

**Definition 1** (Predictive Modeling). *The process of developing a mathematical tool or model that generates an accurate prediction.*

### **Why do predictive models fail?**

1. Inadequate pre-processing of the data.
2. Inadequate model validation.
3. Unjustified extrapolation, e.g.:the application of the model  $\rightarrow$  data is in an application not seen by the model.
4. Over-fitting the model to the existing data.

## **0.1 Prediction Versus Interpretation**

**For applications that require accurate predictions:**

- Historical Data — Interested in accurately projecting the chances of future events, not why they occurred.
- Pricing Homes on Zillow — Interested in accurate price estimates, not how they predicted them.
- Medical — Predict patient response to a treatment based on a significant number of factors.

Secondary considerations are given to the interpretation of the data, not primary. Higher accuracy models are more complex, and as a consequence reduce interpretability.

## **0.2 Ingredients of Predictive Models**

- The best models are influenced and produced by modelers with expert knowledge and field context of the problem.
- These modelers can pre-filter irrelevant information and place more meaningful constraints on data sets.
- Modelers can also put personal biases on data.
- Predictive modeling is not a substitute for intuition, but a complement.
- Traditional experts make better decisions with results of statistical prediction.

### 0.3 Terminology

**Definition 2** (Sample, Data Point, Observation, Instance). *A single independent unit of data (A Customer), or a subset of data points (A Training Sample).*

**Definition 3** (Training Set). *Contains the data used to develop models.*

**Definition 4** (Validation Set). *Contains the data used to evaluate the performance of the final set of candidate models.*

**Definition 5** (Predictors, independent Variables, Attributes, Descriptors). *The data used as input for the prediction equation.*

**Definition 6** (Outcome, Dependent Variables, Target, Class, Response). *The outcome event, or quantity that is being predicted.*

**Definition 7** (Continuous Data). *Has natural, numeric scales. Ex: Blood pressure, cost, quantity.*

**Definition 8** (Categorical, Nominal, Attribute, or Discrete Data). *Has specific values without scale.*

**Definition 9** (Model-Building, -Training, Parameter Estimation). *The process of using data to determine values of model equations.*

### 0.4 Notation

1.  $n$  = the number of data points.
2.  $P$  = the number of predictors.
3.  $y_i$  = the  $i$ th observed value of the outcome,  $i = 1 \dots n$ .
4.  $\hat{y}_i$  = the predicted outcome of the  $i$ th data point,  $i = 1 \dots n$ .
5.  $\bar{y}$  = the average or sample mean of the  $n$  observed values of the outcome.
6.  $\mathbf{y}$  = a vector of all  $n$  outcome values.
7.  $x_{ij}$  = the value of the  $j$ th predictor for the  $i$ th data point,  $i = 1 \dots n$  and  $j = 1 \dots P$ .
8.  $\mathbf{x}_i$  = a collection of the  $P$  predictors for the  $i$ th data point,  $i = 1 \dots n$ .
9.  $\mathbf{X}$  = a matrix of  $P$  predictors for all data points; this matrix has  $n$  rows and  $P$  columns.
10.  $\mathbf{X}'$  = the transpose of  $\mathbf{X}$ ; this matrix has  $P$  rows and  $n$  columns.

## 0.5 Other Notational Guidelines

1.  $C$  = the number of classes in a categorical outcome.
2.  $C_l$  = the value of the  $l$ th class level.
3.  $p$  = the probability of an event.
4.  $p_l$  = the probability of the  $l$ th event.
5.  $P_r[\cdot]$  = the probability of event.
6.  $\sum_{i=1}^n$  = the summation operator over the index  $i$ .
7.  $\Sigma$  = the theoretical covariance matrix.
8.  $E[\cdot]$  = the expected value of  $[\cdot]$ .
9.  $f(\cdot)$  = a function of  $[\cdot]$ ;  $g(\cdot)$  and  $h(\cdot)$  also represent functions throughout the text.
10.  $\beta$  = an unknown or theoretical model coefficient.
11.  $b$  = an estimated model coefficient based on a sample of data points.