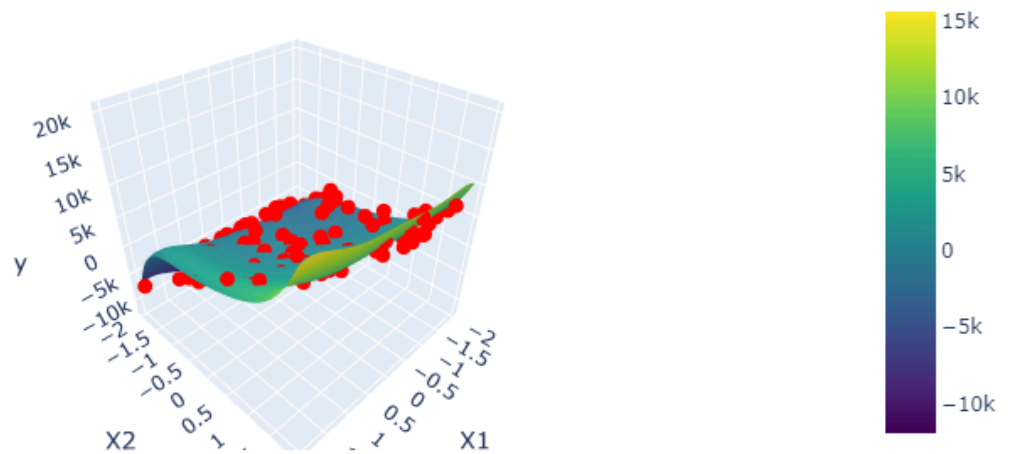
1. Degree 2 and 4 without Regularization: High MSE, similar poor generalization as in 2a.

## With Regularization

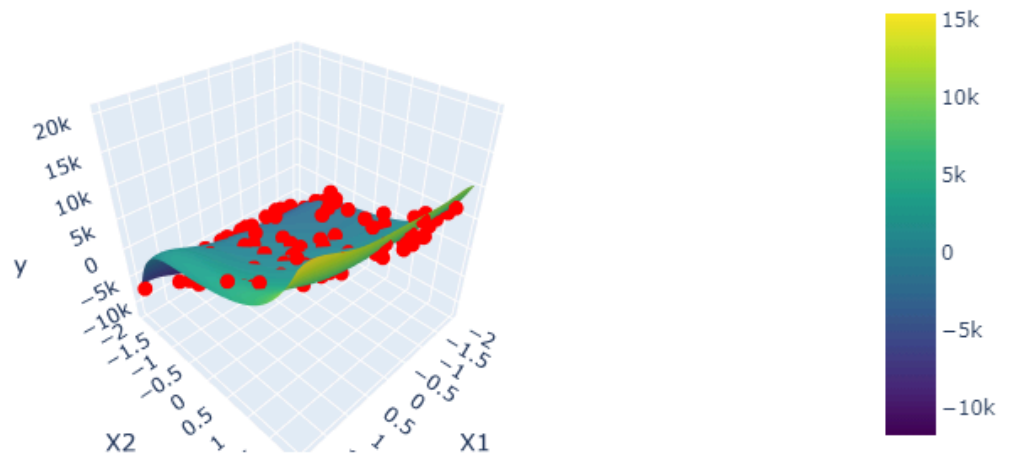
Polynomial Regression (degree=6, lambda=0.001) on Training Data - 2



Polynomial Regression (degree=6, lambda=0.1) on Training Data - 2



## Polynomial Regression (degree=6, lambda=1) on Training Data - 2



### Observations

1. **Degree 6 with Regularization:** Similar to 2a, very high MSE values.

## Dataset 3

Multivariate data

### Observations

1. **Degree 3 with Regularization:** Stable MSE values across all datasets, suggesting good generalization.
2. **Degree 2 without Regularization:** Significantly lower MSE, showing effective performance without regularization.

## E\_RMS Table

Dataset	Degree	$\lambda$	Regularization	Training $E_{RMS}$	Validation $E_{RMS}$	Test $E_{RMS}$
1a	9	0.001	Yes	0.280719	0.5396	0.90606
1a	9	0.1	Yes	0.306639	0.560843	0.662793
1a	9	1	Yes	0.350065	0.63514	0.567244
1a	3		No	0.283378	0.669069	0.578373
1a	6		No	0.25371	27.7856	42.2011
1b	9	0.001	Yes	0.338586	0.411546	0.511182
1b	9	0.1	Yes	0.344825	0.403183	0.513352
1b	9	1	Yes	0.352835	0.369344	0.527465
1b	3		No	0.355957	0.40061	0.508245
1b	6		No	0.343583	0.41459	0.51277
2a	6	0.001	Yes	843.572	4496.38	2427.39
2a	6	0.1	Yes	845.591	4640.6	2476.11
2a	6	1	Yes	857.566	4729.2	2508.36
2a	2		No	1593.56	6032.11	2520.07
2a	4		No	849.883	4849.5	2582.12
2b	6	0.001	Yes	1776.35	4310.96	1861.26
2b	6	0.1	Yes	1776.39	4316.3	1861.48
2b	6	1	Yes	1778.32	4348.22	1862.82
2b	2		No	2318.17	5335.96	2390.16
2b	4		No	1819.4	4597.95	1849.14
3	3	1e-06	Yes	5.21479	5.17761	5.20719
3	3	0.0001	Yes	5.21479	5.17761	5.20719
3	3	0.1	Yes	5.21479	5.17761	5.20719
3	2		No	2.24179	2.24129	2.30676

## Best Combinations

Here are the best combinations for each dataset as observed in the report:

1. Dataset 1(a) - 10 examples, Univariate Input Data:

- **Best Combination:** Degree 9 with Regularization ( $\lambda = 1$ )
  - **Reason:** This combination provides the lowest test Mean Squared Error (MSE) while maintaining generalization, making it the best choice.

## 2. Dataset 1(b) - 50 examples, Univariate Input Data:

- **Best Combination:** Degree 9 with Regularization ( $\lambda = 0.001$ )
  - **Reason:** This configuration shows good generalization and better MSE in validation and test compared to other regularized cases.

## 3. Dataset 2(a) - 25 examples, Bivariate Input Data:

- **Best Combination:** None of the tested combinations are particularly good, but Degree 2 or 4 without regularization might be the least poor options.
  - **Reason:** High MSE values across the board indicate poor generalization ability, even with regularization.

## 4. Dataset 2(b) - 100 examples, Bivariate Input Data:

- **Best Combination:** Similar to Dataset 2(a), none of the combinations performed exceptionally well, but lower-degree polynomials without regularization might be marginally better.

## 5. Dataset 3 - Multivariate Data:

- **Best Combinations:**
  - **Degree 3 with Regularization:** Provides stable MSE values across all datasets, indicating good generalization.
  - **Degree 2 without Regularization:** Shows significantly lower MSE, which suggests effective performance without the need for regularization.