#### Step 1: Data

- 1) How many data samples are included in the dataset?
  - There are about 3047 records of data in the dataset
- 2) Which problem will this dataset try to address?
  - The goal is to predict cancer mortality rate (aka TARGET\_deathRate column in the dataset)
- 3) What is the minimum value and the maximum value in the dataset?
  - The minimum value of the column TARGET\_deathRate is 59.7
  - The maximum value of the column TARGET\_deathRate is 362.8
- 4) How many features in each data samples?
  - There are 33 features in each data sample
  - There is 1 target column TARGET deathRate
- 5) Does the dataset have any missing information? E.g., missing features.
  - Yes, the dataset does have missing features. I Noticed few blanks in PctSomeCol18 24 column and PctPrivateCoverageAlone
- 6) What is the label of this dataset?
  - The label is "TARGET deathRate"
- 7) How many data will you use for training, validation and testing?
  - I will use 70% for training, 10% validation and 20% for testing
- 8) What kind of data pre-processing will you use for your training dataset? For preparing the dataset for training, I applied the following pre-processing steps:
  - A. Handling Missing Values
    - The column PctSomeCol18\_24 was dropped because it contained over 2000 missing values, which would significantly reduce data quality if retained.
    - For PctEmployed16\_Over and PctPrivateCoverageAlone, missing values were replaced with the mean of their respective columns, ensuring no loss of records while maintaining the overall distribution.
  - B. Incompatible Columns
    - binnedInc (categorical): Initially considered one-hot encoding, but due to model compatibility issues and lack of strong predictive value, it was ultimately removed.
    - Geography (string): Dropped since it is not required for prediction and string-based identifiers are not usable by regression or neural network models.

Model	Test R-squared	
	0.5004	
Linear regression	0.5001	
DNN-16	0.4250	
DNN-30-8	0.3340	
DNN-30-16-8	0.2305	
DNN-30-16-8-4	0.1986	

## Step 3: Objective

Used Mean Squared Error as the loss function in my code
criterion = nn.MSELoss()

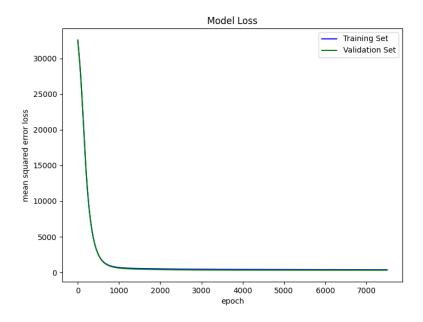
## Step 4: Optimization

Used Stochastic Gradient Descent (SGD) to train all of my models
optimizer = optim.SGD(model.parameters(), Ir=learning\_rate)

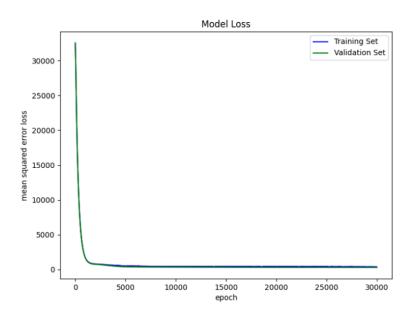
MODEL	LR 0.1	LR 0.01	LR 0.001	LR 0.0001
Linear Regression	Epoch = 700	Epoch = 700	Epoch = 3000	Epoch = 25000
	= 0.5001	= 0.4977	= 0.4961	= 0.4943
DNN-16	Sigmoid Epoch = 350	Sigmoid Epoch = 1000	Sigmoid Epoch = 6000	Sigmoid Epoch = 7500
	= 0.4250	= 0.5058	= 0.5195	= 0.4614
DNN-30-8	Sigmoid Epoch = 180	Sigmoid Epoch = 900	Sigmoid Epoch = 3800	Sigmoid Epoch = 30000
	= 0.3376	= 0.5121	= 0.5031	= 0.4936
DNN-30-16-8	Tanh Epoch = 200	Sigmoid Epoch = 3000	Sigmoid Epoch = 4000	Sigmoid Epoch = 3000
	= 0.2306	= 0.5346	= 0.4614	= 0.4603
DNN-30-16-8-4	Tanh Epoch = 90	Tanh Epoch = 500	Tanh Epoch = 2500	Tanh Epoch = 20000
	= 0.1985	= 0.4251	= 0.4920	= 0.4573

Step 6: Model Performance

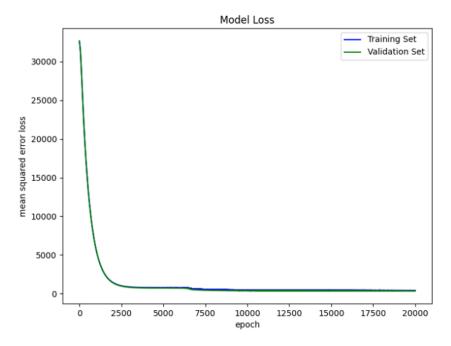
• Model\_DN\_16, LR\_0.0001



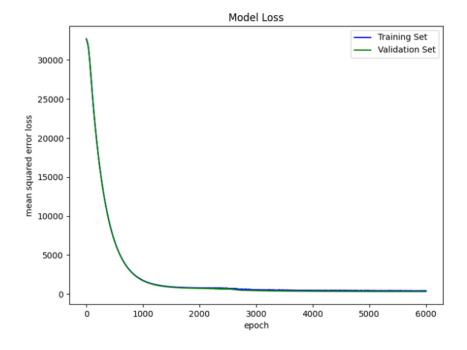
• Model\_DN\_30\_8, LR\_0.0001



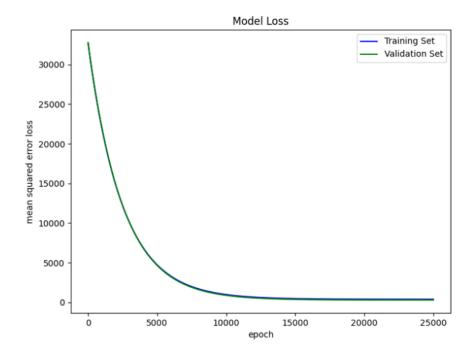
Model\_DN\_30\_16\_8\_4, LR\_0.0001



• Model\_DN\_30\_16\_8, LR\_0.0001



• Model\_Linear Regression, LR\_0.0001



# NOTE:

REST OF THE MODEL PERFORMANCE PLOTS ARE SAVED IN THE IMAGE FOLDER OF ZIP FILE PROVIDED