

# **Antitrust Notice**

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#### GLM I: Introduction to Generalized Linear Models

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Casualty Actuarial Society Ratemaking and Product Development Seminar March 19–21, 2012 Philadelphia, PA

#### Overview

Overview of GLMs

Personal Injury Claims

Intercept Only Models

One Continuous Predictor

One Discrete Predictor

Many Predictors

**Key Concepts** 

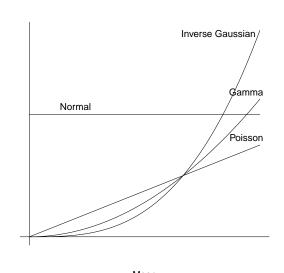
#### Basic GLM Specification

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

- 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

Variance



Mean

#### 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:

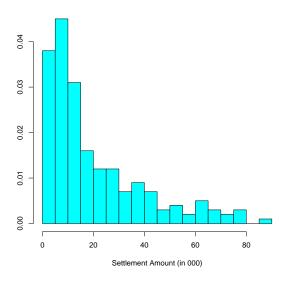
- 1. Injured count
- 2. Accident injury code

- 3. Report delay
- 4. Settlement delay

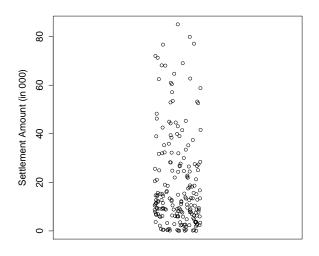
# Variable Descriptions

| Variable       | Type  | Comments  |
|----------------|-------|---|
| Settled Amount | Cont  | range: \$40 to \$85,000                               |
| Injury Codes   | Cat   | Injury level: $1, 2, \dots, 6 = death, 9 = missing$   |
| Legal Rep.     | Bin   | Attorney involved? $1 = \text{Yes}$ , $0 = \text{No}$ |
| 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: $1, 2, \ldots, 5$          |
| 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                |
|                |       |   |

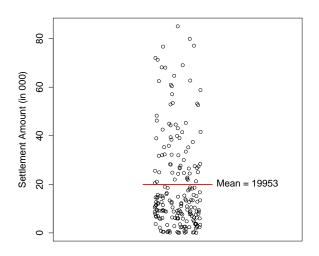
### Histogram of Settlement Amount



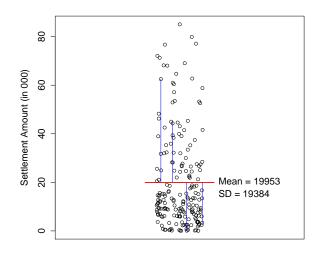
#### Distribution of Settlement Amount



#### Settlement Amount: mean



#### Settlement Amount: mean & standard deviation



## Linear Model—Intercept only

```
Call:
lm(formula = total ~ 1, data = spinj)
Residuals:
  Min 10 Median 30
                           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

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### 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 ATC: 4519.5

Number of Fisher Scoring iterations: 2

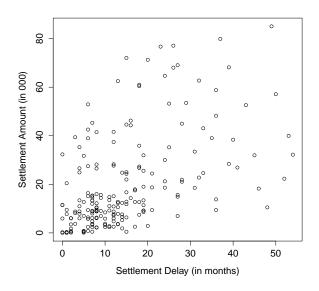
### Generalized Linear Model—Gamma Id—Intercept only

```
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
ATC: 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 ***
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
ATC: 4366.6
Number of Fisher Scoring iterations: 6
```

### Settlement Amount vs. Settlement Delay

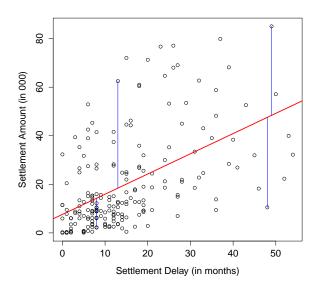


### Linear Model-Intercept and Slope

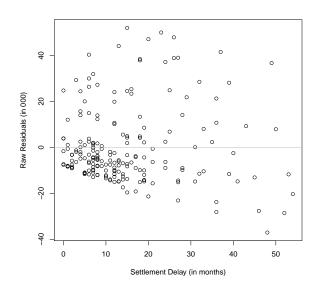
```
Call:
lm(formula = total ~ settle.delay, data = spinj)
Residuals:
  Min 10 Median 30
                           Max
-37059 -10395 -5085 4366 51957
Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept) 7614.05 1861.85 4.089 6.28e-05 ***
settle.delay 832.30
                        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

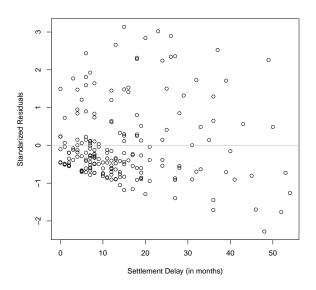
#### Settlement Amount vs. Delay: Least Squares Line



#### Raw Residuals vs. Settlement Delay



# Standarized 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  $\operatorname{sgn}(y - \mu)\sqrt{\operatorname{deviance}}$ 

Standarized Divide residual by  $\sqrt{1-h}$ , which aims to make its variance constant; where h are the diagonal elements of the projection ('hat') matrix,  $H=X(X^tX)^{-1}X^t$ , which maps y into  $\hat{y}$  Studentized Divide residual by  $\sqrt{\phi}$ ; where  $\phi$  is the scale parameter Stan & Stud Divide residual by both standarized and studentized adjustments

#### Deviance

| 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 | $(y_i-\mu_i)^2/(\mu_i^2y_i)$              |

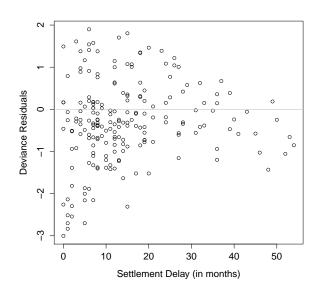
## Gamma Log GLM-Intercept and Slope

```
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 ATC: 4321.8

Number of Fisher Scoring iterations: 7

### Gamma Model: Deviance Residuals vs. Settlement Delay



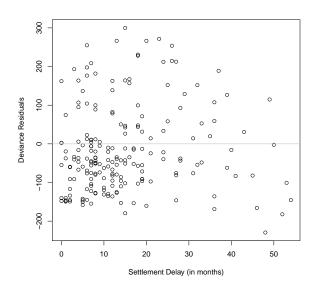
## Poisson Log GLM-Intercept and Slope

Number of Fisher Scoring iterations: 5

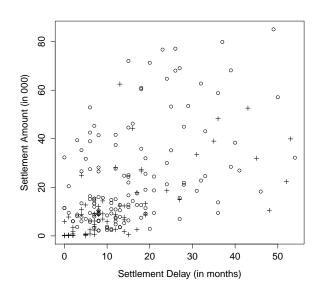
```
Call: glm(formula = tot.amt ~ settle.delay,
          family = poisson(link = "log"), data = spinj)
Deviance Residuals:
   Min
             1Q Median 3Q
                                     Max
-229.41 -92.18 -42.51 35.74 299.99
Coefficients:
             Estimate Std. Error z value Pr(>|z|)
(Intercept) 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
ATC: 2517928
```

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# Poisson Model: Deviance Residuals vs. Settlement Delay



#### Legal Representation?



## Gamma Log GLM-Legal Representation?

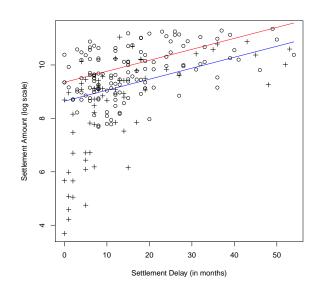
```
Call: glm(formula = total ~ settle.delay + legrep,
         family = Gamma(link = "log"), data = spinj)
Deviance Residuals:
   Min
             10 Median
                             3Q
                                     Max
-2.8152 -0.8183 -0.3115 0.2864 2.6778
Coefficients:
            Estimate 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

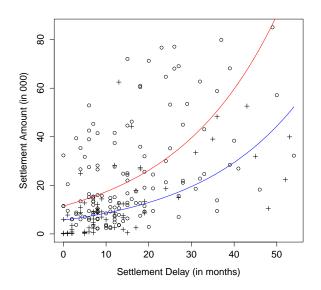
AIC: 4300.9

Number of Fisher Scoring iterations: 8

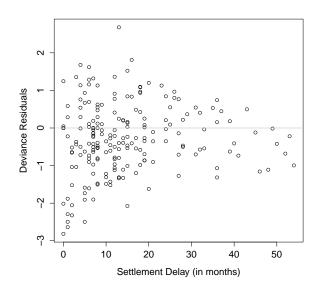
### Legal Representation: Linear Predictor



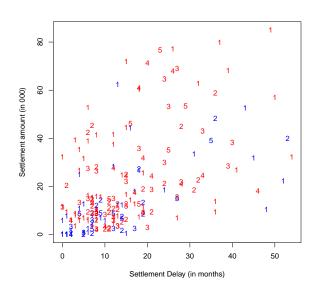
#### Legal Representation: Fitted Values



#### Legal Representation: Deviance Residuals



#### Number of Injured Persons



#### Gamma Log GLM-Many Predictors

Number of Fisher Scoring iterations: 9

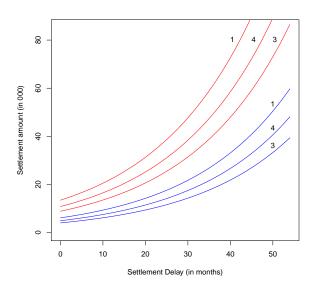
```
Call: glm(formula = total ~ settle.delay + legrep + inj.count,
        family = Gamma(link = "log"), data = spinj)
Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept)
            8.722358
                     0.141721 61.546 < 2e-16 ***
settle.delay
            0.786161  0.139411  5.639  6.01e-08 ***
legrep1
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
```

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#### **Predicted Values**

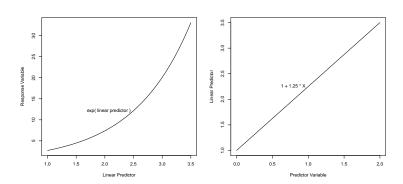
| Settle<br>Delay | _   | Injured<br>Count | Linear Predictor                    | Fitted<br>Value                                     |
|-----------------|-----|------------------|-------------------------------------|---|
| 0               | No  | 1                | $8.7 + 0 \cdot 0.042 = 8.7$    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

### 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 | $(y_i-\mu_i)^2/(\mu_i^2y_i)$              |

#### References



John M. Chambers, William S. Cleveland, Beat Kleiner, and Paul A. Tukey. *Graphical Methods for Data Analysis*.

The Wadsworth Statistics/Probability Series. Wadsworth International Group, Belmont, California, 1983.



Annette J. Dobson.

An introduction to Generalized Linear Models. Chapman & Hall, London, 1990.



Edward W. Frees.

Regression Modeling with Actuarial and Financial Applications. Cambridge University Press, 2010.

#### References

- James Hardin and Joseph Hilbe. Generalized Linear Models and Extensions. Stata Press, College Station, Texas, 2001.
- Piet De Jong and Gillian Z. Heller.

  Generalized Linear Models for Insurance Data.

  Cambridge University Press, 2008.
- W.N. Venables and B.D. Ripley.

  Modern Applied Statistics with S.

  Springer New York, 2002.