

Precipitation Downscaling Project

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Executive Summary:

The goal of this project is to predict precipitation in Texas at time point “t+1” given temperature at time “t”. This analysis uses CPC gauge-based gridded daily precipitation and daily temperature data over roughly the area of Texas from the years 2000 to 2009 to train two different predictive models. Both models applied a principal component analysis (PCA) to downscale the high dimensional temperature data, and then fit models on the selected principal components. In the first approach, K-nearest neighbors (KNN) was applied to the principal components to predict precipitation based on a subsetting of temperature data to allow for comparison to actual values. In the second approach, a linear regression model was applied to the first principal component only to predict precipitation corresponding to the same testing temperature dataset.

In order to validate the fit of the models, mean squared error, mean absolute error, and residuals were calculated. These metrics were also used to compare the performance between the two models. Though upon visual inspection PCA-KNN appears to predict precipitation trends closer to observed values, the PCA-KNN approach exhibited a higher mean squared error and mean absolute error than the PCA-linear regression model despite a lower average residual value. Higher mean absolute error and mean squared error indicates that the PCA-KNN model actually performs worse than the PCA-linear regression model, likely due to large variations and inaccuracies in extreme values. The PCA-linear regression model, on the other hand, exhibits a conservative, “dreary” prediction across all time points, which leads to lower error overall when dealing with extremes.

Methods

Data Management/Preprocessing The original precipitation data was subsetting to only include the years 2000 to 2009 to correspond to the selected temperature data timeframe. Temperature data was subsetting to match the space occupied by the precipitation data and cover roughly the area of Texas. Latitude and longitude were adjusted to ensure consistent formatting across the datasets. To predict on precipitation at time t+1, the precipitation time points used in the analysis were shifted one day from the temperature data time points. Precipitation and temperature data were then split to create training and testing sets, with years 2000 to 2007 included in training and the rest allocated to testing. Temperature data was preprocessed by calculating the mean climatology and obtaining the

anomalies rather than working with raw temperatures to better account for seasonal variation.

1. Setup

1.1 Load Packages

2. Data

2.1 Precipitation 2.1.1 Load data

```
24x24x16365 Array{Union{Missing, Float32}, 3}:
[:, :, 1] =
  1.37582  1.57975  1.45262  ...  0.0      0.0202463  0.0
  1.20963  1.85082  2.28766      0.0308067  0.0602374  0.114836
  1.97182  1.45203  2.09293      0.0490085  0.121709  0.237292
  1.5909   1.33469  1.65327      0.0664271  0.389667  0.678864
  1.43704  1.75647  2.29071      0.0964478  0.24362   1.0564
  1.23506  1.71966  1.24682  ...  0.29584   0.192309  0.594209
  1.34688  2.06668  1.14913      0.50683   0.345075  0.153464
  1.45706  2.27098  0.724487      1.06509   0.576138  0.0801041
  5.20069  3.71962  0.181735      1.43136   0.706655  0.133022
  2.88819  0.759342  0.163341      2.08092   1.4474    0.53837
  1.50673  0.461275  0.207048  ...  missing   missing   missing
  1.43436  0.139224  0.290842      missing   missing   missing
  3.18961  2.13086  2.59861      missing   missing   missing
  5.19326  5.16296  4.65167      missing   missing   missing
  5.3778   4.47193  6.19447      missing   missing   missing
  2.81836  8.9879   5.9169   ...  missing   missing   missing
  1.80613  8.12168   2.24706      missing   missing   missing
  6.17247  15.1932   14.4421      missing   missing   missing
  15.3935  12.0708   20.1535      missing   missing   missing
  16.41    16.3307  23.8688      missing   missing   missing
  12.77    20.6869  18.5016  ...  missing   missing   missing
  14.9623  9.67723   7.38247      missing   missing   missing
  21.7209  10.5213   9.88938      missing   missing   missing
  20.7196  14.5443   15.6206      missing   missing   missing

[:, :, 2] =
  0.0      0.0      0.0      0.0      ...  0.0      0.0      0.0
  0.0      0.0      0.0      0.0      ...  0.0      0.0      0.0
  0.0      0.0      0.0      0.0      ...  0.0      0.0      0.0
  0.0      0.0      0.0      0.0      ...  0.0      0.0      0.0
  0.0      0.0      0.0      0.0      ...  0.0      0.0      0.0
  0.0      0.0      0.0      0.0      ...  0.0      0.0      0.0
  0.0      0.0      0.0      0.0      ...  0.0      0.0      0.0
  0.0      0.0      0.0      0.0      ...  0.0507107  0.0      0.0
  0.0      0.0      0.0      0.0      ...  0.170228  0.0      0.0
  0.0344644  0.0828211  0.0      0.0      ...  0.0242824  0.0      0.0
  0.116297  0.232722  0.0      0.0      ...  missing   missing   missing
  0.0780643  0.120667  0.0      0.0      ...  missing   missing   missing
```

0.790819	0.209601	0.0810387	0.0836833		missing	missing	missing
0.330909	0.0694368	0.202908	0.172532		missing	missing	missing
0.0834752	0.366478	0.193584	0.041639		missing	missing	missing
0.334877	0.392715	0.221516	0.0368422	...	missing	missing	missing
0.171142	0.847358	0.0837831	0.0		missing	missing	missing
0.660464	2.01706	0.12625	0.0		missing	missing	missing
0.309305	0.0865614	0.0	0.0546165		missing	missing	missing
0.206135	0.0	0.0	0.0464391		missing	missing	missing
0.397575	0.355726	0.0	0.0436917	...	missing	missing	missing
1.2978	0.686435	0.064151	0.364207		missing	missing	missing
0.998617	0.906296	0.373319	2.32323		missing	missing	missing
0.708568	2.16601	2.14061	2.34233		missing	missing	missing
[:, :, 3] =							
0.0	0.0	0.0	0.245373	...	0.0	0.0	0.0
0.0	0.0	0.0	0.218143		0.0	0.0	0.0
0.0	0.0	0.174088	0.175847		0.0	0.0	0.0
0.16041	0.0296272	0.77125	0.584347		0.0	0.0	0.0
0.0822869	0.0	0.0870539	0.0920338		0.0	0.0	0.0
0.0	0.0	0.0	0.0	...	0.0	0.0	0.0
0.0	0.0	0.0	0.153964		0.0	0.0	0.0
0.0	0.0	0.0	0.377731		0.0	0.0	0.0
0.0	0.0	0.0	0.0		0.0	0.0	0.0
0.391579	0.0879667	0.0	0.0		0.0	0.0	0.0
0.178373	0.052071	0.0	0.0	...	missing	missing	missing
0.0	0.0	0.0	0.0		missing	missing	missing
0.0	0.0	0.0621035	0.0487938		missing	missing	missing
0.0	0.0	0.509918	0.990183		missing	missing	missing
0.0	0.0	0.0975961	0.627761		missing	missing	missing
0.0	0.0	0.0	0.0	...	missing	missing	missing
0.0	0.0	0.0	0.0		missing	missing	missing
0.0	0.0	0.0	0.0		missing	missing	missing
0.0	0.0	0.0	0.0		missing	missing	missing
0.0	0.0	0.0	0.0	...	missing	missing	missing
0.0707157	0.0	0.0	0.0		missing	missing	missing
0.0	0.0	0.0	0.0		missing	missing	missing
0.0	0.0	0.0	0.0		missing	missing	missing
;;; ...							
[:, :, 16363] =							
0.0	0.0	0.0	0.0	...	0.0	0.0	0.0
0.0	0.0	0.0	0.0		0.0	0.0	0.0
0.0	0.0	0.0	0.0		0.0	0.0	0.0
0.0	0.0	0.0	0.0		0.0	0.0	0.0
0.0	0.0	0.0	0.0	...	0.0	0.0	0.0
0.0	0.0	0.0	0.0		0.0	0.0	0.0
0.0	0.0	0.0	0.0		0.0	0.0	0.0

0.0	0.0	0.0	0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0	...	missing	missing
0.0	0.0	0.0	0.0		missing	missing
0.0	0.0	0.0	0.0		missing	missing
0.0	0.0	0.0	0.0		missing	missing
0.0	0.0259045	0.0	0.0		missing	missing
0.0	0.0	0.207233	0.0	...	missing	missing
0.0	0.0	0.0300441	0.0		missing	missing
0.0	0.0	0.0296209	0.0		missing	missing
0.0761566	0.228387	0.441014	0.237419		missing	missing
0.275329	0.189209	0.147844	0.189911		missing	missing
0.72749	0.303631	0.0207723	0.150705	...	missing	missing
1.02891	0.220527	0.0399159	0.0950709		missing	missing
0.818845	0.127663	0.0	0.0299711		missing	missing
0.641579	0.0371985	0.0	0.0		missing	missing

[:, :, 16364] =

0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	missing	missing
0.0	2.37197	0.0	0.0	0.0	0.0	0.0	0.0	missing	missing	missing
0.0	0.683025	0.0	0.0	0.0	0.0	0.0	0.0	missing	missing	missing
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	missing	missing	missing
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	missing	missing	missing
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	missing	missing
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	missing	missing	missing
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	missing	missing	missing
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	missing	missing	missing
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	missing	missing
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	missing	missing	missing
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	missing	missing	missing
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	missing	missing	missing

[:, :, 16365] =

0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	0.0	0.0

0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0326195	0.0
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	missing	missing
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0		missing	missing
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0		missing	missing
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0		missing	missing
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0		missing	missing
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	missing	missing
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0		missing	missing
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0		missing	missing
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0		missing	missing
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0		missing	missing
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	missing	missing
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0		missing	missing
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0		missing	missing
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0		missing	missing

2.1.2 Close precip dataset

closed Dataset

2.1.3 Filter data

3653-element Vector{DateTime}:
2000-01-01T00:00:00
2000-01-02T00:00:00
2000-01-03T00:00:00
2000-01-04T00:00:00
2000-01-05T00:00:00
2000-01-06T00:00:00
2000-01-07T00:00:00
2000-01-08T00:00:00
2000-01-09T00:00:00
2000-01-10T00:00:00
2000-01-11T00:00:00
2000-01-12T00:00:00
2000-01-13T00:00:00
:
2009-12-20T00:00:00
2009-12-21T00:00:00
2009-12-22T00:00:00
2009-12-23T00:00:00
2009-12-24T00:00:00
2009-12-25T00:00:00
2009-12-26T00:00:00
2009-12-27T00:00:00
2009-12-28T00:00:00
2009-12-29T00:00:00

```
2009-12-30T00:00:00
2009-12-31T00:00:00
```

2.1.4 Subset data

```
3652-element Vector{DateTime}:
```

```
2000-01-02T00:00:00
2000-01-03T00:00:00
2000-01-04T00:00:00
2000-01-05T00:00:00
2000-01-06T00:00:00
2000-01-07T00:00:00
2000-01-08T00:00:00
2000-01-09T00:00:00
2000-01-10T00:00:00
2000-01-11T00:00:00
2000-01-12T00:00:00
2000-01-13T00:00:00
2000-01-14T00:00:00
:
2009-12-20T00:00:00
2009-12-21T00:00:00
2009-12-22T00:00:00
2009-12-23T00:00:00
2009-12-24T00:00:00
2009-12-25T00:00:00
2009-12-26T00:00:00
2009-12-27T00:00:00
2009-12-28T00:00:00
2009-12-29T00:00:00
2009-12-30T00:00:00
2009-12-31T00:00:00
```

2.1.5 Reverse latitude

```
24×24×3652 Array{Union{Missing, Float32}, 3}:
```

```
[:, :, 1] =
```

0.0	0.0	0.0	0.0	...	0.0	0.0	0.0
0.0	0.0	0.0	0.0		0.0	0.0	0.0
0.0	0.0	0.0	0.0		0.0	0.0	0.0
0.0	0.0	0.0	0.0		0.0	0.0	0.0
0.0	0.0	0.0	0.0		0.0	0.0	0.0
0.0	0.0	0.0	0.0	...	0.0	0.0	0.0
0.0	0.0	0.0	0.0		0.0	0.0	0.0
0.0	0.0	0.0	0.0323414		0.0	0.0	0.0
0.0	0.0	0.0	0.135582		0.0	0.0	0.0
0.0	0.0	0.0	0.151623		0.0	0.0	0.031432
missing	missing	missing	missing	...	0.436169	0.167826	0.0222956
missing	missing	missing	missing		0.113203	0.219052	0.0204078
missing	missing	missing	missing		0.0	0.0	0.0
missing	missing	missing	missing		0.0	0.0	0.0

missing	missing	missing	missing	0.0	0.0696384	0.0207805
missing	missing	missing	missing	...	0.0450895	0.0828177
missing	missing	missing	missing	0.0	0.0	0.0513665
missing	missing	missing	missing	0.0764082	0.0628494	0.156683
missing	missing	missing	missing	0.616311	0.172386	0.0214627
missing	missing	missing	missing	3.26764	2.32975	3.03718
missing	missing	missing	missing	...	3.97116	1.08259
missing	missing	missing	missing	1.08442	0.152592	0.607323
missing	missing	missing	missing	7.1412	1.6474	0.0
missing	missing	missing	missing	1.00298	0.780391	0.602726


```
[:, :, 2] =
```

0.0	0.0	0.0	...	0.0	0.861222	1.58391
0.0	0.0	0.0		0.0	0.142009	0.620014
0.0	0.0	0.0		0.0	0.0	0.0304579
0.0	0.0	0.0		0.0	0.0	0.0
0.0	0.0	0.0		0.0	0.0	0.0
0.0	0.0	0.0	...	0.0	0.0279951	0.10282
0.0	0.0	0.0		0.0	0.203494	0.390608
0.0	0.0	0.0		0.299697	1.65651	2.56886
0.0	0.0	0.0		2.07031	0.885224	1.6951
0.0	0.0	0.0		2.42535	2.01437	0.887439
missing	missing	missing	...	4.94332	2.89262	1.55625
missing	missing	missing		11.1041	5.20581	2.63938
missing	missing	missing		21.9379	7.78901	4.3323
missing	missing	missing		28.1783	15.3255	10.0802
missing	missing	missing		34.7224	28.1705	13.5422
missing	missing	missing	...	25.7909	27.0969	15.6146
missing	missing	missing		23.9934	25.7158	22.9741
missing	missing	missing		22.3714	23.8421	28.7882
missing	missing	missing		21.7186	23.8547	24.5314
missing	missing	missing		33.3132	19.215	14.6306
missing	missing	missing	...	55.1564	27.6612	8.24005
missing	missing	missing		37.3152	56.9326	26.5516
missing	missing	missing		24.6176	73.771	51.9611
missing	missing	missing		8.29554	48.9767	73.7095


```
[:, :, 3] =
```

0.0	0.0	0.0	...	0.0	0.0	0.119807
0.0	0.0	0.0		0.0640346	0.081867	0.165219
0.0	0.0	0.0		0.039643	0.118587	0.431731
0.0	0.0	0.0		0.0	0.0213695	0.271837
0.0	0.0	0.0		0.0226516	0.0476442	0.131516
0.0	0.0	0.0	...	0.0273123	0.0882924	0.112214
0.0	0.0	0.0		0.0	0.243156	0.652184
0.0	0.0	0.0		0.0254786	0.409578	1.41301
0.0	0.0	0.0		0.141537	1.47574	2.68291
0.0	0.0	0.0		0.590031	0.550419	2.79478
missing	missing	missing	...	1.84155	0.927045	2.13219
missing	missing	missing		1.7651	0.957743	2.1216

```

missing missing missing 1.64877 0.612912 2.59903
missing missing missing 0.780635 0.469386 1.95597
missing missing missing 1.67376 1.71056 1.23499
missing missing missing ... 0.883307 1.35247 1.12839
missing missing missing 1.0367 1.32528 3.07755
missing missing missing 4.9913 2.11696 0.937954
missing missing missing 10.4429 7.33287 2.03677
missing missing missing 8.51849 7.07041 4.33941
missing missing missing ... 10.3645 10.9053 8.96606
missing missing missing 16.6359 18.2572 10.5232
missing missing missing 7.94858 18.4997 17.5984
missing missing missing 8.29728 18.0127 25.2627

;;; ...

[:, :, 3650] =
1.91927 2.18294 4.04477 5.20947 ... 0.0 0.0 0.0
0.203245 0.580783 2.77413 4.15098 0.0 0.0 0.0
0.237774 1.28198 1.22886 0.721624 0.0 0.0 0.0
1.80391 2.13242 1.09698 0.152618 0.0 0.0 0.0
2.32314 1.18486 0.259382 0.0 0.0 0.0 0.0
1.09869 0.124012 0.0 0.0 ... 0.0 0.0 0.0
0.493284 0.0336855 0.0 0.0 0.0 0.0 0.0
0.0 0.0 0.0 0.0 0.210678 0.0 0.0
0.0 0.0 0.0 0.0 0.244225 0.0478664 0.0
0.0 0.0 0.0 0.0 0.883143 1.56837 0.156057
missing missing missing missing ... 1.22156 1.51986 1.36312
missing missing missing missing 2.79113 3.31666 2.07084
missing missing missing missing 2.48622 2.86405 1.8507
missing missing missing missing 0.999645 7.01725 4.73033
missing missing missing missing 1.48617 2.80321 4.4185
missing missing missing missing ... 0.311555 0.0501361 0.0426737
missing missing missing missing 0.123578 0.0 0.0
missing missing missing missing 0.0 0.0 0.0
missing missing missing missing 0.0 0.0 0.0
missing missing missing missing 0.0 0.0 0.0
missing missing missing missing ... 0.0 0.0 0.0
missing missing missing missing 0.0 0.0 0.0
missing missing missing missing 0.0 0.0 0.0
missing missing missing missing 0.0 0.0 0.0

[:, :, 3651] =
1.25381 1.38226 6.78581 9.23103 ... 0.209643 0.06951 0.0322176
1.77286 1.17652 5.44695 8.42261 0.674334 0.0 0.038588
5.26336 2.10804 3.33561 5.37528 0.916897 0.0899668 0.0
4.11361 1.87916 1.61745 5.03068 0.337145 0.891777 0.0281657
1.10992 0.986878 1.11829 6.67025 3.08982 0.966748 0.270218
2.73123 2.4278 3.59609 5.59022 ... 10.1729 1.6837 0.420492
3.33995 4.18317 3.80484 3.17682 0.469495 1.02329 0.587488
5.67861 5.78922 4.2321 1.34415 0.723784 1.02258 1.35627

```


5.94113	5.36736	3.45825	5.4516		1.6775	1.66165	2.03233
5.672	4.33358	3.37151	5.49953		2.08825	1.21806	2.91972
missing	missing	missing	missing	...	2.48373	1.6086	1.16387
missing	missing	missing	missing		2.07946	1.88346	0.863437
missing	missing	missing	missing		2.36113	1.62808	0.568243
missing	missing	missing	missing		1.84581	1.57609	1.48535
missing	missing	missing	missing		1.86876	1.23442	0.884178
missing	missing	missing	missing	...	1.79652	0.287498	1.20721
missing	missing	missing	missing		2.98471	0.613985	1.21469
missing	missing	missing	missing		2.01611	0.473238	0.555291
missing	missing	missing	missing		1.43036	0.545582	0.477171
missing	missing	missing	missing		0.386973	0.315955	0.0261992
missing	missing	missing	missing	...	0.0	0.0310814	0.0
missing	missing	missing	missing		0.0	0.0293388	0.0
missing	missing	missing	missing		0.0	0.0	0.0
missing	missing	missing	missing		0.0	0.0	0.0


```
[:, :, 3652] =
```

0.0	0.0	0.0	...	0.0	0.194001	0.229313
0.0	0.0	0.0		0.201	0.0	1.05024
0.0	0.0	0.0		0.794761	0.0	0.239922
0.0	0.0	0.0		0.487525	0.0712316	0.0
0.0	0.0	0.0		0.871535	0.725623	0.0546665
0.0	0.0	0.0	...	0.63043	0.834709	0.935677
0.0	0.0	0.0		0.737369	0.396731	1.7525
0.0	0.0	0.0342669		0.810175	0.914279	2.05469
0.0	0.0	0.418512		0.482352	0.6477	0.834548
0.0	0.0364437	0.121333		1.36826	1.77949	1.08005
missing	missing	missing	...	1.58758	4.78649	4.96679
missing	missing	missing		3.17118	5.32441	3.90305
missing	missing	missing		3.59475	4.90733	4.06902
missing	missing	missing		1.31501	5.03513	4.85771
missing	missing	missing		1.21495	2.24849	4.07878
missing	missing	missing	...	1.81574	0.2997	1.09506
missing	missing	missing		1.08286	0.0417713	0.649026
missing	missing	missing		0.0439509	0.0	0.253759
missing	missing	missing		0.0449181	0.0	1.20432
missing	missing	missing		0.264041	0.0	1.47429
missing	missing	missing	...	0.779127	0.235859	0.301635
missing	missing	missing		1.42092	0.159816	0.394416
missing	missing	missing		3.98366	0.940115	1.16965
missing	missing	missing		0.545287	0.645726	1.59136

2.1.6 Convert lon to lon1 format

```
24-element Vector{Float32}:
-101.75
-101.25
-100.75
-100.25
```

```
-99.75
-99.25
-98.75
-98.25
-97.75
-97.25
-96.75
-96.25
-95.75
-95.25
-94.75
-94.25
-93.75
-93.25
-92.75
-92.25
-91.75
-91.25
-90.75
-90.25
```

2.2 Temperature

2.2.1 Load data

```
Dict{Int64, Array{Union{Missing, Float64}, 3}} with 10 entries:
 2003 => [282.275 282.693 ... 291.855 291.843; 282.332 282.663 ... 291.725
291.825...
 2004 => [277.587 277.948 ... 292.118 292.505; 277.904 277.883 ... 292.192
292.599...
 2001 => [282.277 282.732 ... 291.394 291.7; 282.241 282.585 ... 291.294
291.704; ...
 2005 => [275.953 277.028 ... 291.309 291.787; 276.401 277.778 ... 291.375
291.761...
 2008 => [280.872 281.42 ... 290.851 291.7; 280.859 281.331 ... 290.545 291.318;
.....
 2000 => [279.491 280.057 ... 290.199 290.589; 279.531 280.12 ... 290.026
290.425;...
 2002 => [281.375 281.722 ... 292.495 292.614; 281.622 281.596 ... 291.817
292.024...
 2009 => [276.942 277.0 ... 290.54 290.873; 276.963 277.073 ... 290.276 290.623;
.....
 2006 => [281.257 281.87 ... 291.109 291.009; 281.398 281.981 ... 290.75
291.098; ...
 2007 => [282.586 283.225 ... 290.564 290.928; 282.62 283.199 ... 290.448
290.725;...
```

2.2.2 Combine

```
66×27×3653 Array{Union{Missing, Float64}, 3}:
[:, :, 1] =
```

279.491	280.057	280.607	280.917	...	289.866	290.023	290.199	290.589
279.531	280.12	280.553	281.077		289.713	289.881	290.026	290.425
279.28	280.175	280.651	281.016		289.702	289.758	289.926	290.237
276.178	279.946	280.693	280.98		289.692	289.883	290.018	290.189
273.58	279.514	280.61	280.935		289.735	290.065	290.055	290.214
277.067	276.623	280.496	281.015	...	289.657	289.969	289.991	290.13
273.442	275.95	276.05	279.896		289.52	289.897	289.96	290.041
269.142	278.684	276.2	278.425		289.458	289.821	289.979	290.127
268.997	273.748	277.069	274.384		289.484	289.773	289.944	290.203
268.419	268.099	269.77	271.571		289.59	289.863	289.963	290.202
268.58	265.679	268.572	271.113	...	289.57	289.863	289.99	290.169
266.085	266.211	269.975	272.546		289.445	289.784	289.933	290.244
264.757	265.626	270.774	273.479		289.245	289.528	289.771	290.1
⋮				⋮			⋮	
254.899	260.748	262.287	264.109		294.643	296.035	296.605	296.791
255.745	260.112	261.232	262.814	...	295.342	296.097	296.557	296.485
255.495	260.011	261.017	262.453		295.462	295.966	296.195	296.044
254.944	258.618	259.911	260.492		295.328	295.66	295.582	295.705
253.234	258.312	259.692	258.167		295.162	295.313	294.967	295.856
252.139	256.79	259.484	258.771		295.033	294.944	294.888	296.067
251.687	256.465	258.677	260.193	...	295.004	294.775	295.226	296.419
251.198	253.888	259.967	259.31		294.953	294.755	295.744	296.741
250.48	264.074	255.333	255.737		295.092	295.23	295.971	296.872
251.777	261.876	254.365	256.103		295.154	295.015	295.976	297.019
260.38	254.461	255.862	257.353		294.93	294.985	296.056	297.025
261.702	255.281	261.385	260.49	...	294.927	295.222	296.28	296.949
[:, :, 2] =								
277.785	278.38	279.159	280.1	...	289.92	290.165	290.567	290.996
277.716	278.507	279.004	280.103		289.754	290.153	290.443	290.939
277.222	278.548	279.249	279.895		289.765	290.133	290.447	290.816
273.583	278.09	279.134	279.774		289.607	290.211	290.484	290.76
270.987	277.941	278.876	279.638		289.527	290.125	290.331	290.682
275.653	273.91	279.103	279.61	...	289.446	289.938	290.117	290.521
271.838	272.446	273.632	277.845		289.418	289.856	290.03	290.381
267.648	276.67	274.12	276.756		289.386	289.792	289.96	290.179
267.639	272.867	275.305	272.49		289.276	289.648	289.91	290.173
267.453	266.356	268.245	270.097		289.245	289.578	289.889	290.131
266.813	263.233	269.017	273.323	...	289.218	289.583	289.882	290.17
265.723	264.585	269.935	274.421		289.139	289.587	289.797	290.14
264.8	265.987	270.392	273.674		289.145	289.484	289.656	290.002
⋮				⋮			⋮	
252.019	257.008	262.167	267.321		295.426	296.038	296.136	296.542
252.789	256.945	262.659	267.498	...	295.654	296.031	296.384	296.576
252.661	257.45	262.604	267.244		295.556	295.94	296.381	296.708
252.84	257.821	262.894	267.353		295.407	295.767	296.323	296.741
253.045	258.766	262.798	267.269		295.214	295.655	296.287	296.516
253.765	258.141	263.144	267.657		295.108	295.629	296.099	296.47
254.068	259.351	264.516	268.582	...	294.967	295.444	295.739	296.327
254.841	260.144	266.131	268.704		294.756	295.213	295.681	296.433

254.337	265.358	264.149	267.794		294.815	295.29	295.731	296.415
255.676	264.265	264.281	267.633		295.117	295.411	295.79	296.309
262.506	259.107	264.341	267.95		295.309	295.681	295.984	296.424
264.02	260.764	267.133	269.137	...	295.281	295.695	296.087	296.276

[:, :, 3] =

280.828	281.828	282.706	282.996	...	289.165	289.84	290.385	290.991
280.566	281.458	282.289	282.929		288.986	289.586	290.269	290.903
279.681	281.088	281.875	282.584		288.987	289.367	290.123	290.615
275.337	280.444	281.437	282.134		288.897	289.346	290.005	290.397
271.402	278.565	280.973	281.78		288.852	289.269	289.858	290.285
275.433	274.647	280.394	281.478	...	288.788	289.247	289.601	290.126
271.669	272.666	274.223	278.408		288.709	289.104	289.461	290.028
267.078	275.862	273.902	276.648		288.533	288.929	289.433	289.914
266.979	270.142	274.794	273.056		288.451	288.853	289.353	289.877
266.514	265.201	267.255	269.509		288.502	288.914	289.375	289.875
265.282	261.934	267.185	273.036	...	288.627	289.053	289.379	289.887
264.581	261.941	267.859	273.294		288.549	288.959	289.247	289.778
264.328	264.095	269.003	272.317		288.527	288.925	289.253	289.673
:				⋮			:	
253.715	256.142	259.555	264.432		296.103	296.319	296.661	297.081
254.471	256.343	259.741	265.014	...	296.04	296.285	296.455	297.006
254.527	257.199	259.464	265.527		295.941	296.159	296.246	296.607
255.055	258.063	260.314	266.014		295.842	295.988	296.177	296.391
255.322	259.694	261.859	266.884		295.564	295.896	296.202	296.499
255.733	258.6	261.346	267.86		295.304	295.887	296.299	296.556
255.324	258.223	263.353	267.943	...	295.314	295.824	296.36	296.645
255.92	261.045	265.461	269.195		295.348	295.96	296.451	296.689
255.555	266.834	263.836	269.951		295.533	296.057	296.495	296.772
258.12	265.24	265.328	269.328		295.661	296.118	296.474	296.605
264.055	260.288	267.399	270.726		295.79	296.078	296.431	296.375
263.982	262.179	269.563	272.556	...	295.731	295.869	296.299	296.307

;;; ...

[:, :, 3651] =

281.317	281.631	282.055	282.475	...	290.048	290.55	291.19	291.712
281.224	281.523	281.881	282.373		289.75	290.295	290.943	291.675
280.949	281.472	281.794	282.254		289.503	290.161	290.857	291.609
277.409	280.979	281.704	282.156		289.555	290.214	290.808	291.403
272.98	280.201	281.459	281.905		289.799	290.338	290.839	291.307
276.845	275.065	280.384	281.192	...	289.797	290.269	290.817	291.452
270.635	273.917	274.727	276.575		289.674	290.24	290.813	291.307
261.242	275.57	275.241	274.124		289.558	290.273	290.784	291.108
260.93	273.561	274.782	273.303		289.557	290.208	290.711	290.94
264.997	266.236	264.513	266.479		289.322	289.894	290.56	290.772
266.316	263.749	267.508	268.063	...	289.229	289.762	290.265	290.666
265.258	266.517	269.136	269.25		289.265	289.864	290.177	290.521
265.879	266.18	270.629	269.587		289.216	289.707	290.056	290.52
:				⋮			:	

253.361	252.705	252.822	255.428		293.386	294.378	295.293	296.202
254.197	253.381	254.299	256.346	...	293.853	294.744	295.7	296.497
254.319	255.045	255.828	257.537		294.197	295.051	295.971	296.718
254.705	257.057	258.031	259.485		294.546	295.325	296.153	296.919
255.491	259.084	258.295	260.563		294.833	295.686	296.518	297.227
257.814	258.49	257.719	263.299		295.112	295.83	296.529	296.991
259.65	260.214	262.346	264.149	...	295.214	295.795	296.449	297.058
262.037	264.046	265.519	264.893		295.595	296.247	296.771	297.537
261.705	269.298	264.476	266.474		295.95	296.84	297.451	298.056
265.553	268.9	265.503	265.871		296.193	297.139	297.758	298.104
271.039	264.844	269.122	268.14		296.43	297.167	297.752	298.103
272.069	266.797	271.938	271.56	...	296.539	297.341	297.835	298.206

[:, :, 3652] =

280.924	281.381	281.465	281.648	...	290.272	290.567	290.694	290.878
280.926	281.114	281.336	281.738		290.04	290.36	290.518	290.89
280.686	280.924	281.356	281.76		289.859	290.191	290.436	290.894
276.745	280.729	281.195	281.712		289.776	290.063	290.448	290.886
273.125	279.941	281.124	281.767		289.607	289.957	290.539	290.992
276.986	275.677	280.511	281.58	...	289.408	289.944	290.686	291.071
271.443	274.938	275.913	277.995		289.437	290.092	290.683	290.935
264.532	276.826	275.71	275.997		289.46	290.124	290.51	290.797
264.48	273.902	275.87	273.69		289.297	289.85	290.338	290.639
266.745	266.052	266.847	268.699		289.037	289.586	290.106	290.673
266.968	264.586	269.273	269.912	...	288.935	289.517	289.817	290.912
266.727	268.063	269.462	270.111		288.909	289.293	290.075	291.102
267.132	267.341	269.07	270.322		288.931	289.572	290.579	291.034
⋮				⋮			⋮	
253.825	255.068	255.108	255.468		293.021	293.94	294.713	295.478
253.326	254.084	254.525	254.247	...	292.976	293.987	294.779	295.577
253.139	254.191	254.617	255.016		292.884	294.046	294.863	295.511
252.847	255.042	254.682	255.502		293.003	294.099	294.838	295.45
252.355	254.32	253.945	254.113		292.948	294.095	294.828	295.47
252.708	252.801	253.487	256.163		292.884	293.911	294.709	295.344
252.323	253.393	256.56	255.072	...	292.923	293.938	294.682	295.339
253.422	255.511	258.346	256.633		293.218	294.192	294.811	295.435
252.721	259.623	255.915	257.58		293.322	294.28	295.013	295.636
255.707	259.314	256.864	256.163		293.302	294.318	295.097	295.78
260.568	255.255	259.262	258.668		293.4	294.283	295.076	295.804
262.111	257.977	262.539	261.679	...	293.509	294.351	295.143	295.865

[:, :, 3653] =

280.888	281.313	282.14	283.177	...	290.665	290.918	291.42	291.863
280.888	281.057	281.772	282.987		290.509	290.748	291.252	291.796
280.65	281.023	281.667	282.724		290.354	290.692	291.184	291.715
277.474	280.886	281.462	282.516		290.291	290.545	291.052	291.491
274.398	280.377	281.522	282.359		290.31	290.447	290.85	291.264
277.832	276.912	280.86	281.848	...	290.254	290.381	290.608	291.033
273.217	276.21	276.427	278.753		290.199	290.348	290.416	290.835
268.404	277.919	276.007	277.287		290.14	290.3	290.243	290.683

267.641	273.508	277.048	273.979		289.978	290.042	290.167	290.519
268.688	267.861	269.159	269.92		289.6	289.749	290.09	290.394
268.349	264.932	269.081	268.821	...	289.425	289.668	289.92	290.361
267.662	267.312	269.093	269.592		289.326	289.697	290.011	290.348
267.469	267.457	269.342	271.291		289.411	289.792	290.118	290.491
:				⋮			:	
265.747	265.947	264.79	264.938		295.352	296.136	296.881	297.443
264.333	264.415	264.751	265.249	...	295.32	295.854	296.607	297.26
263.013	264.202	263.996	263.92		294.821	295.742	296.315	296.937
260.457	262.073	262.738	261.584		294.357	295.263	296.039	296.637
258.505	261.497	263.556	261.326		293.995	294.959	295.777	296.469
256.852	260.006	261.386	261.116		293.811	294.556	295.434	296.124
254.831	257.297	260.099	260.703	...	293.698	294.391	295.111	295.891
254.33	256.677	259.421	258.532		293.688	294.428	294.999	295.743
252.638	262.568	256.035	256.655		293.43	294.324	294.987	295.735
254.496	261.304	255.382	256.532		293.127	294.067	294.87	295.672
260.685	254.638	258.74	259.026		292.917	293.757	294.664	295.52
261.243	257.23	261.903	261.136	...	292.844	293.684	294.548	295.45

2.2.3 Save lon, lat, time, and temp variables

`open_mfdataset` (generic function with 1 method)

3653-element Vector{DateTime}:

```

2000-01-01T00:00:00
2000-01-02T00:00:00
2000-01-03T00:00:00
2000-01-04T00:00:00
2000-01-05T00:00:00
2000-01-06T00:00:00
2000-01-07T00:00:00
2000-01-08T00:00:00
2000-01-09T00:00:00
2000-01-10T00:00:00
2000-01-11T00:00:00
2000-01-12T00:00:00
2000-01-13T00:00:00
:
2009-12-20T00:00:00
2009-12-21T00:00:00
2009-12-22T00:00:00
2009-12-23T00:00:00
2009-12-24T00:00:00
2009-12-25T00:00:00
2009-12-26T00:00:00
2009-12-27T00:00:00
2009-12-28T00:00:00
2009-12-29T00:00:00
2009-12-30T00:00:00
2009-12-31T00:00:00

```

2.2.4 Flip the temperature latitude

66x27x3653 Array{Union{Missing, Float64}, 3}:

```
[:, :, 1] =  
290.589 290.199 290.023 289.866 ... 280.917 280.607 280.057 279.491  
290.425 290.026 289.881 289.713 ... 281.077 280.553 280.12 279.531  
290.237 289.926 289.758 289.702 ... 281.016 280.651 280.175 279.28  
290.189 290.018 289.883 289.692 ... 280.98 280.693 279.946 276.178  
290.214 290.055 290.065 289.735 ... 280.935 280.61 279.514 273.58  
290.13 289.991 289.969 289.657 ... 281.015 280.496 276.623 277.067  
290.041 289.96 289.897 289.52 ... 279.896 276.05 275.95 273.442  
290.127 289.979 289.821 289.458 ... 278.425 276.2 278.684 269.142  
290.203 289.944 289.773 289.484 ... 274.384 277.069 273.748 268.997  
290.202 289.963 289.863 289.59 ... 271.571 269.77 268.099 268.419  
290.169 289.99 289.863 289.57 ... 271.113 268.572 265.679 268.58  
290.244 289.933 289.784 289.445 ... 272.546 269.975 266.211 266.085  
290.1 289.771 289.528 289.245 ... 273.479 270.774 265.626 264.757  
⋮ ⋮ ⋮ ⋮ ⋮ ⋮ ⋮ ⋮  
296.791 296.605 296.035 294.643 ... 264.109 262.287 260.748 254.899  
296.485 296.557 296.097 295.342 ... 262.814 261.232 260.112 255.745  
296.044 296.195 295.966 295.462 ... 262.453 261.017 260.011 255.495  
295.705 295.582 295.66 295.328 ... 260.492 259.911 258.618 254.944  
295.856 294.967 295.313 295.162 ... 258.167 259.692 258.312 253.234  
296.067 294.888 294.944 295.033 ... 258.771 259.484 256.79 252.139  
296.419 295.226 294.775 295.004 ... 260.193 258.677 256.465 251.687  
296.741 295.744 294.755 294.953 ... 259.31 259.967 253.888 251.198  
296.872 295.971 295.23 295.092 ... 255.737 255.333 264.074 250.48  
297.019 295.976 295.015 295.154 ... 256.103 254.365 261.876 251.777  
297.025 296.056 294.985 294.93 ... 257.353 255.862 254.461 260.38  
296.949 296.28 295.222 294.927 ... 260.49 261.385 255.281 261.702  
[:, :, 2] =  
290.996 290.567 290.165 289.92 ... 280.1 279.159 278.38 277.785  
290.939 290.443 290.153 289.754 ... 280.103 279.004 278.507 277.716  
290.816 290.447 290.133 289.765 ... 279.895 279.249 278.548 277.222  
290.76 290.484 290.211 289.607 ... 279.774 279.134 278.09 273.583  
290.682 290.331 290.125 289.527 ... 279.638 278.876 277.941 270.987  
290.521 290.117 289.938 289.446 ... 279.61 279.103 273.91 275.653  
290.381 290.03 289.856 289.418 ... 277.845 273.632 272.446 271.838  
290.179 289.96 289.792 289.386 ... 276.756 274.12 276.67 267.648  
290.173 289.91 289.648 289.276 ... 272.49 275.305 272.867 267.639  
290.131 289.889 289.578 289.245 ... 270.097 268.245 266.356 267.453  
290.17 289.882 289.583 289.218 ... 273.323 269.017 263.233 266.813  
290.14 289.797 289.587 289.139 ... 274.421 269.935 264.585 265.723  
290.002 289.656 289.484 289.145 ... 273.674 270.392 265.987 264.8  
⋮ ⋮ ⋮ ⋮ ⋮ ⋮ ⋮ ⋮  
296.542 296.136 296.038 295.426 ... 267.321 262.167 257.008 252.019  
296.576 296.384 296.031 295.654 ... 267.498 262.659 256.945 252.789  
296.708 296.381 295.94 295.556 ... 267.244 262.604 257.45 252.661  
296.741 296.323 295.767 295.407 ... 267.353 262.894 257.821 252.84
```


296.516	296.287	295.655	295.214		267.269	262.798	258.766	253.045
296.47	296.099	295.629	295.108		267.657	263.144	258.141	253.765
296.327	295.739	295.444	294.967	...	268.582	264.516	259.351	254.068
296.433	295.681	295.213	294.756		268.704	266.131	260.144	254.841
296.415	295.731	295.29	294.815		267.794	264.149	265.358	254.337
296.309	295.79	295.411	295.117		267.633	264.281	264.265	255.676
296.424	295.984	295.681	295.309		267.95	264.341	259.107	262.506
296.276	296.087	295.695	295.281	...	269.137	267.133	260.764	264.02

[:, :, 3] =

290.991	290.385	289.84	289.165	...	282.996	282.706	281.828	280.828
290.903	290.269	289.586	288.986		282.929	282.289	281.458	280.566
290.615	290.123	289.367	288.987		282.584	281.875	281.088	279.681
290.397	290.005	289.346	288.897		282.134	281.437	280.444	275.337
290.285	289.858	289.269	288.852		281.78	280.973	278.565	271.402
290.126	289.601	289.247	288.788	...	281.478	280.394	274.647	275.433
290.028	289.461	289.104	288.709		278.408	274.223	272.666	271.669
289.914	289.433	288.929	288.533		276.648	273.902	275.862	267.078
289.877	289.353	288.853	288.451		273.056	274.794	270.142	266.979
289.875	289.375	288.914	288.502		269.509	267.255	265.201	266.514
289.887	289.379	289.053	288.627	...	273.036	267.185	261.934	265.282
289.778	289.247	288.959	288.549		273.294	267.859	261.941	264.581
289.673	289.253	288.925	288.527		272.317	269.003	264.095	264.328
:				:			:	
297.081	296.661	296.319	296.103		264.432	259.555	256.142	253.715
297.006	296.455	296.285	296.04	...	265.014	259.741	256.343	254.471
296.607	296.246	296.159	295.941		265.527	259.464	257.199	254.527
296.391	296.177	295.988	295.842		266.014	260.314	258.063	255.055
296.499	296.202	295.896	295.564		266.884	261.859	259.694	255.322
296.556	296.299	295.887	295.304		267.86	261.346	258.6	255.733
296.645	296.36	295.824	295.314	...	267.943	263.353	258.223	255.324
296.689	296.451	295.96	295.348		269.195	265.461	261.045	255.92
296.772	296.495	296.057	295.533		269.951	263.836	266.834	255.555
296.605	296.474	296.118	295.661		269.328	265.328	265.24	258.12
296.375	296.431	296.078	295.79		270.726	267.399	260.288	264.055
296.307	296.299	295.869	295.731	...	272.556	269.563	262.179	263.982

;;; ...

[:, :, 3651] =

291.712	291.19	290.55	290.048	...	282.475	282.055	281.631	281.317
291.675	290.943	290.295	289.75		282.373	281.881	281.523	281.224
291.609	290.857	290.161	289.503		282.254	281.794	281.472	280.949
291.403	290.808	290.214	289.555		282.156	281.704	280.979	277.409
291.307	290.839	290.338	289.799		281.905	281.459	280.201	272.98
291.452	290.817	290.269	289.797	...	281.192	280.384	275.065	276.845
291.307	290.813	290.24	289.674		276.575	274.727	273.917	270.635
291.108	290.784	290.273	289.558		274.124	275.241	275.57	261.242
290.94	290.711	290.208	289.557		273.303	274.782	273.561	260.93
290.772	290.56	289.894	289.322		266.479	264.513	266.236	264.997

290.666	290.265	289.762	289.229	...	268.063	267.508	263.749	266.316
290.521	290.177	289.864	289.265		269.25	269.136	266.517	265.258
290.52	290.056	289.707	289.216		269.587	270.629	266.18	265.879
:				:			:	
296.202	295.293	294.378	293.386		255.428	252.822	252.705	253.361
296.497	295.7	294.744	293.853	...	256.346	254.299	253.381	254.197
296.718	295.971	295.051	294.197		257.537	255.828	255.045	254.319
296.919	296.153	295.325	294.546		259.485	258.031	257.057	254.705
297.227	296.518	295.686	294.833		260.563	258.295	259.084	255.491
296.991	296.529	295.83	295.112		263.299	257.719	258.49	257.814
297.058	296.449	295.795	295.214	...	264.149	262.346	260.214	259.65
297.537	296.771	296.247	295.595		264.893	265.519	264.046	262.037
298.056	297.451	296.84	295.95		266.474	264.476	269.298	261.705
298.104	297.758	297.139	296.193		265.871	265.503	268.9	265.553
298.103	297.752	297.167	296.43		268.14	269.122	264.844	271.039
298.206	297.835	297.341	296.539	...	271.56	271.938	266.797	272.069
[: , : , 3652] =								
290.878	290.694	290.567	290.272	...	281.648	281.465	281.381	280.924
290.89	290.518	290.36	290.04		281.738	281.336	281.114	280.926
290.894	290.436	290.191	289.859		281.76	281.356	280.924	280.686
290.886	290.448	290.063	289.776		281.712	281.195	280.729	276.745
290.992	290.539	289.957	289.607		281.767	281.124	279.941	273.125
291.071	290.686	289.944	289.408	...	281.58	280.511	275.677	276.986
290.935	290.683	290.092	289.437		277.995	275.913	274.938	271.443
290.797	290.51	290.124	289.46		275.997	275.71	276.826	264.532
290.639	290.338	289.85	289.297		273.69	275.87	273.902	264.48
290.673	290.106	289.586	289.037		268.699	266.847	266.052	266.745
290.912	289.817	289.517	288.935	...	269.912	269.273	264.586	266.968
291.102	290.075	289.293	288.909		270.111	269.462	268.063	266.727
291.034	290.579	289.572	288.931		270.322	269.07	267.341	267.132
:				:			:	
295.478	294.713	293.94	293.021		255.468	255.108	255.068	253.825
295.577	294.779	293.987	292.976	...	254.247	254.525	254.084	253.326
295.511	294.863	294.046	292.884		255.016	254.617	254.191	253.139
295.45	294.838	294.099	293.003		255.502	254.682	255.042	252.847
295.47	294.828	294.095	292.948		254.113	253.945	254.32	252.355
295.344	294.709	293.911	292.884		256.163	253.487	252.801	252.708
295.339	294.682	293.938	292.923	...	255.072	256.56	253.393	252.323
295.435	294.811	294.192	293.218		256.633	258.346	255.511	253.422
295.636	295.013	294.28	293.322		257.58	255.915	259.623	252.721
295.78	295.097	294.318	293.302		256.163	256.864	259.314	255.707
295.804	295.076	294.283	293.4		258.668	259.262	255.255	260.568
295.865	295.143	294.351	293.509	...	261.679	262.539	257.977	262.111
[: , : , 3653] =								
291.863	291.42	290.918	290.665	...	283.177	282.14	281.313	280.888
291.796	291.252	290.748	290.509		282.987	281.772	281.057	280.888
291.715	291.184	290.692	290.354		282.724	281.667	281.023	280.65
291.491	291.052	290.545	290.291		282.516	281.462	280.886	277.474

291.264	290.85	290.447	290.31		282.359	281.522	280.377	274.398
291.033	290.608	290.381	290.254	...	281.848	280.86	276.912	277.832
290.835	290.416	290.348	290.199		278.753	276.427	276.21	273.217
290.683	290.243	290.3	290.14		277.287	276.007	277.919	268.404
290.519	290.167	290.042	289.978		273.979	277.048	273.508	267.641
290.394	290.09	289.749	289.6		269.92	269.159	267.861	268.688
290.361	289.92	289.668	289.425	...	268.821	269.081	264.932	268.349
290.348	290.011	289.697	289.326		269.592	269.093	267.312	267.662
290.491	290.118	289.792	289.411		271.291	269.342	267.457	267.469
⋮				⋮			⋮	
297.443	296.881	296.136	295.352		264.938	264.79	265.947	265.747
297.26	296.607	295.854	295.32	...	265.249	264.751	264.415	264.333
296.937	296.315	295.742	294.821		263.92	263.996	264.202	263.013
296.637	296.039	295.263	294.357		261.584	262.738	262.073	260.457
296.469	295.777	294.959	293.995		261.326	263.556	261.497	258.505
296.124	295.434	294.556	293.811		261.116	261.386	260.006	256.852
295.891	295.111	294.391	293.698	...	260.703	260.099	257.297	254.831
295.743	294.999	294.428	293.688		258.532	259.421	256.677	254.33
295.735	294.987	294.324	293.43		256.655	256.035	262.568	252.638
295.672	294.87	294.067	293.127		256.532	255.382	261.304	254.496
295.52	294.664	293.757	292.917		259.026	258.74	254.638	260.685
295.45	294.548	293.684	292.844	...	261.136	261.903	257.23	261.243

2.2.5 Subset temperature to area of Texas instead of US

11×11×3653 Array{Union{Missing, Float64}, 3}:								
[:, :, 1] =								
291.31	290.912	290.253	289.702	...	287.914	286.397	284.341	283.857
291.359	292.577	291.752	290.155		288.822	287.284	284.14	283.579
294.291	293.043	292.334	289.887		287.999	285.853	283.505	282.566
293.938	292.984	290.842	289.412		285.757	285.144	283.36	283.169
294.653	293.992	291.703	288.774		284.997	284.048	283.946	283.252
296.33	295.475	293.28	287.959	...	284.874	284.03	283.941	283.375
296.416	295.53	293.52	290.233		284.907	283.735	282.815	283.406
296.158	295.471	293.695	290.598		283.788	283.558	282.356	282.317
296.254	295.835	294.149	291.078		285.554	283.737	282.379	280.963
296.419	296.219	294.747	290.463		285.086	285.36	283.302	281.299
296.699	296.629	295.455	291.055	...	285.868	286.188	285.094	281.852
[:, :, 2] =								
293.682	294.663	293.619	292.196	...	287.538	285.536	283.818	281.873
294.237	296.073	295.585	293.843		289.812	288.154	285.528	283.329
296.598	295.877	296.526	294.523		291.016	288.691	286.095	284.717
295.727	296.559	295.922	294.954		291.273	289.738	287.398	286.099
295.003	294.79	294.608	295.179		292.05	290.401	288.871	287.385
296.606	296.265	294.972	295.118	...	292.334	290.859	289.587	288.703
296.577	296.147	295.147	292.247		292.261	290.22	289.503	288.252
296.431	295.958	295.31	292.818		291.934	290.013	289.351	286.932
296.602	296.168	295.596	293.396		291.917	290.509	289.26	287.116
296.901	296.689	295.971	292.78		291.516	290.681	289.67	287.883

297.139	297.121	296.391	292.873	...	290.944	290.096	289.802	288.823
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[:, :, 3] =

292.085	292.854	291.909	290.942	...	282.91	281.018	278.14	274.995
292.797	294.604	294.149	292.369		285.145	283.904	280.893	277.093
295.131	294.607	294.906	293.185		286.17	284.659	282.005	278.211
295.704	295.811	295.002	294.017		286.947	285.41	282.396	279.592
294.688	294.379	294.588	294.73		288.94	286.131	283.493	280.8
296.732	296.152	294.6	295.283	...	290.251	287.954	285.039	281.771
296.851	296.422	295.191	292.597		291.292	289.293	285.617	282.495
296.851	296.337	295.571	292.925		292.529	290.314	287.826	282.528
296.955	296.507	295.852	293.724		293.374	291.678	289.932	283.792
297.132	296.845	296.222	293.522		293.705	292.632	291.251	287.52
297.489	297.374	296.658	293.777	...	294.035	292.909	291.896	290.317

;;; ...

[:, :, 3651] =

277.637	279.352	280.108	281.083	...	273.931	273.617	275.306	274.236
279.415	281.736	281.757	281.47		273.664	273.975	275.22	274.934
281.824	281.551	281.937	280.699		273.825	271.055	273.781	275.18
283.619	282.434	281.41	279.992		274.713	274.063	272.582	272.55
290.401	288.971	283.602	279.458		275.537	274.975	273.296	271.246
291.291	289.646	286.947	279.293	...	275.802	275.287	274.882	271.582
291.106	289.657	287.574	283.707		275.556	275.47	274.682	273.165
290.998	289.551	287.295	284.176		276.106	275.714	273.903	271.966
290.664	289.202	287.009	283.783		276.523	275.666	273.65	272.001
290.345	288.809	286.681	283.336		276.449	275.371	273.636	272.243
290.107	288.66	286.521	282.775	...	275.964	275.335	273.596	271.933

[:, :, 3652] =

282.926	283.67	283.073	282.293	...	278.147	277.116	276.803	275.597
283.743	284.613	283.219	282.454		278.879	278.547	276.846	274.946
283.814	283.506	282.525	281.945		278.359	276.623	276.224	274.951
284.123	283.35	282.25	281.368		278.279	277.436	275.468	275.422
289.271	288.837	284.726	281.413		278.38	277.147	275.375	274.571
293.833	292.335	289.518	282.214	...	277.842	277.168	276.265	274.742
294.774	293.582	291.862	287.782		278.187	277.193	276.447	274.786
294.876	294.443	292.837	289.089		278.675	277.503	275.921	273.769
295.069	294.504	292.929	289.629		278.731	277.876	275.461	273.733
294.882	293.951	292.589	288.941		279.335	278.16	275.93	273.846
294.284	293.195	291.91	288.445	...	279.817	278.683	276.713	274.464

[:, :, 3653] =

287.912	288.563	288.156	285.831	...	278.462	275.737	275.032	273.965
289.026	288.426	287.168	285.252		279.342	277.259	275.216	273.717
287.971	287.57	286.571	284.772		278.773	276.701	274.801	272.97
287.954	285.956	285.194	284.001		279.628	277.56	273.967	273.266
291.792	290.58	287.419	283.298		279.328	278.125	275.064	273.112

293.837	292.142	289.986	282.892	...	279.816	278.361	276.646	273.587
294.218	293.081	290.791	287.237		279.674	279.13	277.867	274.631
294.547	293.709	290.968	287.526		280.181	280.313	278.919	274.998
295.046	293.815	291.302	287.042		280.62	280.52	279.354	275.451
295.541	293.893	291.453	286.918		281.327	280.75	279.292	276.5
295.487	294.167	291.61	287.925	...	282.428	281.333	279.876	278.086

2.2.6 Subset temperature time

3652-element Vector{DateTime}:

```

2000-01-01T00:00:00
2000-01-02T00:00:00
2000-01-03T00:00:00
2000-01-04T00:00:00
2000-01-05T00:00:00
2000-01-06T00:00:00
2000-01-07T00:00:00
2000-01-08T00:00:00
2000-01-09T00:00:00
2000-01-10T00:00:00
2000-01-11T00:00:00
2000-01-12T00:00:00
2000-01-13T00:00:00
:
2009-12-19T00:00:00
2009-12-20T00:00:00
2009-12-21T00:00:00
2009-12-22T00:00:00
2009-12-23T00:00:00
2009-12-24T00:00:00
2009-12-25T00:00:00
2009-12-26T00:00:00
2009-12-27T00:00:00
2009-12-28T00:00:00
2009-12-29T00:00:00
2009-12-30T00:00:00

```

2.3 Split data into training and testing sets

24×24×731 Array{Union{Missing, Float32}, 3}:

```

[:, :, 1] =
0.230964  0.211317  0.0417734  ...  0.0  0.0  0.0  0.0  0.0  0.0  0.0
0.239267  0.203205  0.0957546  ...  0.0  0.0  0.0  0.0  0.0  0.0  0.0
0.228309  0.229123  0.159002    ...  0.0  0.0  0.0  0.0  0.0  0.0  0.0
0.212892  0.23749   0.212364    ...  0.0  0.0  0.0  0.0  0.0  0.0  0.0
0.179382  0.213751  0.242253    ...  0.0  0.0  0.0  0.0  0.0  0.0  0.0
0.083235  0.157233  0.225679    ...  0.0  0.0  0.0  0.0  0.0  0.0  0.0
0.0       0.0227302 0.12858    ...  0.0  0.0  0.0  0.0  0.0  0.0  0.0
0.0       0.0       0.0       ...  0.0  0.0  0.0  0.0  0.0  0.0  0.0
0.0       0.0       0.0       ...  0.0  0.0  0.0  0.0  0.0  0.0  0.0
0.0       0.0       0.0       ...  0.0  0.0  0.0  0.0  0.0  0.0  0.0

```


0.0	0.356025	0.0		0.0	0.0	0.0	0.0	0.0	0.0
0.0	0.0	0.0		0.0	0.0	0.0	0.0	0.0	0.0
missing	missing	missing	...	0.0	0.0	0.0	0.0	0.0	0.0
missing	missing	missing		0.0	0.0	0.0	0.0	0.0	0.0
missing	missing	missing		0.0	0.0	0.0	0.0	0.0	0.0
missing	missing	missing		0.0	0.0	0.0	0.0	0.0	0.0
missing	missing	missing		0.0	0.0	0.0	0.0	0.0	0.0
missing	missing	missing	...	0.0	0.0	0.0	0.0	0.0	0.0
missing	missing	missing		0.0	0.0	0.0	0.0	0.0	0.0
missing	missing	missing		0.0	0.0	0.0	0.0	0.0	0.0
missing	missing	missing		0.0	0.0	0.0	0.0	0.0	0.0
missing	missing	missing		0.0	0.0	0.0	0.0	0.0	0.0
missing	missing	missing	...	0.0	0.0	0.0	0.0	0.0	0.0
missing	missing	missing		0.0	0.0	0.0	0.0	0.0	0.0
missing	missing	missing		0.0	0.0	0.0	0.0	0.0	0.0
missing	missing	missing		0.0	0.0	0.0	0.0	0.0	0.0

;;; ...

[:, :, 729] =

1.91927	2.18294	4.04477	5.20947	...	0.0	0.0	0.0
0.203245	0.580783	2.77413	4.15098		0.0	0.0	0.0
0.237774	1.28198	1.22886	0.721624		0.0	0.0	0.0
1.80391	2.13242	1.09698	0.152618		0.0	0.0	0.0
2.32314	1.18486	0.259382	0.0		0.0	0.0	0.0
1.09869	0.124012	0.0	0.0	...	0.0	0.0	0.0
0.493284	0.0336855	0.0	0.0		0.0	0.0	0.0
0.0	0.0	0.0	0.0		0.210678	0.0	0.0
0.0	0.0	0.0	0.0		0.244225	0.0478664	0.0
0.0	0.0	0.0	0.0		0.883143	1.56837	0.156057
missing	missing	missing	missing	...	1.22156	1.51986	1.36312
missing	missing	missing	missing		2.79113	3.31666	2.07084
missing	missing	missing	missing		2.48622	2.86405	1.8507
missing	missing	missing	missing		0.999645	7.01725	4.73033
missing	missing	missing	missing		1.48617	2.80321	4.4185
missing	missing	missing	missing	...	0.311555	0.0501361	0.0426737
missing	missing	missing	missing		0.123578	0.0	0.0
missing	missing	missing	missing		0.0	0.0	0.0
missing	missing	missing	missing		0.0	0.0	0.0
missing	missing	missing	missing		0.0	0.0	0.0
missing	missing	missing	missing	...	0.0	0.0	0.0
missing	missing	missing	missing		0.0	0.0	0.0
missing	missing	missing	missing		0.0	0.0	0.0
missing	missing	missing	missing		0.0	0.0	0.0

[:, :, 730] =

1.25381	1.38226	6.78581	9.23103	...	0.209643	0.06951	0.0322176
1.77286	1.17652	5.44695	8.42261		0.674334	0.0	0.038588
5.26336	2.10804	3.33561	5.37528		0.916897	0.0899668	0.0
4.11361	1.87916	1.61745	5.03068		0.337145	0.891777	0.0281657

1.10992	0.986878	1.11829	6.67025		3.08982	0.966748	0.270218
2.73123	2.4278	3.59609	5.59022	...	10.1729	1.6837	0.420492
3.33995	4.18317	3.80484	3.17682		0.469495	1.02329	0.587488
5.67861	5.78922	4.2321	1.34415		0.723784	1.02258	1.35627
5.94113	5.36736	3.45825	5.4516		1.6775	1.66165	2.03233
5.672	4.33358	3.37151	5.49953		2.08825	1.21806	2.91972
missing	missing	missing	missing	...	2.48373	1.6086	1.16387
missing	missing	missing	missing		2.07946	1.88346	0.863437
missing	missing	missing	missing		2.36113	1.62808	0.568243
missing	missing	missing	missing		1.84581	1.57609	1.48535
missing	missing	missing	missing		1.86876	1.23442	0.884178
missing	missing	missing	missing	...	1.79652	0.287498	1.20721
missing	missing	missing	missing		2.98471	0.613985	1.21469
missing	missing	missing	missing		2.01611	0.473238	0.555291
missing	missing	missing	missing		1.43036	0.545582	0.477171
missing	missing	missing	missing		0.386973	0.315955	0.0261992
missing	missing	missing	missing	...	0.0	0.0310814	0.0
missing	missing	missing	missing		0.0	0.0293388	0.0
missing	missing	missing	missing		0.0	0.0	0.0
missing	missing	missing	missing		0.0	0.0	0.0


```
[:, :, 731] =
```

0.0	0.0	0.0	...	0.0	0.194001	0.229313
0.0	0.0	0.0		0.201	0.0	1.05024
0.0	0.0	0.0		0.794761	0.0	0.239922
0.0	0.0	0.0		0.487525	0.0712316	0.0
0.0	0.0	0.0		0.871535	0.725623	0.0546665
0.0	0.0	0.0	...	0.63043	0.834709	0.935677
0.0	0.0	0.0		0.737369	0.396731	1.7525
0.0	0.0	0.0342669		0.810175	0.914279	2.05469
0.0	0.0	0.418512		0.482352	0.6477	0.834548
0.0	0.0364437	0.121333		1.36826	1.77949	1.08005
missing	missing	missing	...	1.58758	4.78649	4.96679
missing	missing	missing		3.17118	5.32441	3.90305
missing	missing	missing		3.59475	4.90733	4.06902
missing	missing	missing		1.31501	5.03513	4.85771
missing	missing	missing		1.21495	2.24849	4.07878
missing	missing	missing	...	1.81574	0.2997	1.09506
missing	missing	missing		1.08286	0.0417713	0.649026
missing	missing	missing		0.0439509	0.0	0.253759
missing	missing	missing		0.0449181	0.0	1.20432
missing	missing	missing		0.264041	0.0	1.47429
missing	missing	missing	...	0.779127	0.235859	0.301635
missing	missing	missing		1.42092	0.159816	0.394416
missing	missing	missing		3.98366	0.940115	1.16965
missing	missing	missing		0.545287	0.645726	1.59136

2.4 Preprocessing

2.4.1 Preprocess function

preprocess (generic function with 1 method)

2.4.2 Preprocess temperature

121×731 Matrix{Float64}:

-5.29162	-9.1721	-15.5382	...	-12.5594	-16.6589	-11.3703
-6.70991	-9.24213	-14.7215		-11.7596	-15.5633	-11.2349
-7.4606	-8.66663	-13.3073		-12.1199	-16.0275	-14.037
-6.92191	-7.86488	-12.4948		-10.7737	-13.6388	-13.1353
-3.20601	-4.50872	-9.2725		-6.62104	-6.23282	-7.36258
-3.4356	-4.63927	-9.81719	...	-6.78153	-6.1819	-3.63953
-3.22345	-4.44058	-9.73227		-6.78883	-6.39981	-2.73266
-2.99846	-4.25502	-9.58033		-6.43223	-6.53416	-2.6556
-2.86815	-4.19236	-9.76109		-6.29229	-6.88516	-2.47985
-2.13038	-3.76691	-9.79706		-6.33689	-7.23386	-2.69678
-1.22141	-3.01323	-9.46011	...	-6.3053	-7.52295	-3.3459
-6.42715	-9.37341	-15.0048		-12.5938	-16.2631	-11.9445
-7.70089	-9.0709	-14.0989		-12.2314	-15.7781	-12.9018
:			⋮			:
-10.2321	-12.4993	-19.5512		-17.3013	-16.7106	-13.594
-13.3914	-17.6619	-18.4287	...	-14.2943	-13.5519	-12.1915
-13.0153	-17.016	-19.3798		-14.0214	-13.0043	-12.993
-12.6853	-16.3206	-19.573		-13.7198	-13.2715	-13.5
-12.877	-16.4353	-20.6092		-17.123	-16.7478	-13.8757
-12.8246	-15.6923	-19.4179		-18.9243	-18.1075	-14.7821
-12.6756	-15.5844	-20.1762	...	-19.3405	-17.9994	-14.8398
-11.9979	-15.4558	-20.235		-16.4364	-15.8031	-14.1816
-11.0614	-15.3627	-20.6582		-16.4999	-15.7922	-13.9889
-10.5836	-14.6008	-20.3137		-15.9558	-15.87	-14.1375
-10.4757	-13.8554	-19.9689		-16.3106	-16.5766	-14.9735
-10.7302	-13.3602	-20.1791	...	-16.9336	-17.4919	-14.9605

2.4.3 Reshape precip

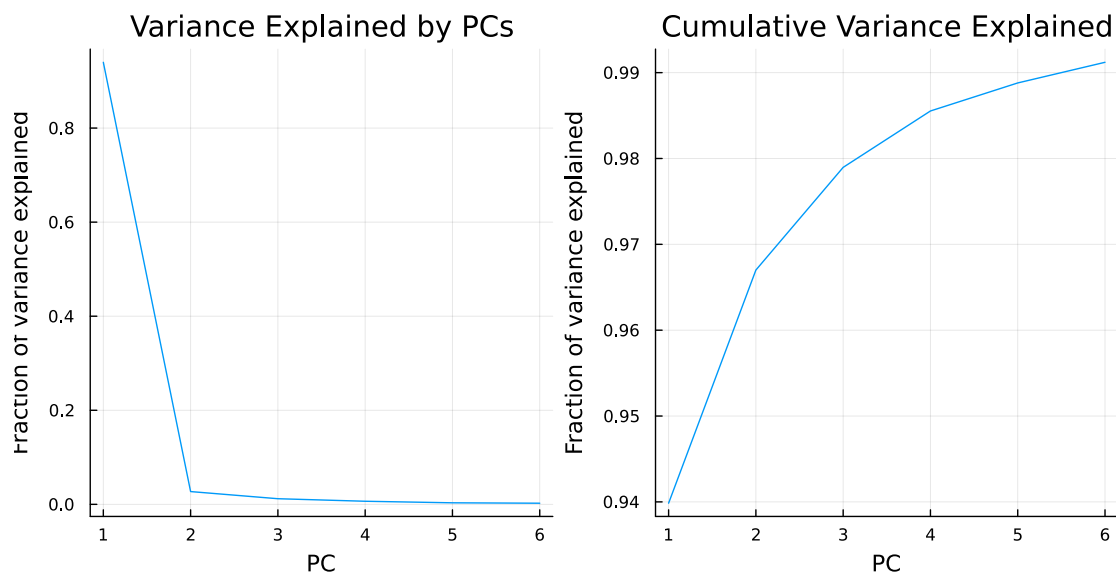
576×731 reshape(::Array{Union{Missing, Float32}, 3}, 576, 731) with eltype Union{Missing, Float32}:

0.230964	0.0	0.0	0.0	...	1.91927	1.25381	0.0
0.239267	0.0	0.0	0.0		0.203245	1.77286	0.0
0.228309	0.0	0.0	0.0		0.237774	5.26336	0.0
0.212892	0.0	0.0	0.0		1.80391	4.11361	0.0
0.179382	0.0	0.0	0.0		2.32314	1.10992	0.0
0.083235	0.0	0.0	0.0	...	1.09869	2.73123	0.0
0.0	0.0	0.0	0.0		0.493284	3.33995	0.0
0.0	0.0	0.0	0.0		0.0	5.67861	0.0
0.0	0.0	0.0	0.0		0.0	5.94113	0.0
0.0	0.0	0.0	0.0		0.0	5.672	0.0
missing	missing	missing	missing	...	missing	missing	missing
missing	missing	missing	missing		missing	missing	missing
missing	missing	missing	missing		missing	missing	missing
:			⋮				:
0.0	0.0	0.0	0.0		1.8507	0.568243	4.06902

0.0	0.0	0.0	0.0	...	4.73033	1.48535	4.85771
0.0	0.0	0.0	0.0		4.4185	0.884178	4.07878
0.0	0.0	0.0	0.153876		0.0426737	1.20721	1.09506
0.0	0.0	0.0	0.0202588		0.0	1.21469	0.649026
0.0	0.0	0.0	0.0		0.0	0.555291	0.253759
0.0	0.0	0.0	0.0	...	0.0	0.477171	1.20432
0.0	0.0	0.0	0.0		0.0	0.0261992	1.47429
0.0	0.0	0.0	0.0		0.0	0.0	0.301635
0.0	0.0	0.0	0.0		0.0	0.0	0.394416
0.0	0.0	0.0	0.0		0.0	0.0	1.16965
0.0	0.0203687	0.0	0.0	...	0.0	0.0	1.59136

3. Principal Component Analysis 3.1 Fit PCA model

3.2 Plot variance to determine number of PCs to keep



3.3 Transform PCs

```
6x731 Matrix{Float64}:
100.247    130.942    179.171    ...    159.314    159.572    128.187
  3.03334     9.58764   -11.2996    ...    -8.46464   -14.8938    -5.77997
  2.22337    -3.58652    0.21244    ...    -1.39438    -8.892     -7.86022
  2.6004     -3.18065   -10.1462    ...    -3.68285    -0.234302   10.0624
  0.0852445   -1.64579    0.984028    ...    -4.64237    -0.923782    0.668111
  4.67906     2.65186   -1.97371    ...     0.396902    0.305045   -4.68379
```

3.4 Save first three PCs

```
2921-element Vector{Float64}:
 3.9052599528110425
 -2.2538113042128542
 -12.537201539433823
 -11.024843758228124
 -0.5726203694196917
```

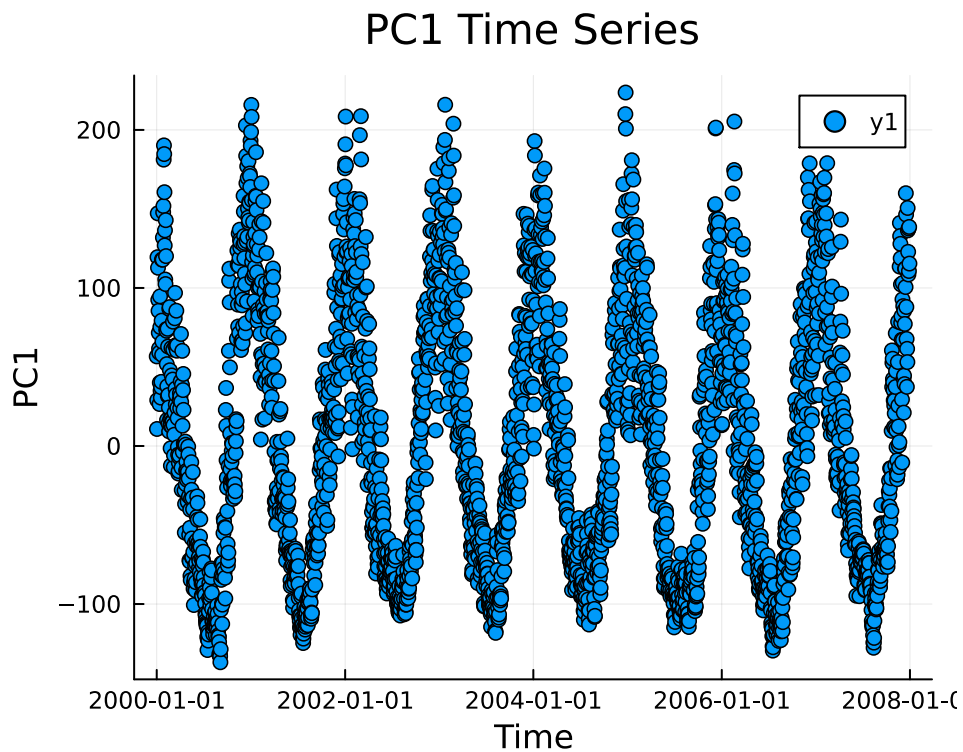
```

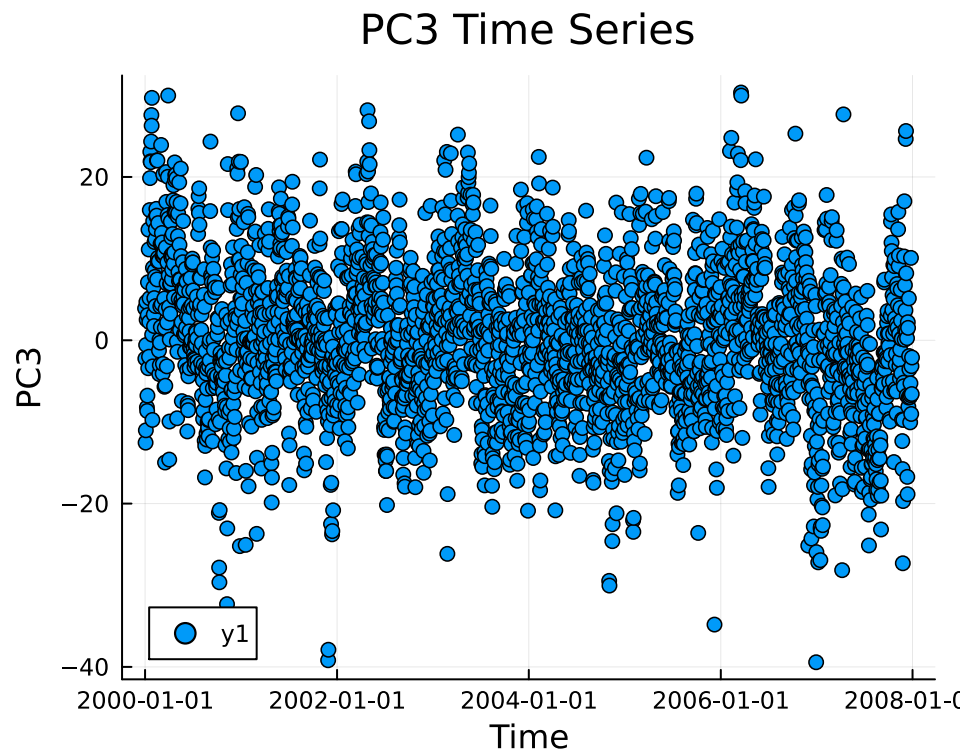
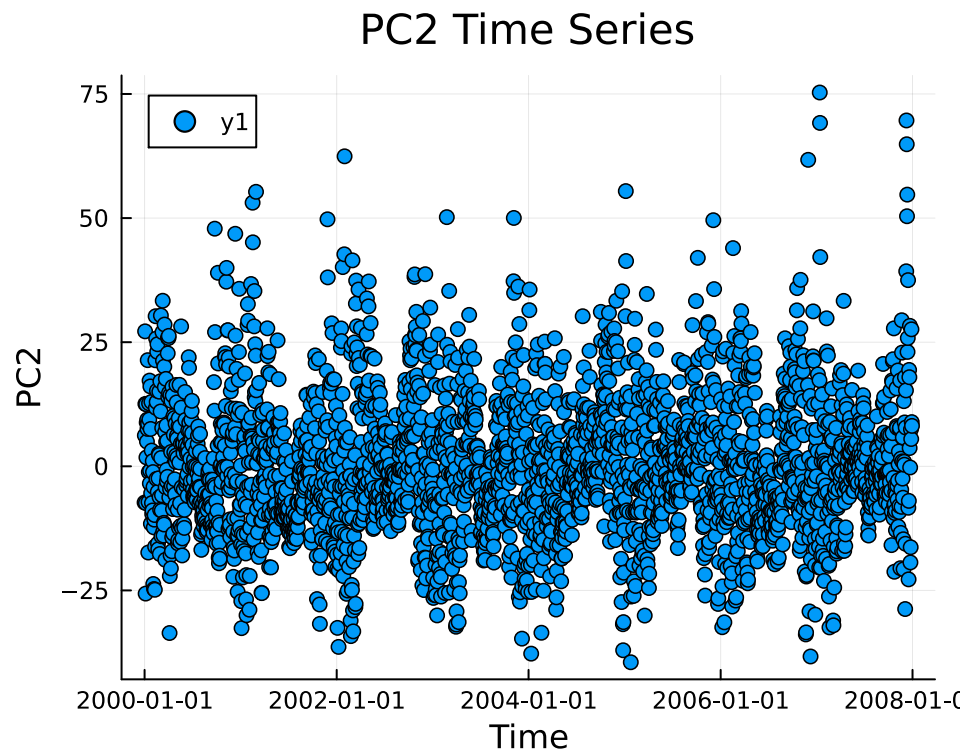
4.701612681874441
2.5183907610932286
-8.215061876405912
-8.658143941931053
-6.778889657459626
3.078553367661374
7.098802390680696
-3.423404610358853
⋮
-1.9310563142374508
-3.550451553216818
-6.775211071460527
-6.952704268141444
-10.052420445554452
-5.265265614927885
-6.733298155961106
10.088840782427953
-9.024376763999939
-3.1651858900584364
-6.582681154712115
-2.110570791578569

```

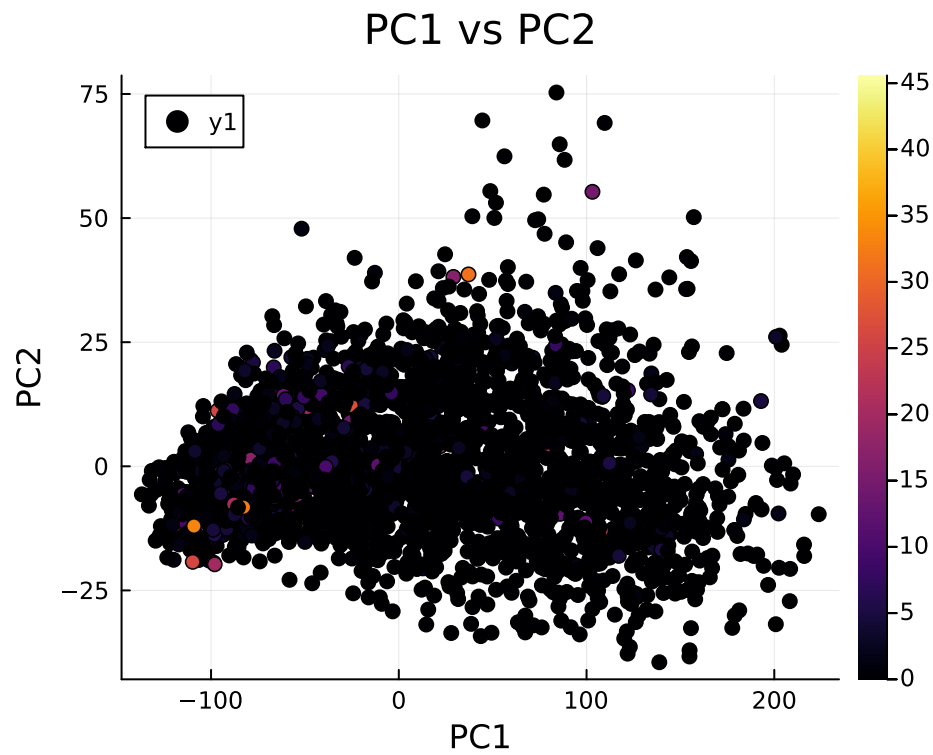
3.5 Plot PCA

3.5.1 Plot time series of first three PCs

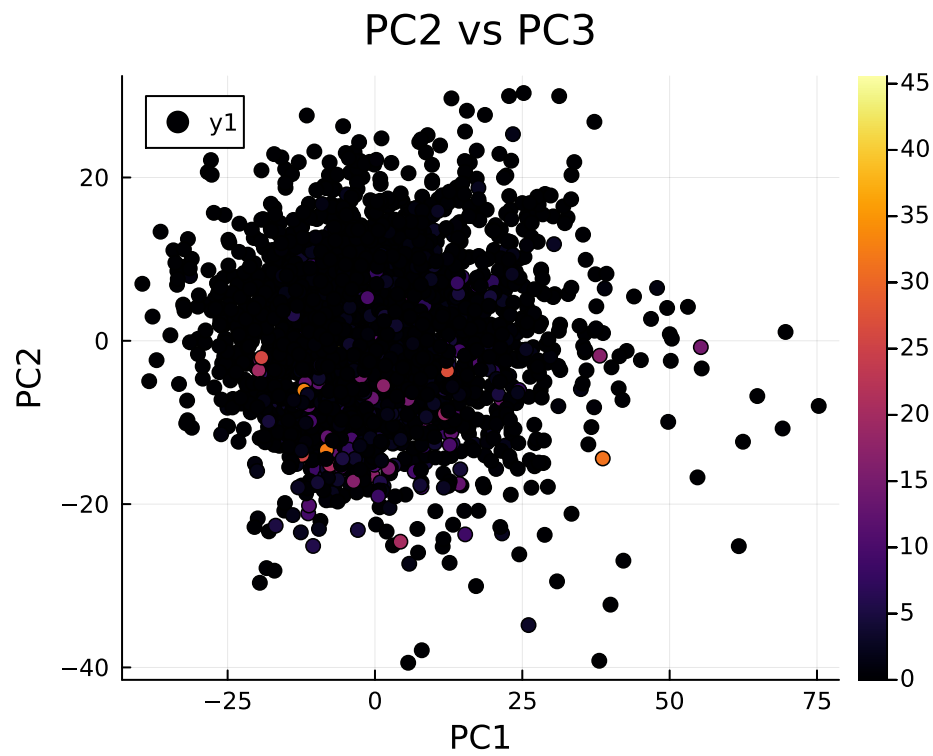




3.5.2 Plot first 2 PCs and mean precipitation



3.5.3 Plot the second and third PCs with mean precipitation



4. Approach 1: KNN

Given the high dimensional structure of the gridded temperature data, a principal components analysis was performed to downscale the features while preserving variation to allow for modeling across space and time. PCA effectively reduces the number of columns (in this case, locations) by projecting the data onto new PC axes. The number of principal components to retain was determined by plotting the variance explained by the principal components and the cumulative variance. After analyzing these figures, a break in variance is observed after principal component number two. The cumulative variance explained plot reveals that approximately 97.8% of the variance is explained by the first three principal components, so three principal components were retained for further analysis. The PCA model was fit to the training temperature data and then the temperature training and testing data were transformed onto the PCA space.

KNN was then used to predict the precipitation data for the testing temperature data given temperature and precipitation over the training period. The KNN function calculates the Euclidean distance between the new datapoint and the existing K data points and assigns weights based on these distances. Weighted sampling is then conducted to obtain the predicted value based on the indexed position sampled. Three was chosen as the hyperparameter for the number of neighbors because this number of neighbors is standard in KNN analysis and serves as an appropriate baseline. The resultant predicted precipitation values were compared to the actual precipitation training data to evaluate the fit of the model using residuals, MAE, and MSE.

4.1 Define KNN function

```
knn (generic function with 1 method)
```

4.2 Combining PCA and KNN

```
predict_knn (generic function with 1 method)
```

4.3 Test the model

```
24x24x731 Array{AbstractFloat, 3}:
```

```
[:, :, 1] =
```

```
0.0 0.0 0.0 0.0 0.0 0.0 0.0 ... 0.0 0.0 0.0
0.0 0.0 0.0 0.0 0.0 0.0 0.0 ... 0.0 0.0 0.0
0.0 0.0 0.0 0.0 0.0 0.0 0.0 ... 0.0 0.0 0.0
0.0 0.0 0.0 0.0 0.0 0.0 0.0 ... 0.0 0.0 0.0
0.0 0.0 0.0 0.0 0.0 0.0 0.0 ... 0.0 0.0 0.0
0.0 0.0 0.0 0.0 0.0 0.0 0.0 ... 0.0 0.0 0.0
0.0 0.0 0.0 0.0 0.0 0.0 0.0 ... 0.0 0.0 0.0
0.0 0.0 0.0 0.0 0.0 0.0 0.0649847 ... 0.0 0.0 0.0
0.0 0.0 0.0 0.0 0.0 0.0 0.0218424 ... 0.0 0.0 0.0
0.0 0.0 0.0 0.0 0.0 0.0 0.0 ... 0.0 0.0 0.0
0.0 0.0 0.0 0.0 0.0 0.0 0.0 ... 0.0 0.0 0.0
0.0 0.0 0.0 0.0 0.0 0.0 0.0 ... 0.0533735 0.0 0.0
0.0 0.0 0.0 0.0 0.0 0.0 0.0 ... 0.0 0.0209779 0.0901863
0.0 0.0 0.0 0.0 0.0 0.0 0.0 ... 0.0302493 0.0 0.0415976
0.0 0.0 0.0 0.0 0.0 0.0 0.0 ... 0.19425 0.0473435 0.0
0.0 0.0 0.0 0.0 0.0 0.0 0.0 ... 0.0884871 0.0421871 0.0
```

0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0780542	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0	0.0	...	0.0211939	0.0
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.176734
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.152089

[:, :, 2] =

1.75447	1.41588	1.19977	0.848477	...	0.0	0.0	0.0
3.56247	1.01861	1.52315	1.49084		0.0	0.0	0.0
9.91041	1.38696	1.36932	2.96705		0.0	0.0	0.0
8.48843	2.42787	1.51287	2.51147		0.0	0.0	0.0
1.6389	1.39423	1.54824	0.957531		0.0	0.0	0.0
4.94228	2.9396	0.673247	0.0492278	...	0.0	0.0	0.0
6.01258	5.55706	4.37366	0.18723		0.0	0.0	0.0
7.2367	11.1376	5.32148	0.156859		0.0	0.0	0.0
6.91768	7.26536	1.85251	0.0910493		0.0	0.0	0.0
4.00505	2.66892	1.61451	0.683203		0.0	0.0	0.0
0.0	0.0	0.0	0.0	...	0.0	0.0	0.0
0.0	0.0	0.0	0.0		0.0	0.0	0.0
0.0	0.0	0.0	0.0		0.0	0.0	0.0
0.0	0.0	0.0	0.0		0.0	0.0	0.0
0.0	0.0	0.0	0.0		0.0	0.0	0.0
0.0	0.0	0.0	0.0	...	0.0334075	0.0	0.0
0.0	0.0	0.0	0.0		0.0	0.0	0.0
0.0	0.0	0.0	0.0		0.0	0.0	0.0
0.0	0.0	0.0	0.0		0.0	0.0	0.0
0.0	0.0	0.0	0.0		0.0	0.0	0.0
0.0	0.0	0.0	0.0	...	0.0	0.0	0.0
0.0	0.0	0.0	0.0		0.0	0.0	0.0777392
0.0	0.0	0.0	0.0		0.0723409	0.0859267	0.147741
0.0	0.0	0.0	0.0		0.0262213	0.0489934	0.195904

[:, :, 3] =

1.49692	1.99355	2.05643	...	0.0	0.0	0.0	0.0	0.0
0.833292	0.72057	1.09722		0.0	0.0	0.0	0.0	0.0
3.12063	1.12997	0.357355		0.0	0.0	0.0	0.0	0.0
2.17121	1.34081	0.0870448		0.0	0.0	0.0	0.0	0.0
0.146885	0.0664999	0.178861		0.0	0.0	0.0	0.0	0.0
0.464428	0.344984	0.791974	...	0.0	0.0	0.0	0.0	0.0
1.5742	2.66726	1.56431		0.0	0.0	0.0	0.0	0.0
1.55001	4.35322	2.61319		0.0	0.0	0.0	0.0	0.0
0.500756	2.29461	2.9136		0.0	0.0	0.0	0.0	0.0
0.479857	1.89894	2.37473		0.0	0.0	0.0	0.0	0.0
0.0	0.0	0.0	...	0.0	0.0	0.0	0.0	0.0
0.0	0.0	0.0		0.0	0.0	0.0	0.0	0.0
0.0	0.0	0.0		0.0	0.0	0.0	0.0	0.0
0.0	0.0	0.0		0.035775	0.0	0.0	0.0	0.0

0.0	0.0	0.0		0.0511538	0.0	0.0	0.0	0.0	0.0
0.0	0.0	0.0	...	0.0340634	0.0	0.0	0.0	0.0	0.0
0.0	0.0	0.0		0.0237441	0.0	0.0	0.0	0.0	0.0
0.0	0.0	0.0		0.0	0.0	0.0	0.0	0.0	0.0
0.0	0.0	0.0		0.0	0.0	0.0	0.0	0.0	0.0
0.0	0.0	0.0		0.0	0.0	0.0	0.0	0.0	0.0
0.0	0.0	0.0	...	0.0	0.0	0.0	0.0	0.0	0.0
0.0	0.0	0.0		0.0	0.0	0.0	0.0	0.0	0.0
0.0	0.0	0.0		0.0	0.0	0.0	0.0	0.0	0.0
0.0	0.0	0.0		0.0	0.0	0.0	0.0	0.0	0.0

```
;;; ...
```

```
[:, :, 729] =
```

0.0	0.0	0.0		0.0	0.0	0.0	...	0.140855	0.0414727	0.436889
0.0	0.0	0.0		0.0	0.0	0.0		0.598811	0.0493965	0.314328
0.0	0.0	0.0		0.0	0.0	0.0		0.240909	0.100539	0.407857
0.0	0.0	0.0		0.0	0.0	0.0		0.0943053	0.266252	0.592172
0.0	0.0	0.0		0.0	0.0	0.0		0.082403	0.117491	0.129018
0.0	0.0	0.0		0.0	0.0	0.0	...	0.292574	0.660883	0.528695
0.0	0.0	0.0		0.0	0.0	0.0		0.236851	0.314418	0.267703
0.0	0.0	0.0		0.0	0.0	0.0		0.430934	0.222956	0.587176
0.0	0.0	0.0513685		0.0	0.0	0.0		0.180304	0.129312	0.219643
0.0	0.0	0.0623262		0.0	0.0	0.0		0.345806	0.176304	0.689809
0.0	0.0	0.0		0.0	0.0	0.0	...	0.470213	0.479811	0.99668
0.0	0.0	0.0		0.0	0.0	0.0		0.903811	0.805968	0.633223
0.0	0.0	0.0		0.0	0.0	0.0		1.55173	1.94446	1.05149
0.0	0.0	0.0		0.0	0.0	0.0		0.751358	2.42407	0.892646
0.0	0.0	0.0		0.0	0.0	0.0		0.84299	1.61934	1.53261
0.0	0.0	0.0		0.0	0.0	0.0	...	0.411869	1.02282	0.344983
0.0	0.0	0.0		0.0	0.0	0.0		0.954351	1.51298	0.313492
0.0	0.0	0.0		0.0	0.0	0.0		2.95927	3.57947	1.06551
0.0	0.0	0.0		0.0	0.0	0.0		1.85423	1.76604	1.70267
0.0	0.0	0.0		0.0	0.0	0.0		0.606824	0.320045	1.13114
0.0	0.0	0.0		0.0	0.0	0.0	...	1.41541	0.305823	0.435956
0.0	0.0	0.0		0.0	0.0	0.0		0.600175	0.637315	0.814799
0.0	0.0	0.0		0.0	0.0	0.0		0.798034	0.480626	0.410468
0.0	0.0	0.0		0.0	0.0	0.0		1.01349	1.12795	0.425947

```
[:, :, 730] =
```

0.0	0.0	0.0250906	...	0.0	0.0	0.0	0.0
0.0	0.0	0.05093		0.0	0.0	0.0	0.0
0.0	0.0434867	0.163335		0.0	0.0	0.0	0.0
0.084721	0.455731	0.452126		0.0	0.0	0.0	0.0
0.376434	0.272408	0.550661		0.0	0.0	0.0	0.0
1.80568	0.370495	1.58513	...	0.0	0.0	0.0	0.0
4.14398	2.07312	4.16797		0.0	0.0	0.0	0.0
10.2574	9.56134	5.12458		0.0	0.0	0.0	0.0
9.65954	6.17192	4.32784		0.0	0.0	0.0	0.0
7.42415	5.88919	6.10839		0.0	0.0	0.0	0.0

0.0	0.0	0.0	...	0.0	0.0	0.0	0.0
0.0	0.0	0.0		0.0	0.0	0.0	0.0
0.0	0.0	0.0		0.0	0.0	0.0	0.0
0.0	0.0	0.0		0.0	0.301538	0.0756907	0.0
0.0	0.0	0.0		0.0	0.87682	0.408131	0.0
0.0	0.0	0.0	...	0.0	0.025333	0.36935	0.0624869
0.0	0.0	0.0		0.0	0.0	0.0766167	0.0
0.0	0.0	0.0		0.0	0.0	0.032296	0.100898
0.0	0.0	0.0		0.0	0.0	0.0	0.0469467
0.0	0.0	0.0		0.0	0.0	0.0	0.0
0.0	0.0	0.0	...	0.0	0.0	0.0	0.0
0.0	0.0	0.0		0.0	0.0	0.0	0.0
0.0	0.0	0.0		0.0	0.0	0.0	0.0532102
0.0	0.0	0.0		0.0	0.0	0.0	0.0


```
[:, :, 731] =
```

0.0	0.0	0.0	0.0	0.0	0.0	...	0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0		0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0		0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0		0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0		0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0	...	0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0		0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0		0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0		0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0		0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0251531	...	0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0		0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0		0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0		0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0		0.0208791	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0	...	3.16174	0.382311	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0		6.56101	1.23528	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0		0.167616	0.41249	0.159337	0.0
0.0	0.0	0.0	0.0	0.0	0.0		0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0		0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0	...	0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0		0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0		0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0		0.0	0.0	0.0	0.0

4.4 Analyze Fit

4.4.1 Reshape data for plotting

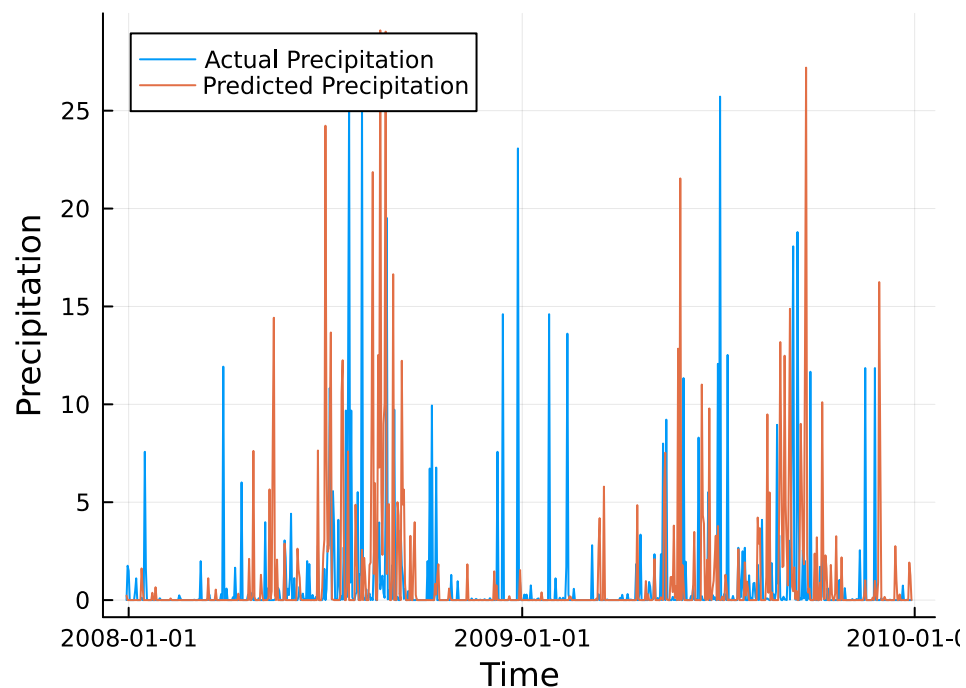
```
576x731 Matrix{AbstractFloat}:
```

0.0	1.75447	1.49692	0.0	...	0.0	0.0	0.0
0.0	3.56247	0.833292	0.0		0.0	0.0	0.0
0.0	9.91041	3.12063	0.0		0.0	0.0	0.0
0.0	8.48843	2.17121	0.0		0.0	0.084721	0.0

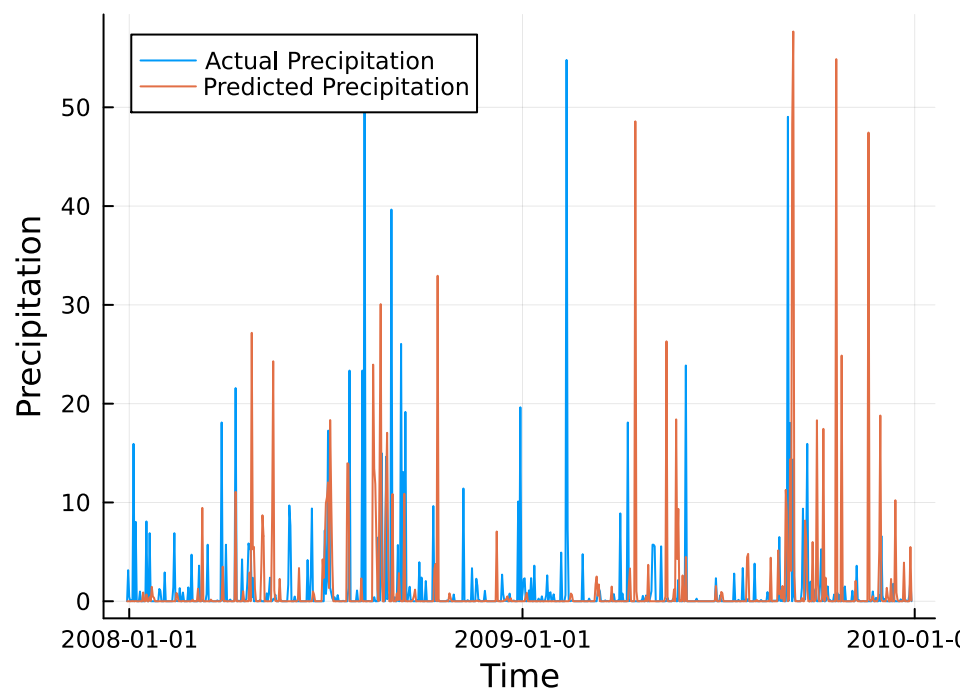
0.0	1.6389	0.146885	0.0	0.0	0.376434	0.0	
0.0	4.94228	0.464428	0.0	...	0.0	1.80568	0.0
0.0	6.01258	1.5742	0.0	0.0	4.14398	0.0	
0.0	7.2367	1.55001	0.0	0.0	10.2574	0.0	
0.0	6.91768	0.500756	0.0	0.0	9.65954	0.0	
0.0	4.00505	0.479857	0.0	0.0	7.42415	0.0	
0.0	0.0	0.0	0.0	...	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
⋮				⋮			⋮
0.0901863	0.0	0.0	1.0249	1.05149	0.0	0.0	
0.0415976	0.0	0.0	0.360796	...	0.892646	0.0	0.0
0.0	0.0	0.0	0.414263	1.53261	0.0	0.0	
0.0	0.0	0.0	0.108951	0.344983	0.0624869	0.0	
0.0	0.0	0.0	0.212448	0.313492	0.0	0.0	
0.0	0.0	0.0	0.0	1.06551	0.100898	0.0	
0.0	0.0	0.0	0.0	...	1.70267	0.0469467	0.0
0.0	0.0	0.0	0.0796868	1.13114	0.0	0.0	
0.0	0.0	0.0	0.118671	0.435956	0.0	0.0	
0.0	0.0777392	0.0	0.0	0.814799	0.0	0.0	
0.176734	0.147741	0.0	0.0	0.410468	0.0532102	0.0	
0.152089	0.195904	0.0	0.0	...	0.425947	0.0	0.0

4.4.2 Time series actual vs predicted test precipitation

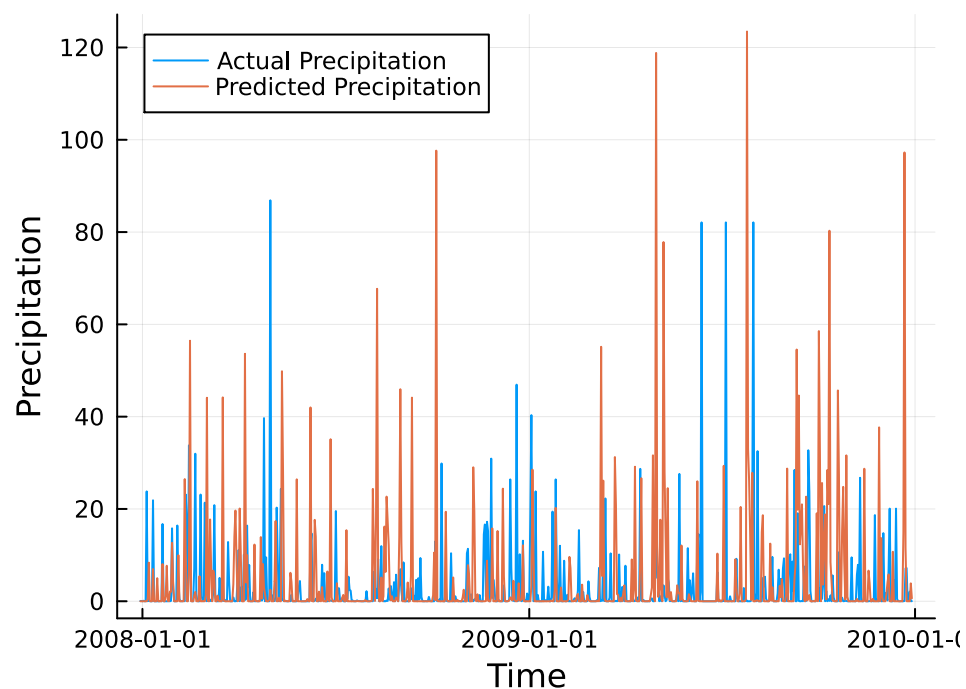
Precipitation at t+1 Predicted by KNN- GC 1



Precipitation at $t+1$ Predicted by KNN- GC 15

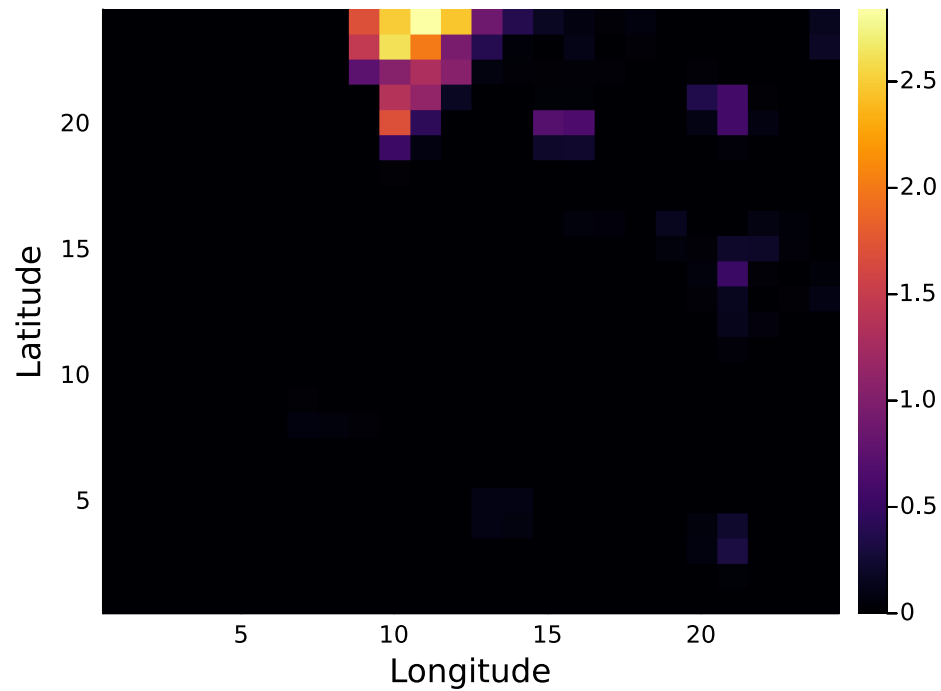


Precipitation at $t+1$ Predicted by KNN- GC 40

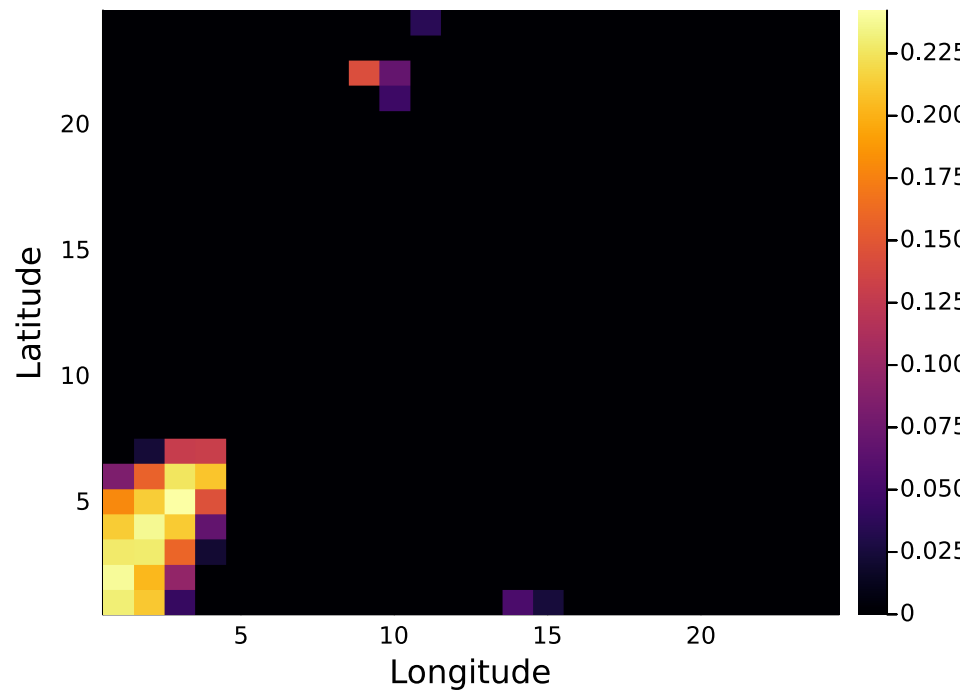


4.4.3 Heatmaps

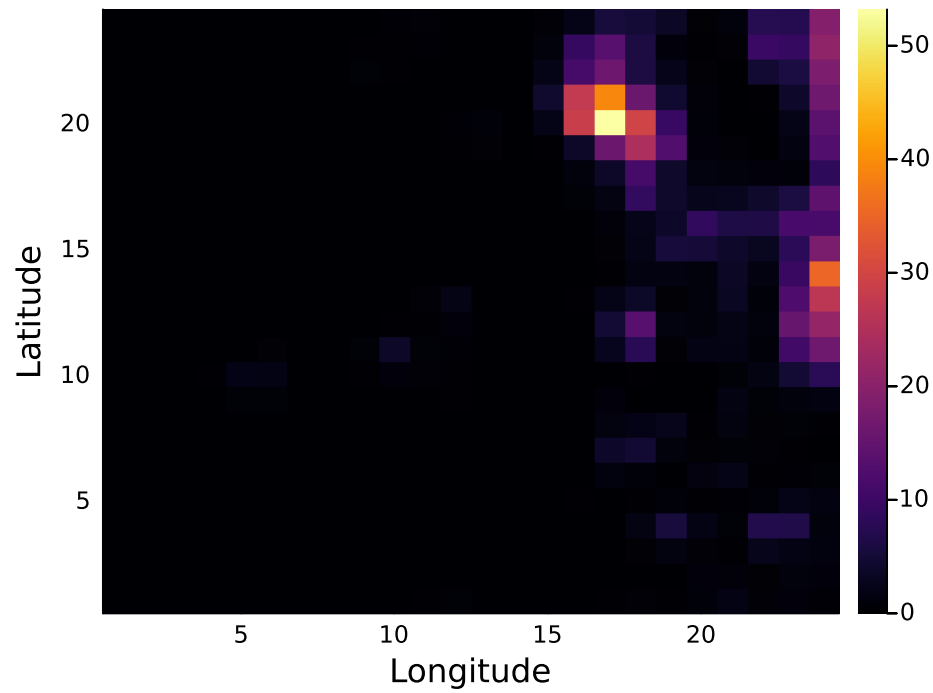
Predicted Precipitation at t+1 Predicted by KNN



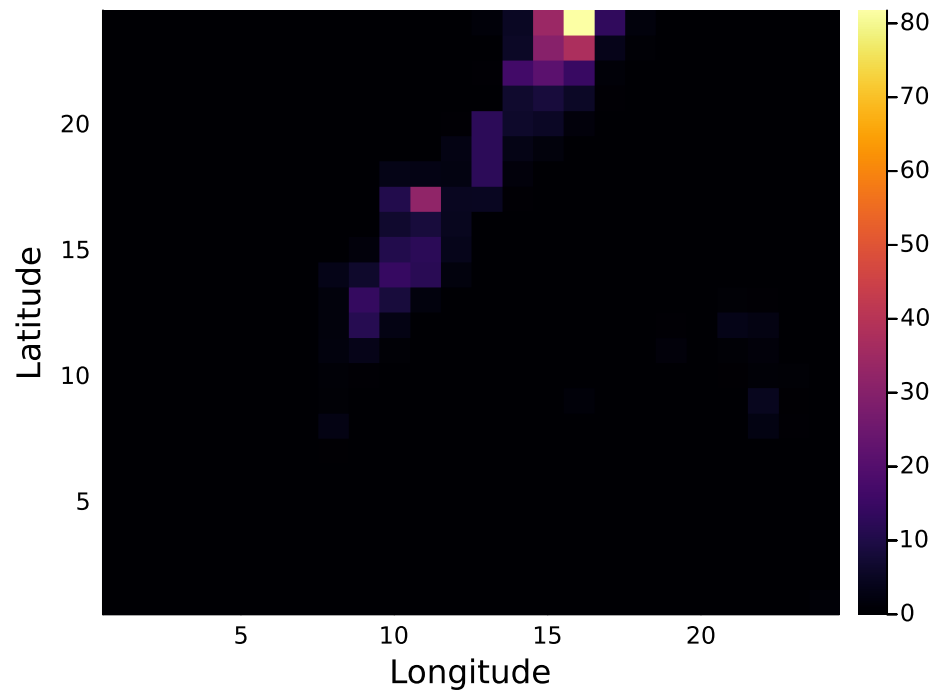
Actual Precipitation at t+1 Predicted by KNN



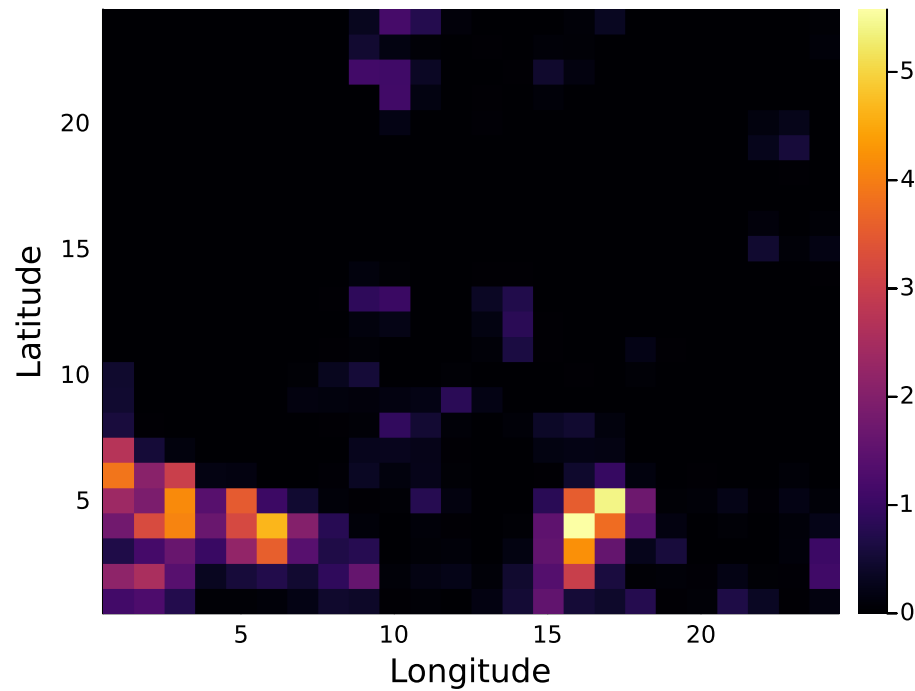
Predicted Precipitation at t+1 Predicted by KNN



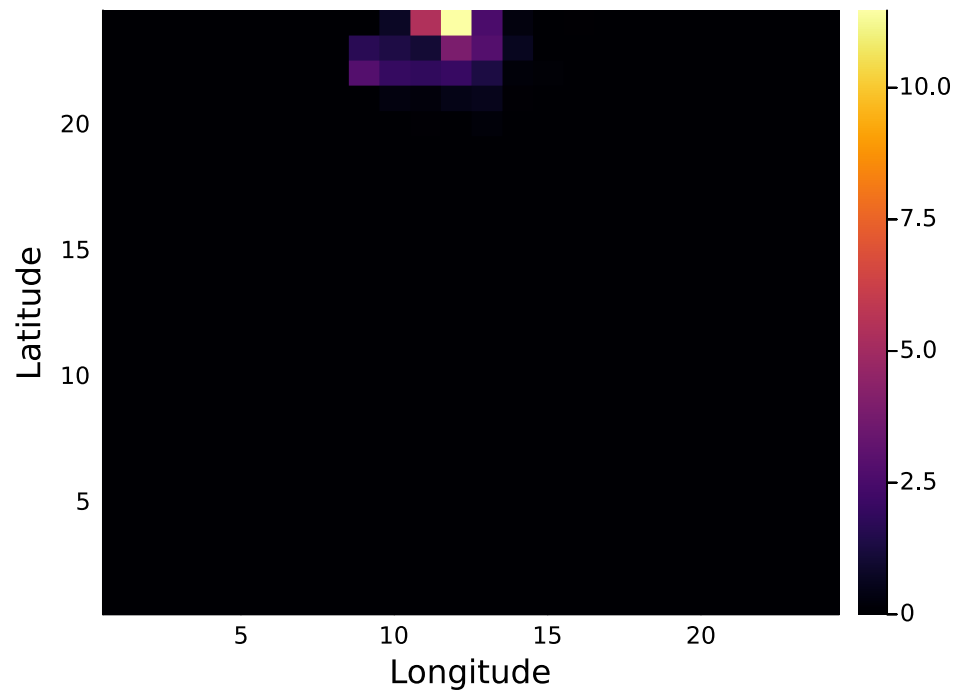
Actual Precipitation at t+1 Predicted by KNN



Predicted Precipitation at t+1 Predicted by KNN



Actual Precipitation at t+1 Predicted by KNN



4.4.4 Evaluate fit using MSE

```
mean_se (generic function with 1 method)
```

4.4.5 Evaluate fit using MAE

```
mean_abs_error (generic function with 1 method)
```

4.4.6 Evaluate fit using residuals

```
residuals (generic function with 1 method)
```

5. Approach 2: PCA/Linear Regression

Similarly to Approach 1, Approach 2 employed PCA to reduce the dimensions of the temperature data. Only the first principal component was retained for further linear regression to allow for the creation of simple vectors that are compatible with linear regression syntax. Linear regression was performed on precipitation at each grid cell location following the first principal component of the temperature data. The temperature test data was then applied to this linear model to predict precipitation and was similarly compared to actual observed precipitation to evaluate the model using residuals, MAE, and MSE.

5.1 Linear Regression

```
576×2921 Matrix{AbstractFloat}:
```

```
0.0      0.0      0.0      0.0      ...  0.0      0.0      0.0  0.0
0.0      0.0      0.0      0.0      ...  0.0      0.0      0.0  0.0
0.0      0.0      0.0      0.0      ...  0.0      0.0      0.0  0.0
0.0      0.0      0.0      0.0      ...  0.0      0.0      0.0  0.0
0.0      0.0      0.0      0.0      ...  0.0      0.0      0.0  0.0
0.0      0.0      0.0      0.0      ...  0.0      0.0      0.0  0.0
0.0      0.0      0.0      0.0      ...  0.0      0.0      0.0  0.0
0.0      0.0      0.0      0.0      ...  0.0      0.0      0.0  0.0
0.0      0.0      0.0      0.0      ...  0.0      0.0      0.0  0.0
0.0      0.0      0.0      0.0      ...  0.0      0.0      0.0  0.0
0.0      0.0      0.0      0.0      ...  0.0      0.0      0.0  0.0
0.0      0.0      0.0      0.0      ...  0.0      0.0      0.0  0.0
0.0      0.0      0.0      0.0      ...  0.0      0.0      0.0  0.0
0.0      0.0      0.0      0.0      ...  0.0      0.0      0.0  0.0
```

```
⋮
0.0      4.3323    2.59903  0.0857328  ...  4.40795  0.0384247  0.0  0.0
0.0      10.0802   1.95597  0.103022   ...  2.42879  0.0         0.0  0.0
0.0207805 13.5422    1.23499  0.0         ...  2.51666  0.0         0.0  0.0
0.0338578 15.6146    1.12839  0.0         ...  2.64994  0.0472129  0.0  0.0
0.0513665 22.9741    3.07755  0.0         ...  2.00045  0.0460711  0.0  0.0
0.156683  28.7882    0.937954 0.0         ...  1.97276  0.0751787  0.0  0.0
0.0214627 24.5314    2.03677  0.0         ...  2.50186  0.0         0.0  0.0
3.03718    14.6306    4.33941  0.0         ...  2.24995  0.0         0.0  0.0
3.51391    8.24005   8.96606  0.0         ...  2.30591  0.0429012  0.0  0.0
0.607323   26.5516   10.5232  0.0         ...  4.07364  0.0         0.0  0.0
0.0        51.9611   17.5984  0.206645   ...  2.48726  0.659939   0.0  0.0
0.602726   73.7095   25.2627  1.76726    ...  3.56594  2.77694    0.0  0.0
```

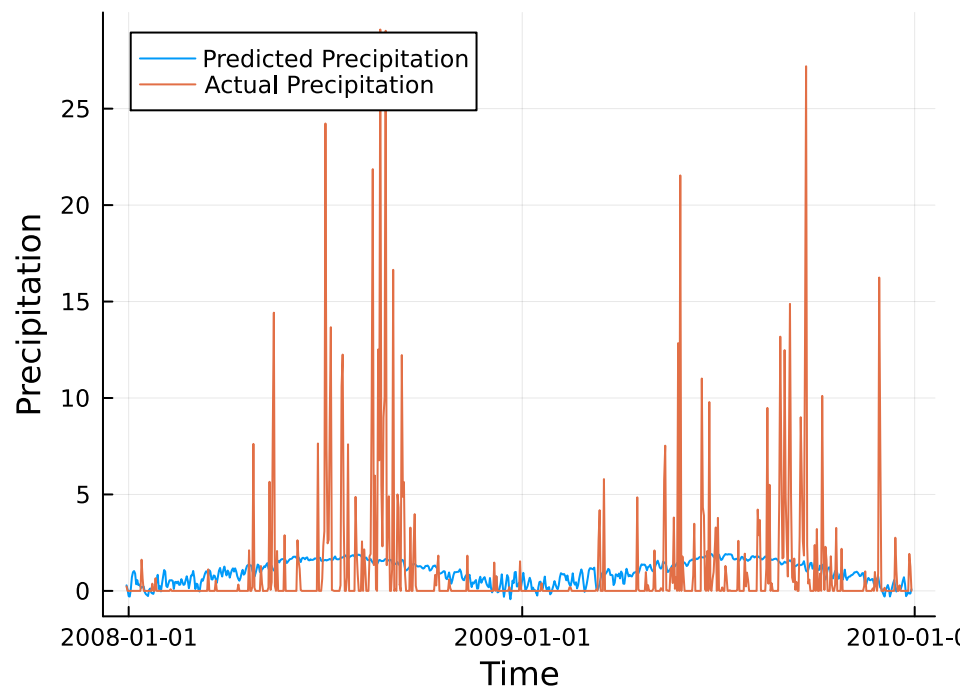
```
731×576 Matrix{Float64}:
```

```
0.288686  0.326512  0.585564  ...  2.21833  2.42359  2.47836  2.48521
```

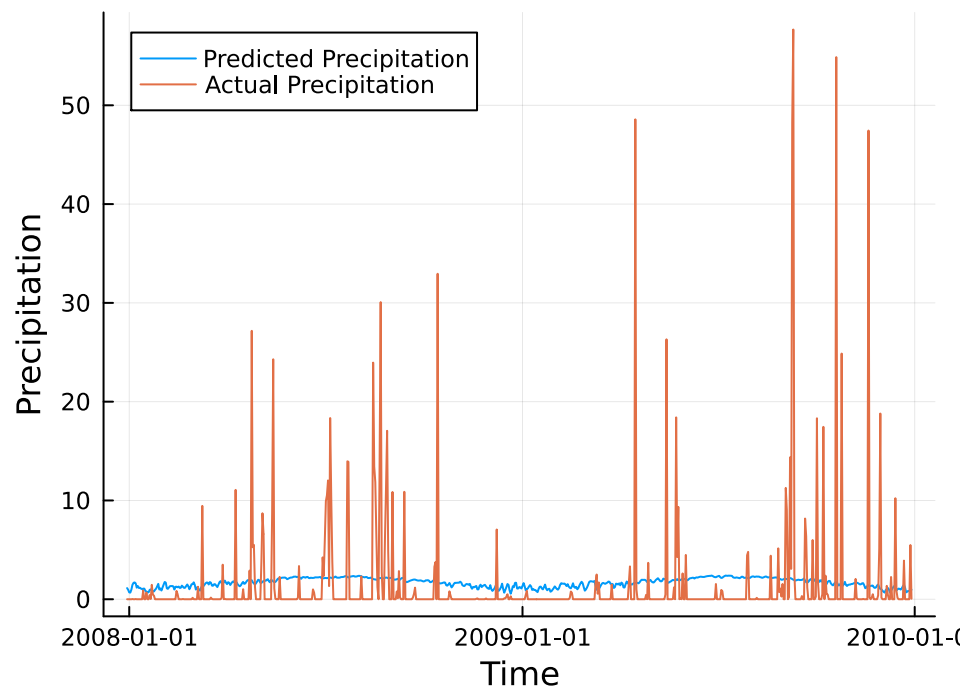
0.0693047	0.0866469	0.344425		1.94787	2.19186	2.22505	2.27144
-0.275387	-0.29023	-0.0344529		1.52294	1.82776	1.82705	1.93557
-0.293983	-0.310562	-0.0548933		1.50001	1.80811	1.80557	1.91745
0.0960588	0.115899	0.373833		1.98085	2.22012	2.25594	2.29751
0.714566	0.792159	1.05368	...	2.74335	2.87345	2.9701	2.9002
0.967155	1.06833	1.33132		3.05475	3.14026	3.26176	3.14632
1.02983	1.13686	1.40022		3.13202	3.20647	3.33413	3.2074
0.901364	0.996399	1.25901		2.97364	3.07077	3.18579	3.08221
0.343025	0.385925	0.645293		2.28532	2.48099	2.5411	2.53816
0.558737	0.62178	0.8824	...	2.55125	2.70885	2.79017	2.74835
0.2988	0.337571	0.596682		2.2308	2.43427	2.49004	2.49506
0.332966	0.374927	0.634237		2.27292	2.47036	2.52949	2.52836
:			⋮				:
0.176906	0.204296	0.462699		2.08052	2.30552	2.34929	2.37629
0.0974754	0.117448	0.37539	...	1.9826	2.22161	2.25757	2.29889
0.269593	0.305637	0.564579		2.19479	2.40342	2.45631	2.46661
0.526515	0.586549	0.846982		2.51152	2.67481	2.75297	2.71696
0.707819	0.784782	1.04627		2.73504	2.86632	2.96231	2.89362
0.356321	0.400463	0.659908		2.30171	2.49503	2.55645	2.55111
-0.272868	-0.287476	-0.0316845	...	1.52604	1.83042	1.82995	1.93802
-0.178059	-0.183814	0.0725277		1.64292	1.93056	1.93943	2.0304
-0.066708	-0.0620658	0.194923		1.7802	2.04818	2.068	2.13891
-0.13347	-0.135062	0.121539		1.69789	1.97766	1.99091	2.07385
-0.135314	-0.137078	0.119512		1.69562	1.97572	1.98878	2.07206
0.0889984	0.10818	0.366072	...	1.97215	2.21266	2.24779	2.29063

5.2 Analyze Fit 5.2.1 Plot time series of actual vs predicted precipitation at different grid cells

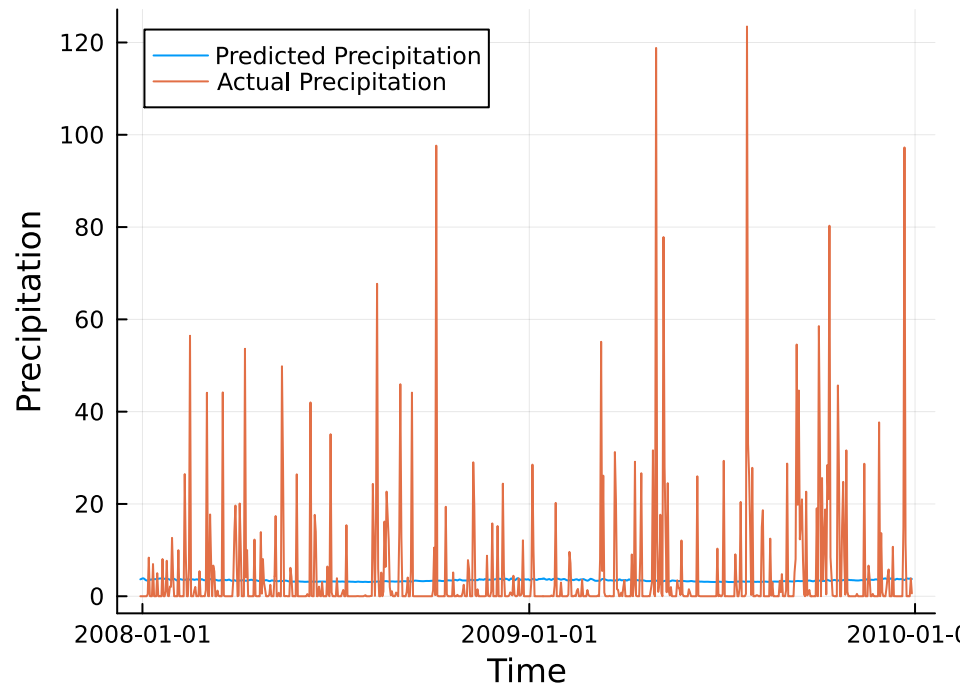
Precipitation at $t+1$ Predicted by Lin Reg- GC



Precipitation at $t+1$ Predicted by Lin Reg- GC :



Precipitation at $t+1$ Predicted by Linear Regression



5.2.2 Evaluate Fit using MSE

5.2.3 Evaluate fit using MAE

5.2.4 Evaluate fit using residuals

6. Compare

When plotting the predicted precipitation and the actual precipitation for the test data, upon visual inspection it seems that the PCA-KNN model more closely resembles the trends of the actual precipitation. The PCA-KNN predicted precipitation of varying levels while the PCA-linear regression model predicted nearly constant, low-level precipitation year-round, with some reflection of the seasonal trends as the maximum periods of the curves occur over the same range of time. However, upon closer analysis, the PCA-KNN model predictions differed significantly from the actual precipitation values. When considering only the residuals of actual precipitation minus predicted precipitation, the PCA-KNN model appears to perform better because the average of the residuals is lower than the PCA-linear regression model. However, when considering the mean absolute error and the mean squared error, the PCA-linear regression model outperformed the PCA-KNN model with significantly lower MAE and MSE, indicating overall that predictions are closer to actual values than the PCA-KNN model. Residuals might be lower than MSE or MAE for a model if there are both undershoots and overshoots when considering the entire dataset of predictions. For example, if the PCA-KNN model predicted 0 mm on a day that should be 40 mm and 40 mm on a day that should be 0 mm, the averaged residuals would be zero since they cancel out. MSE and MAE, however, are a more robust measure of model performance because they account for differences in sign via squaring and absolute value respectively.

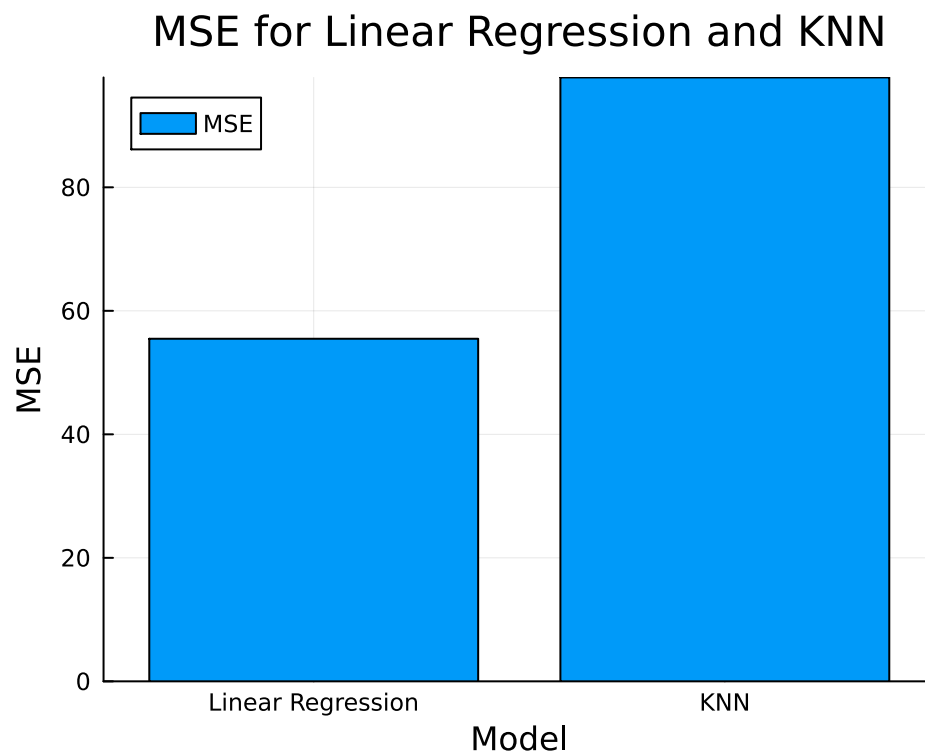
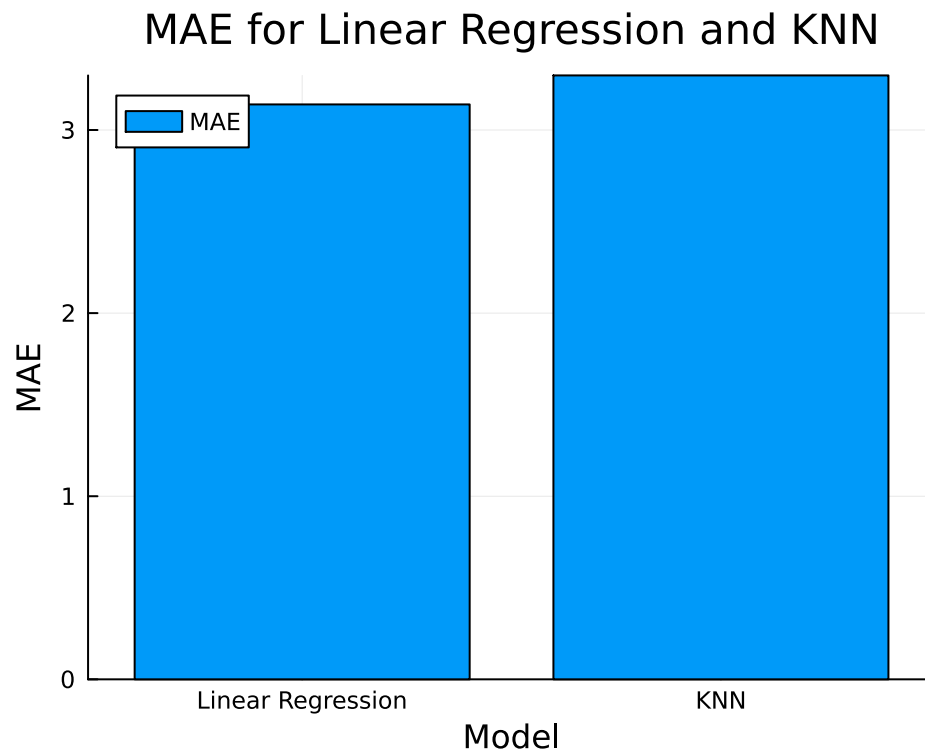
MSE places a greater emphasis on large errors but is sensitive to outliers, which explains why the difference in MSE is greater between the two models than MAE.

The PCA-KNN model appears to be capturing the general shape of the data better, but local predictions are not very reliable based on the MAE and MSE calculations. This model could be further optimized by optimizing the hyperparameters n_{pca} and K . N_{pca} is the number of principal components retained in the model. Three was selected as the n_{pca} value in this analysis based on fraction of variance explained, but a systematic approach could be taken to test all values of PCs and retain the number with the lowest MSE/MAE. A similar systematic approach could be used to optimize the K parameter as well.

The PCA-linear regression model is demonstrating a “dreary” effect by predicting low level precipitation across the time series. While there are periods of slight increase and decrease over the year corresponding to seasonal variation, this model generally does not capture the shape of the data or minimum/maximum values well. Only the first principal component was retained for simplicity of the model, but to improve the predictions more principal components should be considered. Additional principal components may more accurately capture heavy precipitation patterns and predict values closer to actual precipitation on heavy rainfall days.

A limitation of both models is the handling of missing precipitation values. This analysis elected to replace missing values with 0.0, but this is likely altering predictive accuracy by introducing bias. An improvement would be to remove missing values altogether before conducting data analysis. Another major limitation of both models is only using temperature data to predict precipitation. Based on prior knowledge, pressure data is also important in predicting precipitation and model performance likely would have significantly improved if more variables beyond temperature were included.

-0.23472928f0



7. Conclusion

This report provides preliminary models to predict precipitation at time $t+1$ based on temperature at time t over Texas. Both approaches employ a principal component analysis

to downscale high dimensional spatial data while retaining variance explained within the data. Approach 1 then applies K-nearest neighbors to predict precipitation using weighted sampling indexes of the three closest neighbors. After PCA, Approach 2 applies a linear regression to the first principal component at each grid cell of precipitation following temperature. When comparing predicted and actual precipitation across the two models, Approach 1 appears to capture the range of precipitation values more accurately but exhibits a higher mean squared error and mean absolute error, indicating poorer overall model performance. Approach 2 predicts low level precipitation across the entire time series and thus has lower mean squared error and mean absolute error values due to a lack of appropriate extremes.

If the goal of the model prediction was to minimize MSE and MAE, then Approach 2 would be a better model. Both models would be significantly improved by the optimization of chosen model parameters and the inclusion of additional climate variables, including pressure, to predict precipitation.