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GLM I: Introduction to Generalized Linear Models

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Ratemaking and Product Management Seminar
March 27–29, 2017
San Diego, CA

Overview

Overview of GLMs

Personal Injury Claims

Intercept Only Models

One Continuous Predictor

One Discrete Predictor

Many Predictors

Key Concepts

Standard Linear Model Specification

$$y = \beta_0 + x_1\beta_1 + \cdots + x_k\beta_k + \epsilon \quad \text{with } \epsilon \in N(0, \sigma^2)$$

Standard Linear Model Specification

$$y = \beta_0 + x_1\beta_1 + \cdots + x_k\beta_k + \epsilon \quad \text{with } \epsilon \in N(0, \sigma^2)$$

A better way to think about this would be

$$\mathbb{E}[y] = \beta_0 + x_1\beta_1 + \cdots + x_k\beta_k$$

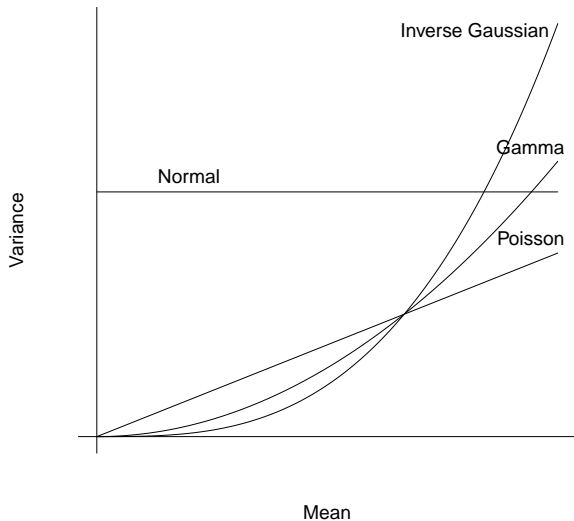
where $y \in N(\mu, \sigma^2)$ and $\mu = \beta_0 + x_1\beta_1 + \cdots + x_k\beta_k$ is the linear predictor.

Generalized Linear Model Specification

$$g(\mathbb{E}[y]) = \beta_0 + x_1\beta_1 + \cdots + x_k\beta_k + \text{offset}$$

1. The link function is g
2. The distribution of y is a member of the exponential family
3. The explanatory variables x_i may be continuous or discrete
4. Offset terms have a known coefficient of 1 in the linear predictor

Mean–Variance Relationship



Personal Injury Dataset

The dataset contains 22,036 settled personal injury claims. These claims arose from accidents occurring from July 1989 through January 1999. This is the `persinj.xls` dataset featured in the book by de Jong & Heller [2].

I have taken a random sample of 200 claims.

The variables are:

1. Settled Amount
2. Injury codes
3. Legal representation
4. Accident month
5. Report month
6. Finalization month
7. Operational time

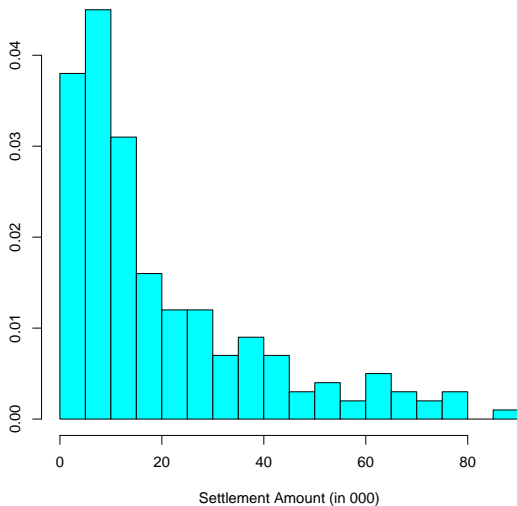
Derived variables:

1. Injured count
2. Accident injury code
3. Report delay
4. Settlement delay

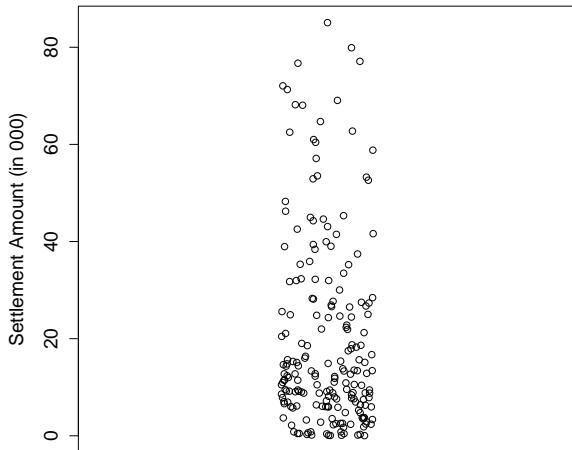
Variable Descriptions

Variable	Type	Comments
Settled Amount	Cont	range: \$40 to \$85,000
Injury Codes	Cat	Injury level: 1, 2, ..., 6 = death, 9 = missing
Legal Rep.	Bin	Attorney involved? 1 = Yes, 0 = No
Accident Month	Coded	1 = July 1989, 120 = June 1999
Report Month	Coded	same as accident month
Fin. Month	Coded	same as accident month
Injured Count	Count	Number of persons injured: 1, 2, ..., 5
Acc. Injury	Cat	Highest injury code among those injured
Report Delay	Cont	# months between accident and report
Settle. Delay	Cont	# months between report and settlement

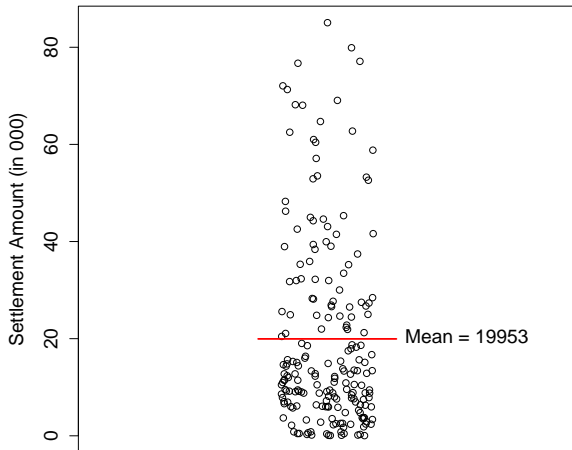
Histogram of Settlement Amount



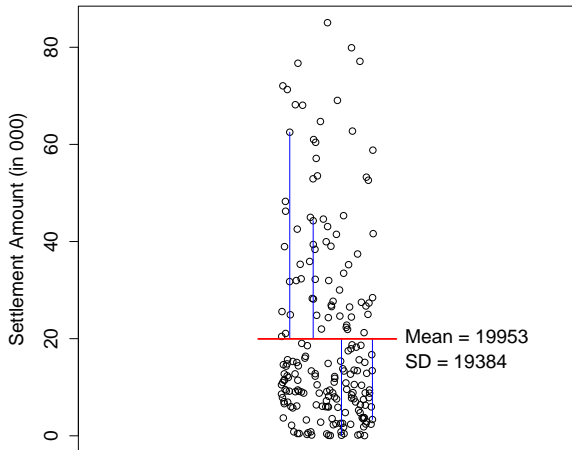
Distribution of Settlement Amount



Settlement Amount: mean



Settlement Amount: mean & standard deviation



Linear Model—Intercept only

Call:

```
lm(formula = total ~ 1, data = spinj)
```

Residuals:

Min	1Q	Median	3Q	Max
-19913	-13570	-7199	7591	65110

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	19953	1371	14.56	<2e-16 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 19380 on 199 degrees of freedom

Generalized Linear Model—Normal Id—Intercept only

```
Call: glm(formula = total ~ 1,
          family = gaussian(link = identity), data = spinj)
```

```
Deviance Residuals:
```

Min	1Q	Median	3Q	Max
-19913	-13570	-7199	7591	65110

```
Coefficients:
```

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	19953	1371	14.56	<2e-16 ***

```
---
```

```
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
(Dispersion parameter for gaussian family taken to be 375744867)
```

```
Null deviance: 7.4773e+10  on 199  degrees of freedom  
Residual deviance: 7.4773e+10  on 199  degrees of freedom  
AIC: 4519.5
```

```
Number of Fisher Scoring iterations: 2
```

Generalized Linear Model—Gamma Id—Intercept only

```
Call: glm(formula = total ~ 1,
           family = Gamma(link = identity), data = spinj)
```

```
Deviance Residuals:
```

Min	1Q	Median	3Q	Max
-3.2293	-0.9588	-0.4165	0.3407	1.9043

```
Coefficients:
```

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	19953	1371	14.56	<2e-16 ***

```
---
```

```
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
(Dispersion parameter for Gamma family taken to be 0.9438079)
```

```
Null deviance: 252.05  on 199  degrees of freedom  
Residual deviance: 252.05  on 199  degrees of freedom  
AIC: 4366.6
```

```
Number of Fisher Scoring iterations: 3
```


Generalized Linear Model—Gamma Log—Intercept only

```
Call: glm(formula = total ~ 1,
          family = Gamma(link = "log"), data = spinj)
```

```
Deviance Residuals:
```

Min	1Q	Median	3Q	Max
-3.2293	-0.9588	-0.4165	0.3407	1.9043

```
Coefficients:
```

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	9.9011	0.0687	144.1	<2e-16 ***

```
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```

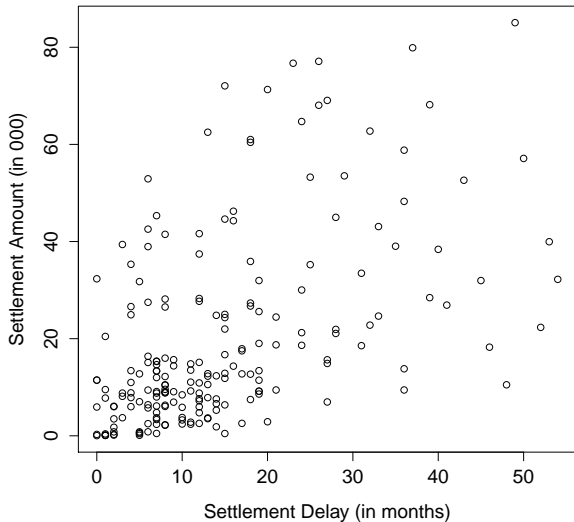
```
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
(Dispersion parameter for Gamma family taken to be 0.9438079)
```

```
Null deviance: 252.05  on 199  degrees of freedom  
Residual deviance: 252.05  on 199  degrees of freedom  
AIC: 4366.6
```

```
Number of Fisher Scoring iterations: 6
```

Settlement Amount vs. Settlement Delay



Linear Model—Intercept and Slope

Call:

```
lm(formula = total ~ settle.delay, data = spinj)
```

Residuals:

Min	1Q	Median	3Q	Max
-37059	-10395	-5085	4366	51957

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	7614.05	1861.85	4.089	6.28e-05	***
settle.delay	832.30	97.44	8.542	3.50e-15	***

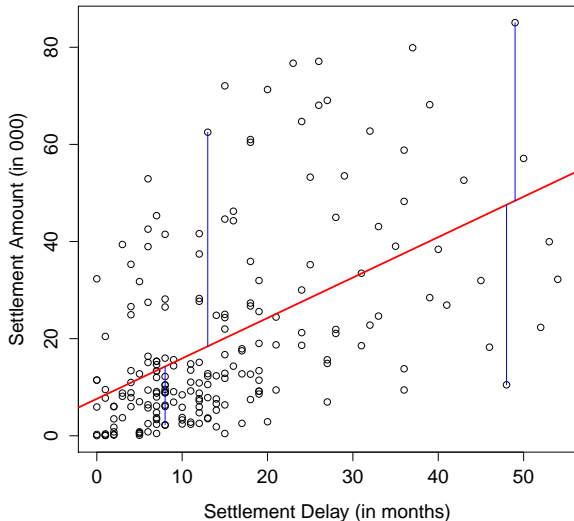
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 16610 on 198 degrees of freedom

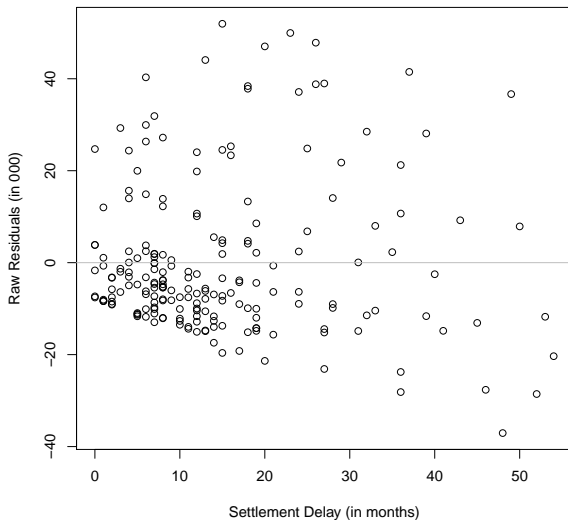
Multiple R-squared: 0.2693, Adjusted R-squared: 0.2656

F-statistic: 72.96 on 1 and 198 DF, p-value: 3.504e-15

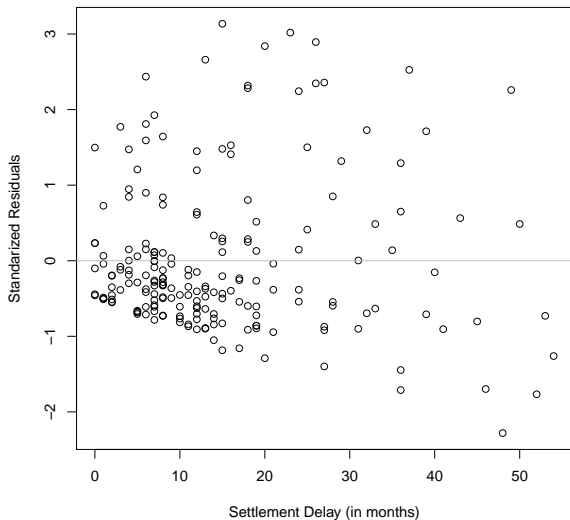
Settlement Amount vs. Delay: Least Squares Line



Raw Residuals vs. Settlement Delay



Standardized Residuals vs. Settlement Delay



Many Flavors of Residuals

Raw $y - \hat{y}$ or $y - \mu$ or $y - \mathbb{E}[y]$

Pearson $(y - \mu) / \sqrt{V}$

Deviance $\text{sgn}(y - \mu) \sqrt{\text{deviance}}$

Standardized Divide residual by $\sqrt{1 - h}$, which aims to make its variance constant; where h are the diagonal elements of the projection ('hat') matrix, $H = X(X^t X)^{-1} X^t$, which maps y into \hat{y}

Studentized Divide residual by $\sqrt{\phi}$; where ϕ is the scale parameter

Stan & Stud Divide residual by both standardized and studentized adjustments

Deviance

Distribution	Contribution to Squared Deviance
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Normal	$(y_i - \mu_i)^2$
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Poisson	$2\{y_i \log(y_i/\mu_i) - y_i + \mu_i\}$
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Gamma	$2\{-\log(y_i/\mu_i) + (y_i - \mu_i)/\mu_i\}$
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Inverse Gaussian	$(y_i - \mu_i)^2 / (\mu_i^2 y_i)$
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Gamma Log GLM–Intercept and Slope

```
Call: glm(formula = total ~ settle.delay,
          family = Gamma(link = "log"), data = spinj)
```

```
Deviance Residuals:
```

	Min	1Q	Median	3Q	Max
	-3.0008	-0.8017	-0.3145	0.1991	1.8982

```
Coefficients:
```

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	9.187173	0.102174	89.917	< 2e-16 ***
settle.delay	0.040473	0.005347	7.569	1.39e-12 ***

```
---
```

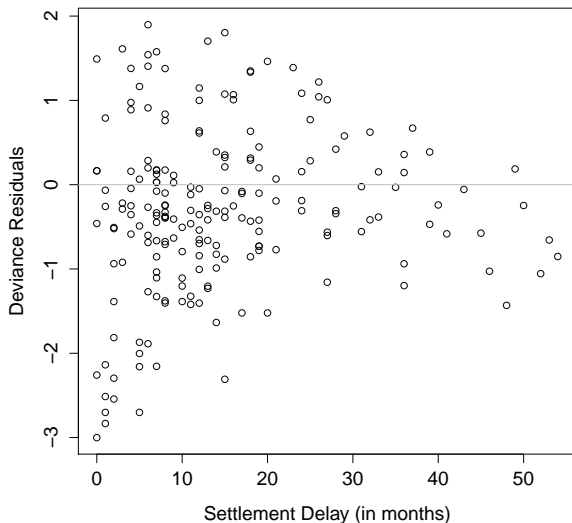
```
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
(Dispersion parameter for Gamma family taken to be 0.8310652)
```

```
Null deviance: 252.05  on 199  degrees of freedom  
Residual deviance: 206.47  on 198  degrees of freedom  
AIC: 4321.8
```

```
Number of Fisher Scoring iterations: 7
```

Gamma Model: Deviance Residuals vs. Settlement Delay



Poisson Log GLM—Intercept and Slope

```
Call: glm(formula = tot.amt ~ settle.delay,
           family = poisson(link = "log"), data = spinj)
```

```
Deviance Residuals:
```

Min	1Q	Median	3Q	Max
-229.41	-92.18	-42.51	35.74	299.99

```
Coefficients:
```

	Estimate	Std. Error	z value	Pr(> z)
(Intercept)	9.323e+00	8.583e-04	10862.1	<2e-16 ***
settle.delay	3.280e-02	3.338e-05	982.7	<2e-16 ***

```
---
```

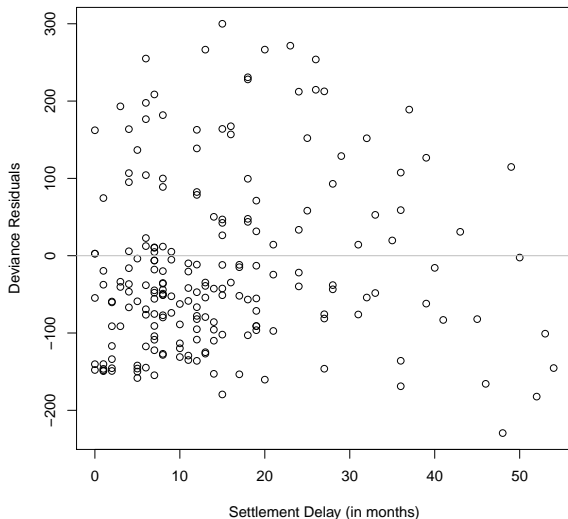
```
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
(Dispersion parameter for poisson family taken to be 1)
```

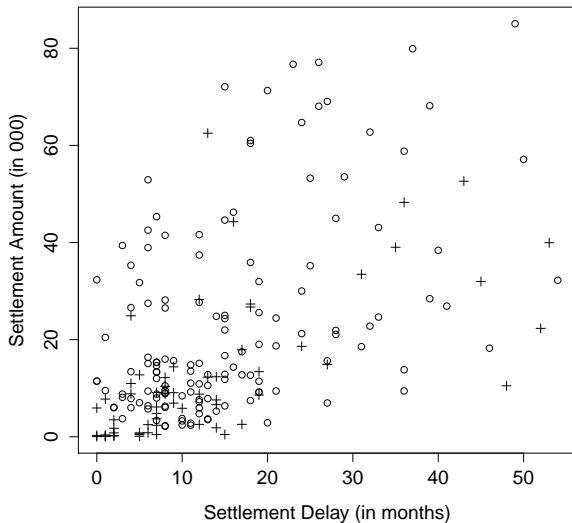
```
Null deviance: 3366902  on 199  degrees of freedom  
Residual deviance: 2515703  on 198  degrees of freedom  
AIC: 2517928
```

```
Number of Fisher Scoring iterations: 5
```

Poisson Model: Deviance Residuals vs. Settlement Delay



Legal Representation?



Gamma Log GLM—Legal Representation?

```
Call: glm(formula = total ~ settle.delay + legrep,
          family = Gamma(link = "log"), data = spinj)
```

```
Deviance Residuals:
```

	Min	1Q	Median	3Q	Max
	-2.8152	-0.8183	-0.3115	0.2864	2.6778

```
Coefficients:
```

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	8.64459	0.13476	64.148	< 2e-16 ***
settle.delay	0.04112	0.00539	7.628	9.96e-13 ***
legrep1	0.70702	0.13989	5.054	9.85e-07 ***

```
---
```

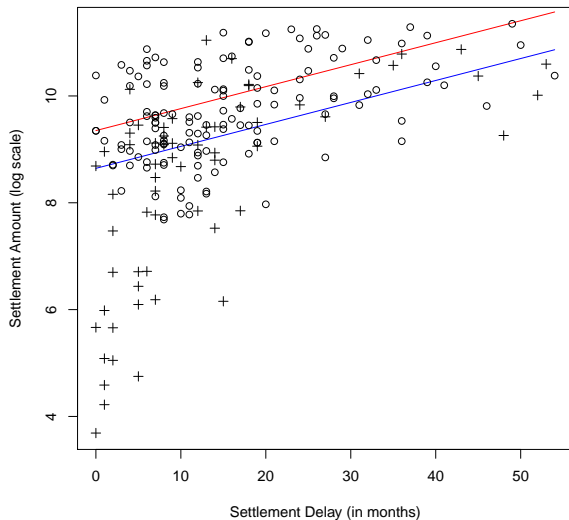
```
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
(Dispersion parameter for Gamma family taken to be 0.8354751)
```

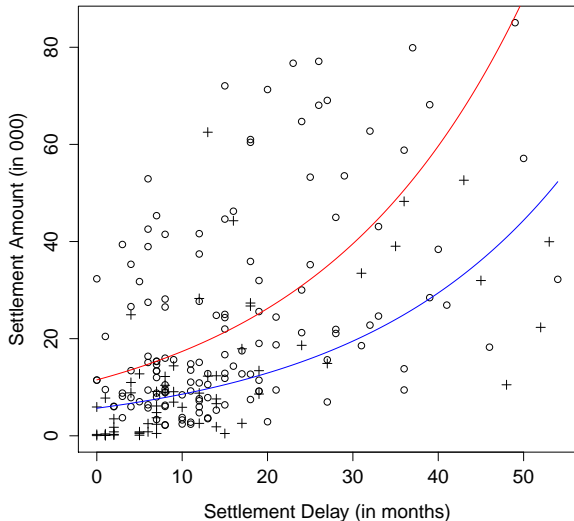
```
Null deviance: 252.05  on 199  degrees of freedom
Residual deviance: 186.98  on 197  degrees of freedom
AIC: 4300.9
```

```
Number of Fisher Scoring iterations: 8
```

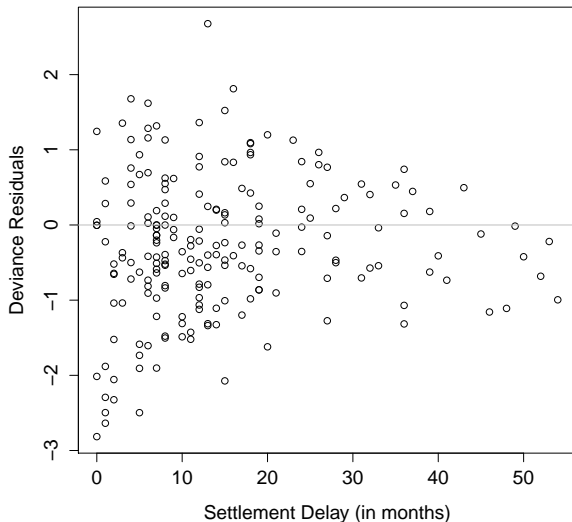
Legal Representation: Linear Predictor



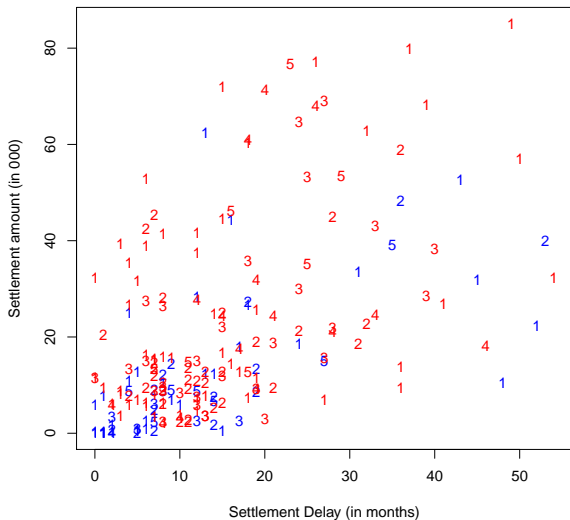
Legal Representation: Fitted Values



Legal Representation: Deviance Residuals



Number of Injured Persons



Gamma Log GLM—Many Predictors

```
Call: glm(formula = total ~ settle.delay + legrep + inj.count,
          family = Gamma(link = "log"), data = spinj)
```

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	8.722358	0.141721	61.546	< 2e-16	***
settle.delay	0.042138	0.005222	8.069	7.38e-14	***
legrep1	0.786161	0.139411	5.639	6.01e-08	***
inj.count2	-0.300230	0.160788	-1.867	0.0634	.
inj.count3	-0.416338	0.177247	-2.349	0.0198	*
inj.count4	-0.216891	0.244640	-0.887	0.3764	
inj.count5	0.005267	0.254395	0.021	0.9835	

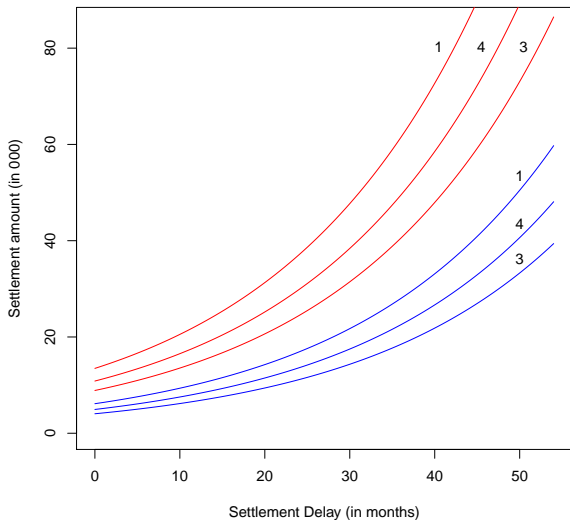
```
Null deviance: 252.05 on 199 degrees of freedom
Residual deviance: 181.44 on 193 degrees of freedom
AIC: 4302
```

```
Number of Fisher Scoring iterations: 9
```

Predicted Values

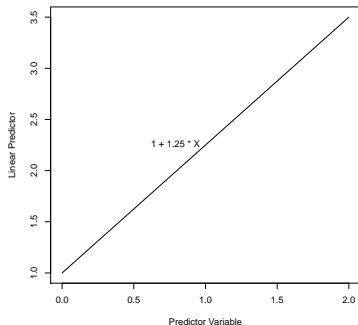
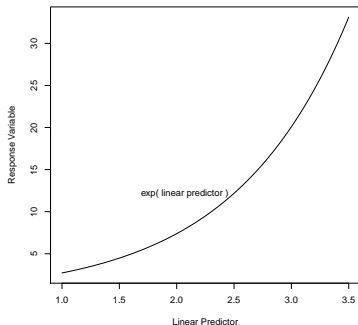
Settle Delay	Legal Rep?	Injured Count	Linear Predictor	Fitted Value
0	No	1	$8.7 + 0 \cdot 0.042 = 8.7$	$e^{8.7} = 6003$
0	Yes	1	$8.7 + 0 \cdot 0.042 + 0.79 = 9.5$	$e^{9.5} = 13360$
10	No	4	$8.7 + 10 \cdot 0.042 - 0.22 = 8.5$	$e^{8.9} = 7332$

Many Predictors: Fitted Values



Summary Key Concepts: Link Function

The link function is the bridge between the space of the linear predictor and the space of the response.






Summary Key Concepts: Deviance

The deviance tells us how to measure the distance between an observation and its fitted value.

Distribution	Contribution to Squared Deviance
Normal	$(y_i - \mu_i)^2$
Poisson	$2\{y_i \log(y_i/\mu_i) - (y_i - \mu_i)\}$
Gamma	$2\{-\log(y_i/\mu_i) + (y_i - \mu_i)/\mu_i\}$
Inverse Gaussian	$(y_i - \mu_i)^2/(\mu_i^2 y_i)$

References

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