Modeling For Inference

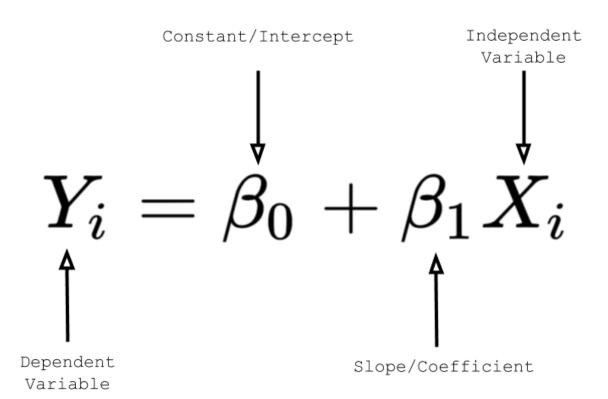
Modeling for inference

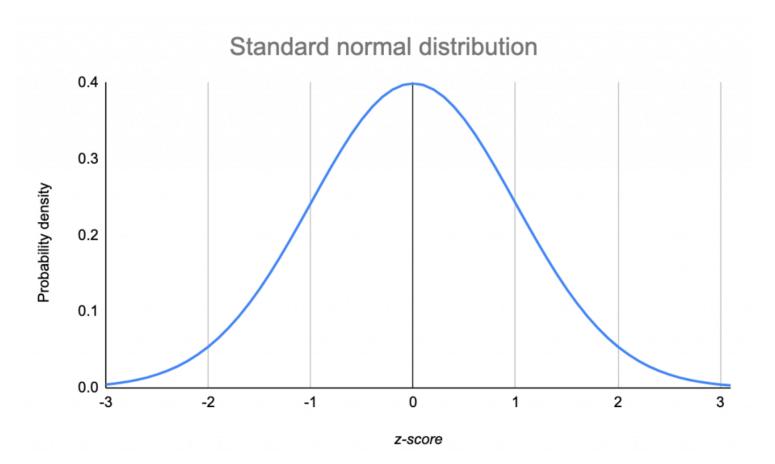
- Descriptive or causal inference
- Focus on understanding the data generating process and its component parts
- Concerned with statistical significance
- Theory driven
- Hypothesis testing
- Regression
- White box

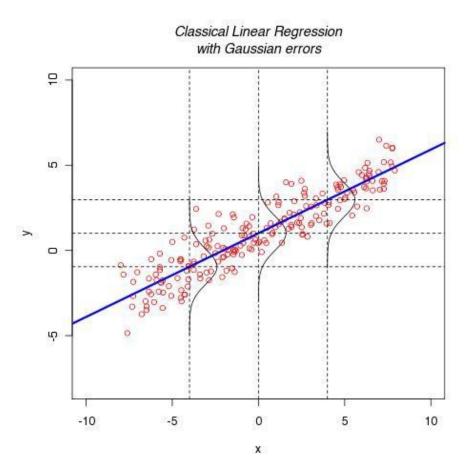
Modeling for prediction

- Forecasting
- Not concerned with model components
- Concerned with out of sample prediction
- Less theory driven
- Wider variety of methods used
- Black box

Inferential Modeling With Regression





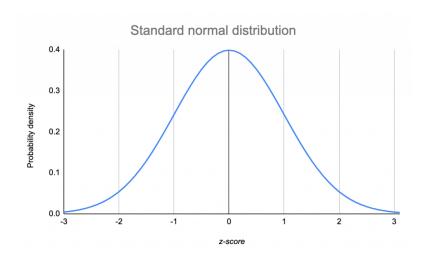


Generalized Linear Models

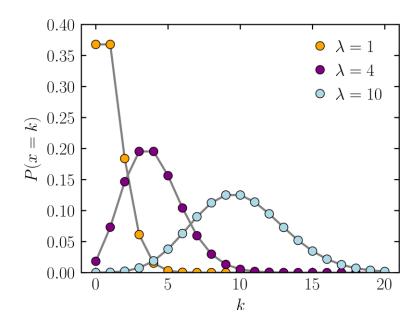
- 1. An exponential distribution for modeling Y
- 2. A linear predictor
- 3. A link function

Probability Distribution: A mathematical function that gives the probability of a random phenomenon occurring in terms of its sample space.

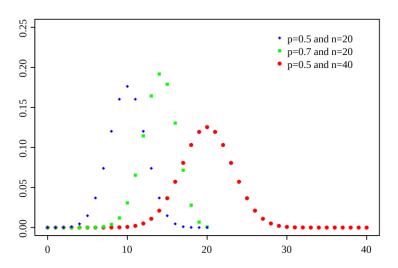
Normal: Continuous random variables



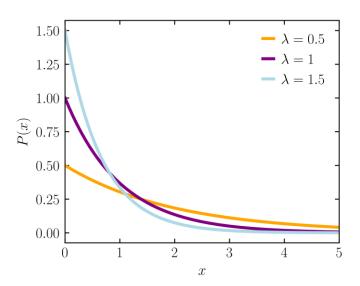
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- Poisson and negative binomial: Count of events



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- Binomial: Number of positive results in independent experiments.



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- Poisson and negative binomial: Count of events
- Binomial: Number of positive results in independent experiments.
- Exponential, Gamma, Weibull: Time to event.



A linear predictor

$$\beta_0 + \beta_1 X_i$$

A link function

• "Links" the linear predictor to the probability distribution.

Given link function g:

$$E(Y|X) = \mu = g^{-1}(\beta_0 + \beta_1 X_i)$$

Distribution	Support of distribution	Typical uses	Link name	Link function, $\mathbf{X}oldsymbol{eta}=g(\mu)$	Mean function
Normal	real: $(-\infty, +\infty)$	Linear-response data	Identity	$\mathbf{X}oldsymbol{eta}=\mu$	$\mu = \mathbf{X} \boldsymbol{eta}$
Exponential	real: $(0,+\infty)$	Exponential-response data, scale parameters	Negative inverse	$\mathbf{X}oldsymbol{eta} = -\mu^{-1}$	$\mu = -(\mathbf{X}oldsymbol{eta})^{-1}$
Gamma					
Inverse Gaussian	real: $(0,+\infty)$		Inverse squared	$\mathbf{X}oldsymbol{eta}=\mu^{-2}$	$\mu = (\mathbf{X}oldsymbol{eta})^{-1/2}$
Poisson	integer: $0,1,2,\ldots$	count of occurrences in fixed amount of time/space	Log	$\mathbf{X}oldsymbol{eta} = \ln(\mu)$	$\mu = \exp(\mathbf{X}oldsymbol{eta})$
Bernoulli	integer: $\{0,1\}$	outcome of single yes/no occurrence	-	$\mathbf{X}oldsymbol{eta} = \ln\!\left(rac{\mu}{1-\mu} ight)$	$\mu = rac{\exp(\mathbf{X}oldsymbol{eta})}{1+\exp(\mathbf{X}oldsymbol{eta})} = rac{1}{1+\exp(-\mathbf{X}oldsymbol{eta})}$
Binomial	integer: $0,1,\ldots,N$	count of # of "yes" occurrences out of N yes/no occurrences		$\mathbf{X}oldsymbol{eta} = \ln\!\left(rac{\mu}{n-\mu} ight)$	
	integer: $[0,K)$			$\mathbf{X}oldsymbol{eta} = \ln\!\left(rac{\mu}{1-\mu} ight)$	
Categorical	K-vector of integer: $[0,1]$, where exactly one element in the vector has the value 1	outcome of single K-way occurrence			
Multinomial	$ extit{ extit{K-vector of integer: } [0,N]}$	count of occurrences of different types (1 K) out of N total K-way occurrences			

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- 1. An exponential distribution for modeling Y Theory driven
- 2. A linear predictor ← The hard part
- 3. A link function \leftarrow Done by R

How is your degree different from a stats or information science degree?

What makes modeling human behavior so difficult?

Vocabulary

- Inferential modeling
- Predictive modeling
- GLM
- Probability distribution
- Link function
- · Normal, Poisson, Binomial, and Exponential distributions