



The Actuarial Profession

making financial sense of the future

A Simple Multi-State Reserving Model

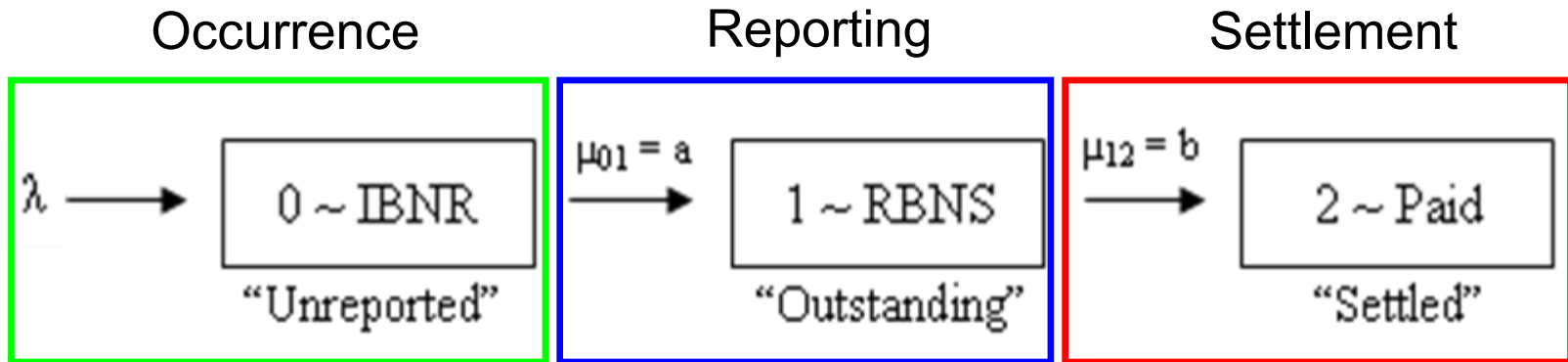
ASTIN Colloquium

Wednesday, 20th June 2007

What I'm Going to Talk to You About

- Treating claims development as a process
- Looking at simplest possible model
- Deriving closed form expressions
- Demonstrating practical extensions to model
- Discussing results of initial investigations
- Presenting a more general model form

The Simplest Model



- Inspired by Norberg, Hesselager and Hachtmeister
- Losses arriving as a Poisson process, rate λ
 - into Incurred But Not Reported (IBNR) State 0 (“Unreported”)
- Transitioning at instantaneous rate a
 - into Reported But Not Settled (RBNS) State 1 (“Outstanding”)
- Transitioning at instantaneous rate b
 - into Paid State 2 (“Settled”)

Instantaneous Transition Rates

$$a = \lim_{h \rightarrow 0} \frac{\Pr(\text{Claim in State 1 at } t + h \mid \text{Loss in State 0 at } t)}{h}$$

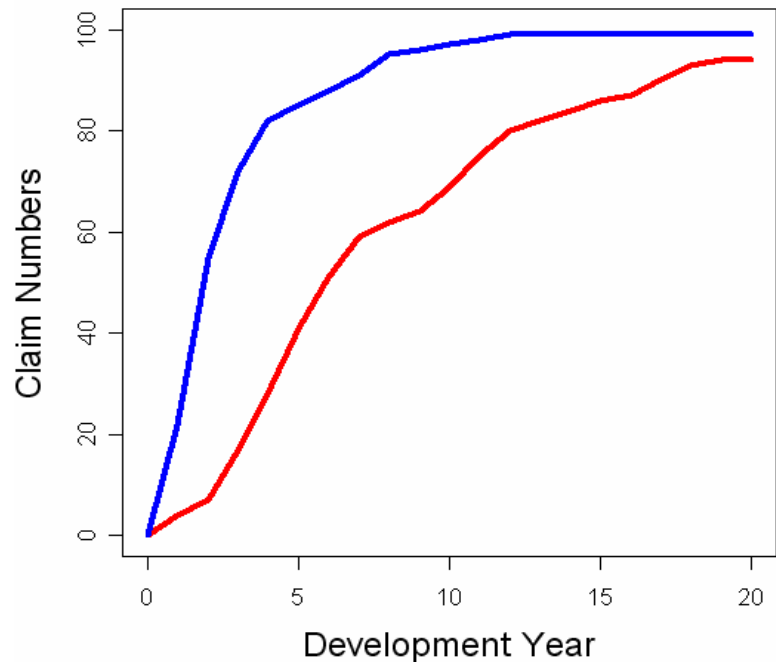
- Limiting probability of moving from State 0 to State 1
 - analogous to “Force of Mortality” used by actuaries
- Waiting times in State 0 distributed as Exponential random variable
 - parameter value a
 - expected value of $1/a$ and variance of $(1/a)^2$
- Higher rate reduces waiting time in state
 - e.g. $a = 2$ implies average waiting time of 0.5 years

Claims Number Development

Observed Data

t	IBNR $N(0,t)$	RBNS $N(1,t)$	Settled $N(2,t)$	Total $N(.,t)$
0	0	0	0	0
1	77	18	4	99
2	44	48	7	99
3	27	55	17	99
4	17	54	28	99
5	14	44	41	99
6	11	37	51	99
7	8	32	59	99
8	4	33	62	99
9	3	32	64	99
10	2	28	69	99
...
∞	0	0	99	99

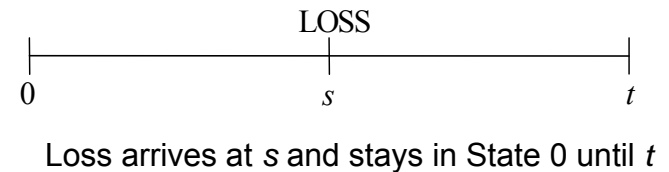
Simulated Paid and Incurred Claims Development Data



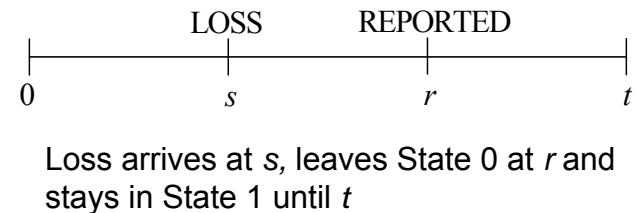
During Accident Year ($t \leq 1$)

$$E[N(s,t)] = \lambda t * [\alpha(0,t), \alpha(1,t), \alpha(2,t)]$$

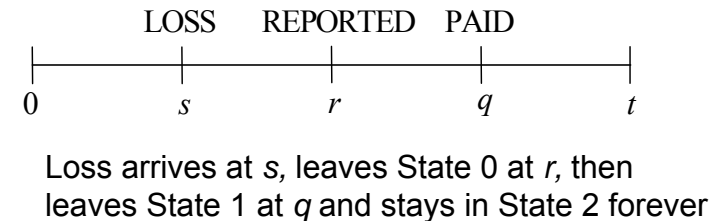
$$\begin{aligned} \lambda t * \alpha(0,t) &= \int_0^t \lambda \exp[-a(t-s)] ds \\ &= \frac{\lambda}{a} [1 - \exp(-at)] \end{aligned}$$



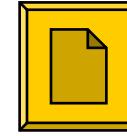
$$\begin{aligned} \lambda t * \alpha(1,t) &= \int_0^t \lambda \int_s^t a \exp[-a(r-s)] \exp[-b(t-r)] dr ds \\ &= \frac{\lambda a}{(a-b)b} [1 - \exp(-bt)] - \frac{\lambda}{(a-b)} [1 - \exp(-at)] \end{aligned}$$



$$\begin{aligned} \lambda t * \alpha(2,t) &= \int_0^t \lambda \int_s^t a \exp[-a(r-s)] \int_r^t b \exp[-b(q-r)] dq dr ds \\ &= \lambda a \left\{ \frac{t}{a} - \frac{[1 - \exp(-at)]}{a^2} - \frac{[1 - \exp(-bt)]}{(a-b)b} + \frac{[1 - \exp(-at)]}{a(a-b)} \right\} \end{aligned}$$



After Accident Year ($t > 1$)



- Transition Intensity Matrix

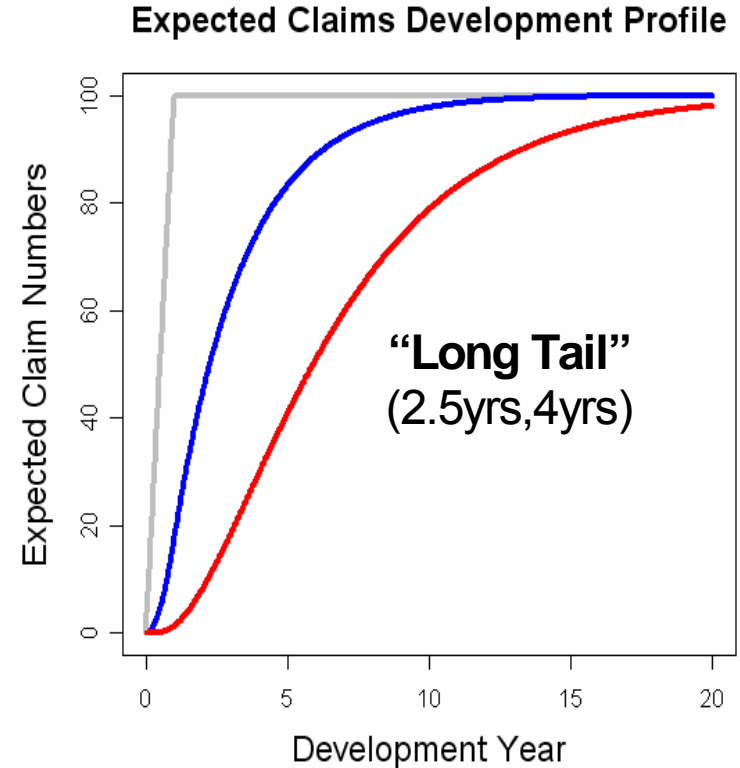
$$Q = \begin{pmatrix} -a & a & 0 \\ 0 & -b & b \\ 0 & 0 & 0 \end{pmatrix}$$

- Kolmogorov's Forward Equation

$$\begin{aligned} E[N(s,t)] &= \lambda \pi(t) \\ &= \lambda \pi(1) \exp[Q(t-1)] \end{aligned}$$

where,

$$\pi(1) = [\alpha(0,1), \alpha(1,1), \alpha(2,1)]$$



Conditional (Next Value) Expectations

- Beyond $t=1$, only transitions between states can occur
- Run-off process fully specified, by transition intensity matrix Q
- Transition probability matrix for reporting interval d

$$P(d) = \exp[Q.d]$$

- Expected next values can be determined from

$$E[N(s, y, t) | N(s, y, t-1)] = N(s, y, t-1)P(1)$$

- need to assume total claims figure n_y
- Can be used to compare actual with expected values

$$\text{SoS} = \sum_{y=1}^{10} \sum_{t=1}^{11-y} \{N(s, y, t) - E[N(s, y, t) | N(s, y, t-1)]\}^2$$

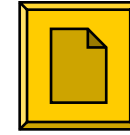
- Can search for parameter values n_y , a and b

Conditional Distribution

- For a multi-state Markov chain
 - conditional upon total number of claims = n_y
 - number of claims in each state at time t (≥ 1)
 - distributed as Multinomial($n_y, \pi(t)$)

$$\Pr(n_{IBNR}, n_{RBNS}, n_{Paid}) = \frac{n_{IBNR}! n_{RBNS}! n_{Paid}!}{(n_{IBNR} + n_{RBNS} + n_{Paid})!} (p_{IBNR})^{n_{IBNR}} (p_{RBNS})^{n_{RBNS}} (p_{Paid})^{n_{Paid}}$$

- where
 - $n_{IBNR}, n_{RBNS}, n_{Paid}$ are claims counts at time t
 - $p_{IBNR}, p_{RBNS}, p_{Paid}$ are $\pi(0,t), \pi(1,t)$ and $\pi(2,t)$ respectively



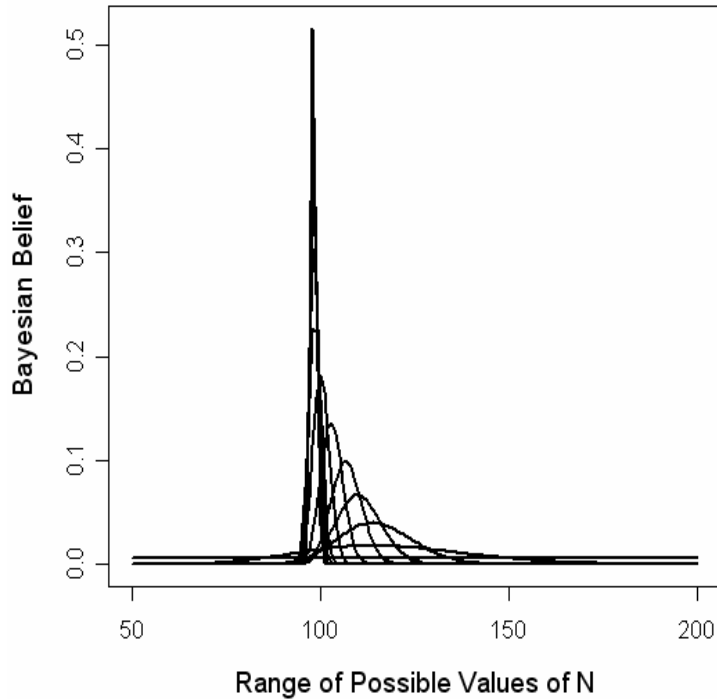
Bayesian Formulation

$$\Pr(N_y | N(s \in \{1,2\}, y, t)) = \frac{\Pr(N(s \in \{1,2\}, y, t) | N_y) \Pr(N_y)}{\sum_{\forall N_y} \Pr(N(s \in \{1,2\}, y, t) | N_y) \Pr(N_y)}$$

- Given O/S (State 1) and Paid (State 2) counts at time t
- Can calculate probabilities of different value of N_y
- Assume that $n_{IBNR} = N_y - n_{RBNS} - n_{Paid}$

Bayesian Results

Successive Bayesian Belief Profiles for Accident Year 1



Expected Values of N_y :

Accident Year	Development Year										
	0	1	2	3	4	5	6	7	8	9	10
1	125.0	118.5	115.0	110.4	107.0	103.0	100.2	98.6	98.4	98.2	98.2
2	125.0	118.5	118.3	119.1	117.1	118.1	118.1	119.7	120.7	121.5	
3	125.0	123.7	119.3	118.5	116.9	114.1	112.1	110.8	109.6		
4	125.0	113.4	102.4	103.8	103.2	103.6	103.2	103.3			
5	125.0	103.1	100.2	100.1	104.4	107.1	109.1				
6	125.0	83.1	82.7	85.2	84.4	87.3					
7	125.0	97.9	99.2	98.8	98.8						
8	125.0	87.8	93.8	95.9							
9	125.0	113.4	92.6								
10	125.0	70.8									

Standard Deviation (i.e square-root of Variance) for N_y :

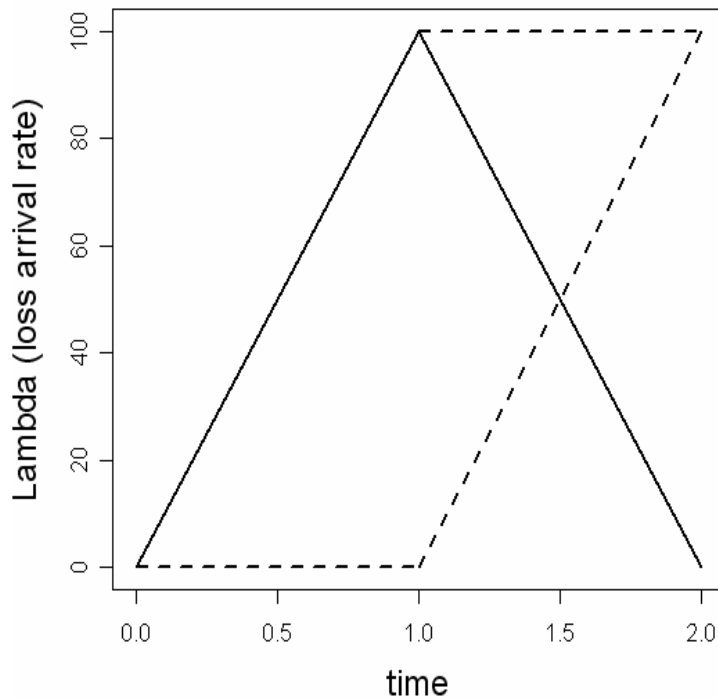
Accident Year	Development Year										
	0	1	2	3	4	5	6	7	8	9	10
1	43.6	22.2	10.0	6.0	4.0	2.9	2.2	1.7	1.3	1.0	0.8
2	43.6	22.2	10.1	6.1	4.2	3.0	2.2	1.6	1.2	0.9	
3	43.6	22.6	10.2	6.1	4.2	3.0	2.3	1.8	1.4		
4	43.6	21.8	9.4	5.7	3.9	2.8	2.1	1.6			
5	43.6	20.8	9.3	5.6	3.8	2.7	2.0				
6	43.6	18.0	8.4	5.2	3.5	2.5					
7	43.6	20.3	9.2	5.6	3.8						
8	43.6	18.9	8.9	5.5							
9	43.6	21.8	9.1								
10	43.6	14.7									

75% Percentile (actually, the first integer greater than the 75% percentile) Values for N_y :

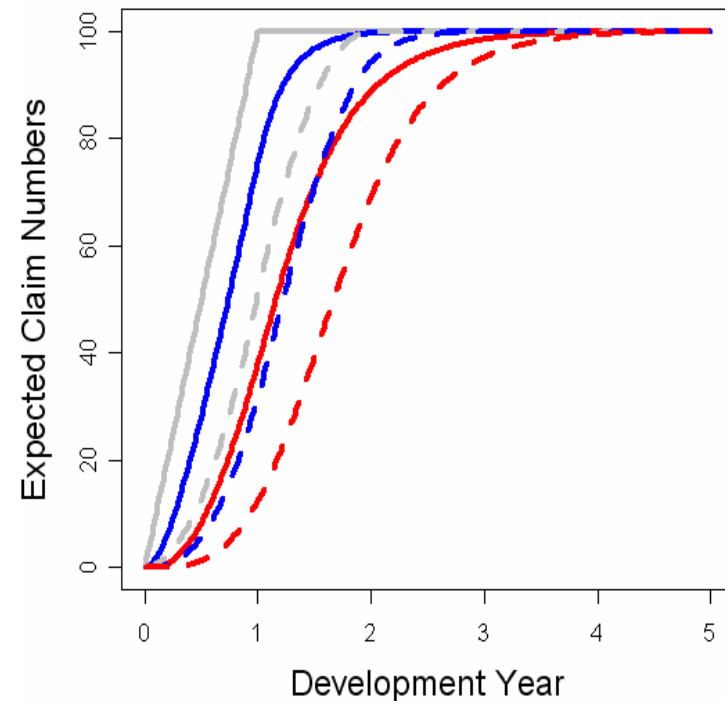
Accident Year	Development Year										
	0	1	2	3	4	5	6	7	8	9	10
1	163	133	121	114	110	105	102	100	99	99	99
2	163	133	125	123	120	120	120	121	121	122	
3	163	138	126	123	120	116	114	112	110		
4	163	127	108	108	106	105	105	104			
5	163	116	106	104	107	109	110				
6	163	94	88	89	87	89					
7	163	111	105	102	101						
8	163	100	100	99							
9	163	127	98								
10	163	79									

Extension: YOA, not Accident Year

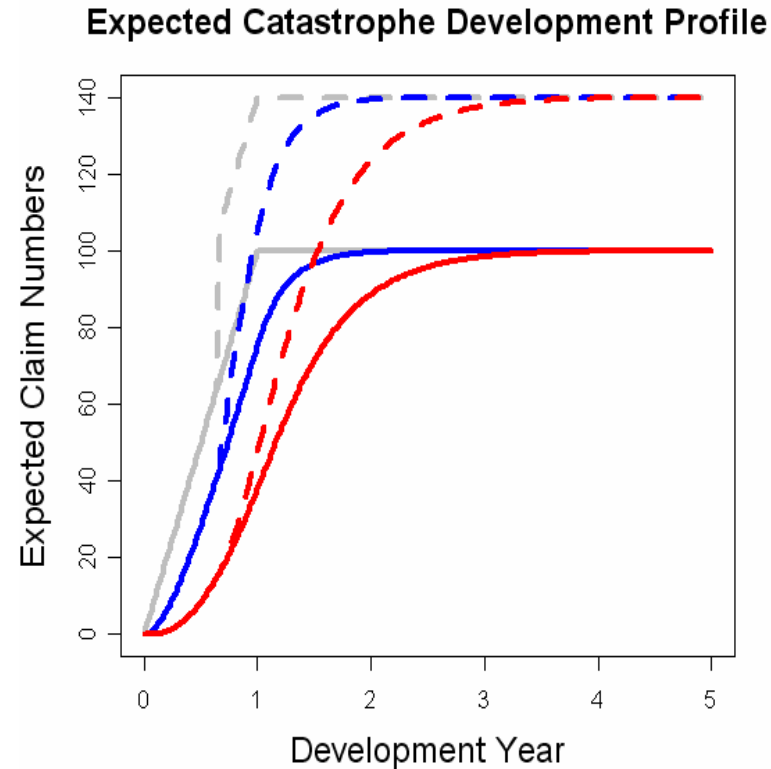
Loss Arrival Rate for Year of Account



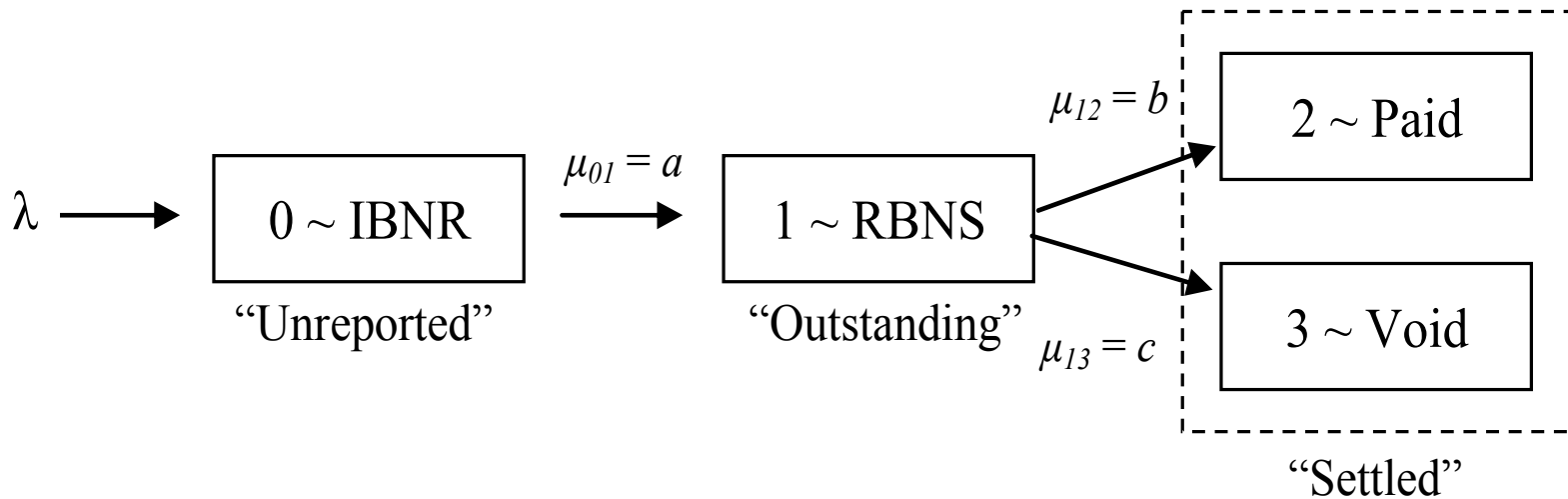
Expected YOA Development Profile



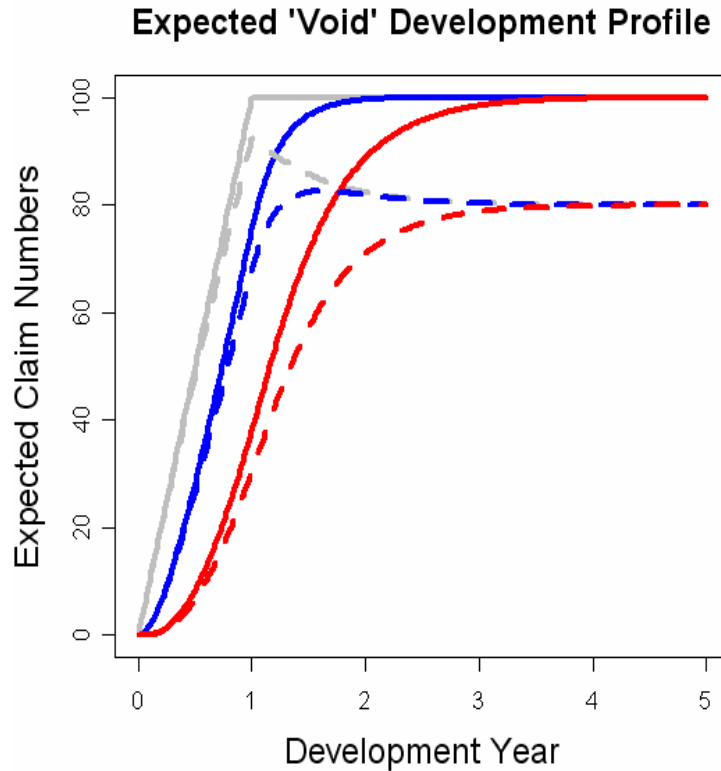
Extension: Catastrophes



Negative Development - “Void” Claims

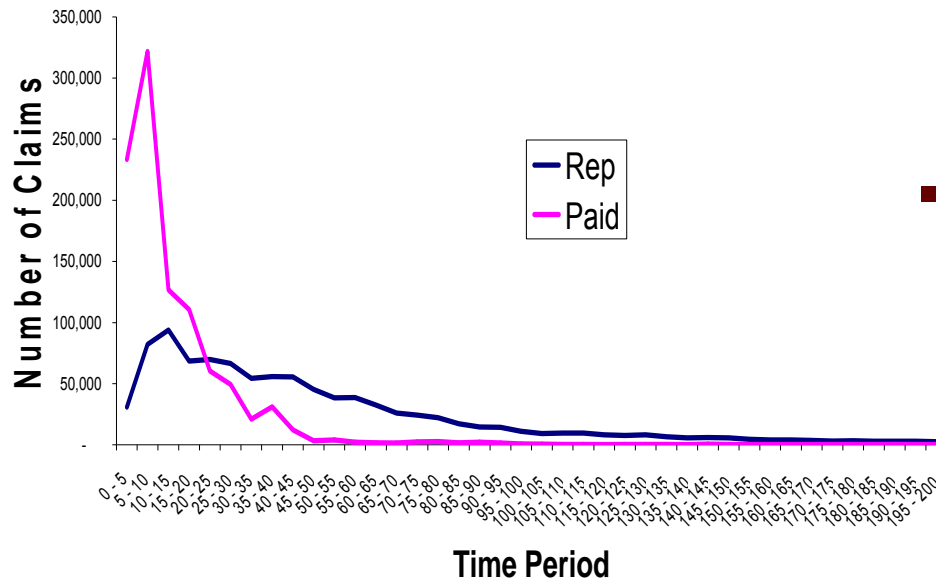


Negative Development - “Void” Claims



Initial Investigations

Reporting and Settlement Delay for General Warranty



- Initial analysis of claims level data
 - indicates Exponential assumption not valid
- Three likely reasons
 - mixing of business
 - changes in processing rates over time
 - multiple stages in settlement process

A Not So Simple General Model

- Structure
- Simulation
- Testing Projection Methods
- Link to Bornhuetter-Ferguson Method

Modelling the Claims Process

O = Occurrence Process

R = Reporting Process

S = Settlement Process

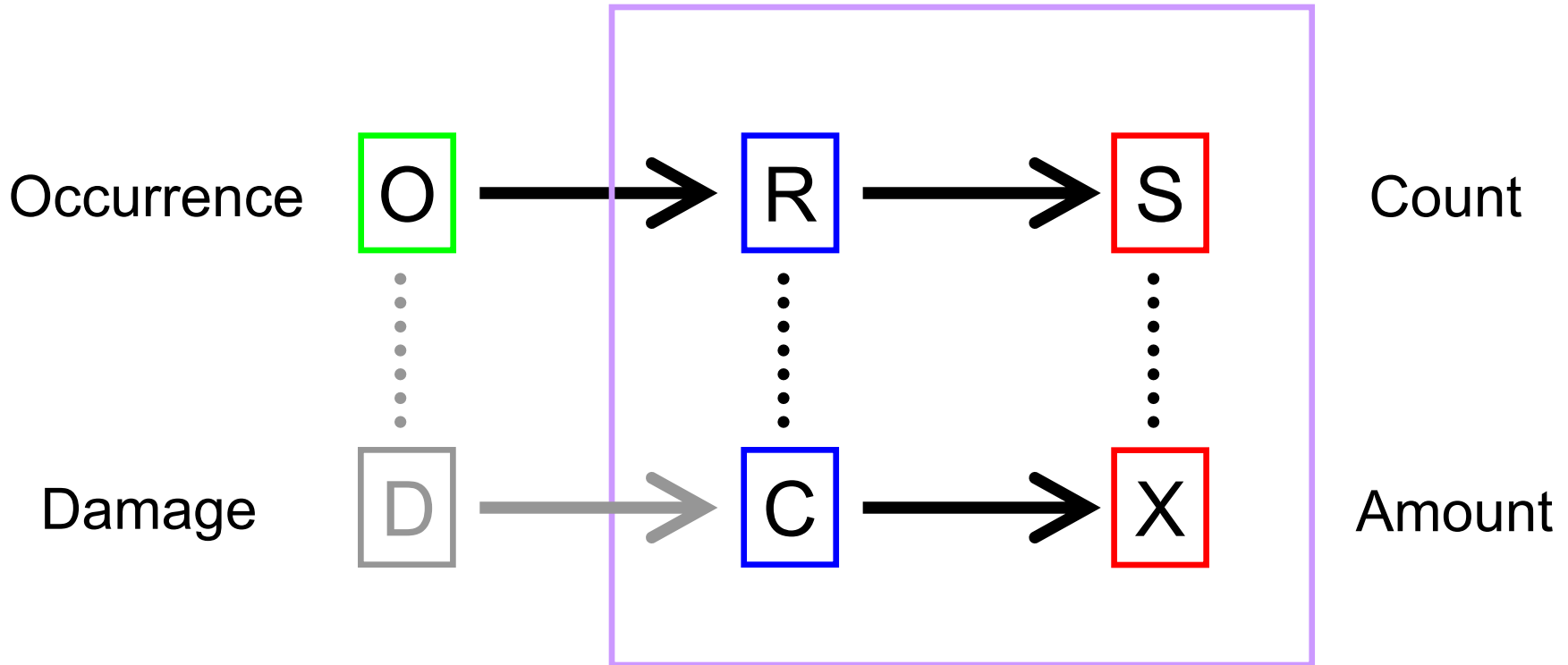
C = Case Estimation Process

X = Claims Amount Process

Case Estimates, Outstanding and Incurred Claims

- On notification of claim
 - individual case estimate struck
 - held as “Outstanding” claim until paid
 - “Paid” + “Outstanding” = “Incurred” (actually “Reported”)
- Many causes of settlement delays
 - loss-adjusting
 - quotations for work
 - restoration, repair or replacement
 - disputes – investigation, arbitration, trial

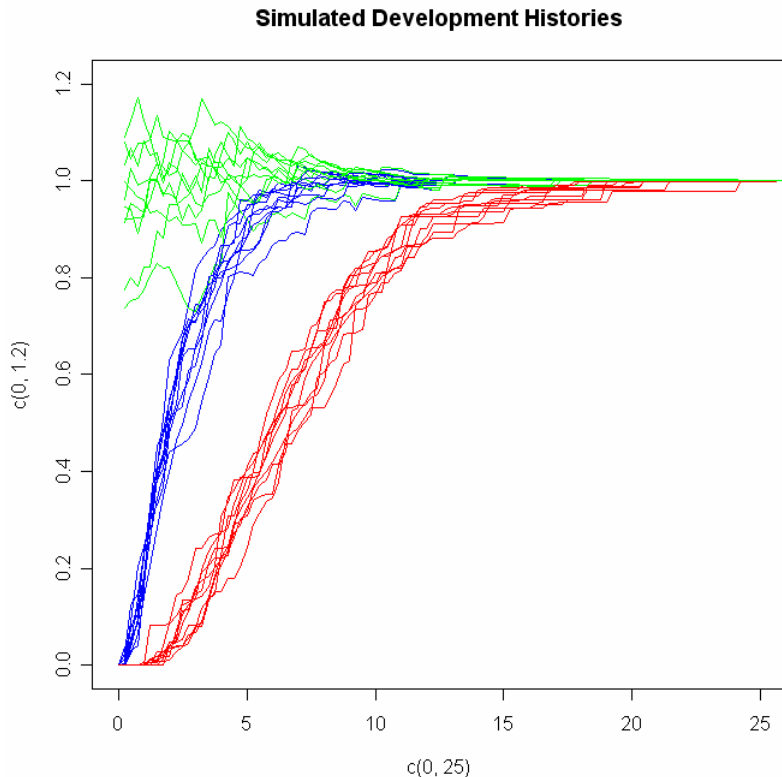
Modelling the Claims Process



Simulation is Easy...

- O_i = occurrence time for i'th claim
- R_i = reporting time for i'th claim
- S_i = settlement time for i'th claim
- CE_i = case estimate for i'th claim
- X_i = paid amount for i'th claim
- So, claim i is reported by time t if $t \geq R_i$
- and, claim i is paid by time t if $t \geq S_i$

Simulation is Easy...



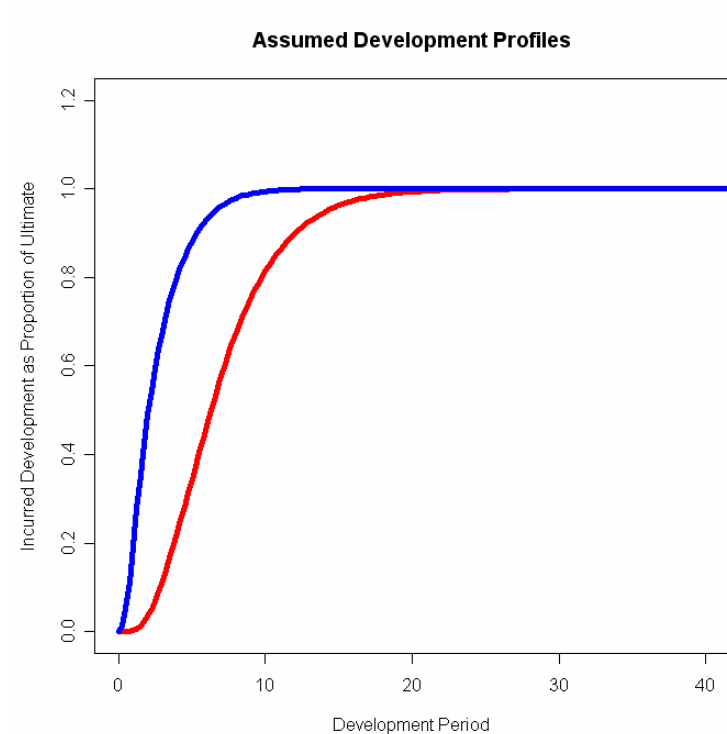
- **Red:** Paid Claims
- **Blue:** Reported Claims
- **Green:** (Estimated) Ultimate
 - applying Incurred Bornhuetter-Ferguson (IBF) method
 - based on benchmark incurred development patterns

Underlying Model

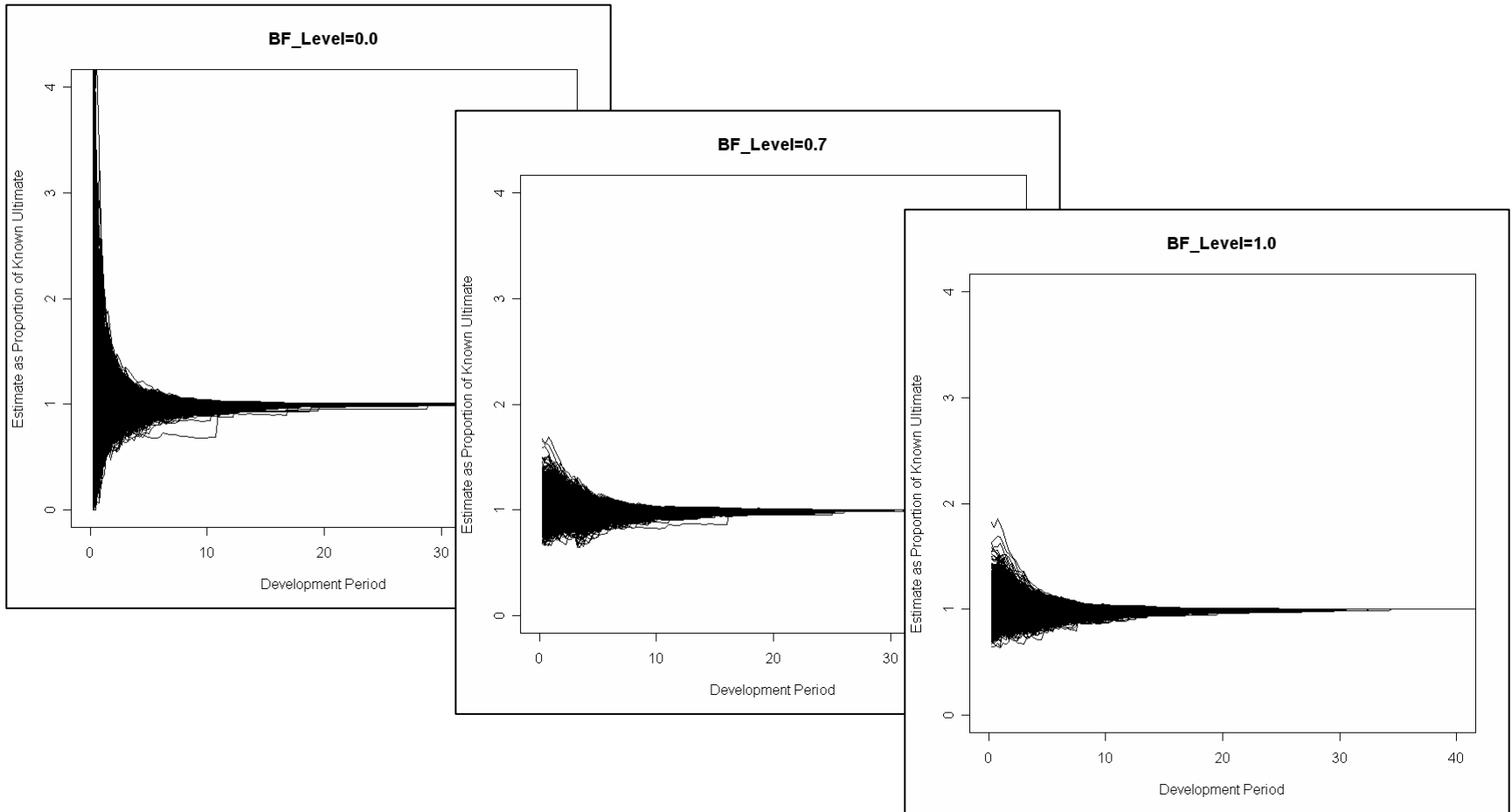
- Inception: all annual covers, incepting at 1/1
- **O** Poisson loss process, $\Lambda = 100$
- **R-O** Exponential reporting delays, 2.5 years ave
- **S-R** Gamma settlement delays
 - equivalent to sum of two Exponential random variables
 - average 2 years
 - giving average delay of 4 years
- **CE** Estimation Errors $\sim \text{Normal}(+10\%, 5\%)$
 - note positive bias, or “redundancy”
- **X** Paid Amounts $\sim \text{logNormal}(8.8, 0.8)$

More Model Details

- “Benchmark” incurred development
 - running simulation model & averaging
- In “Bornhuetter-Ferguson” framework
 - Initial estimate on known parameters
 - average (logNormal) claim size ~£9,140
 - Poisson loss arrival rate = 100
 - so, BF_Estimate ~£914k
 - “cheating”
 - just stochastic variation to worry about
 - “BF_Level” varied (0%,70%,100%)
 - percentage developed at which “Incurred Development Factor” method applied
 - using benchmark development profile



Testing Projection Methods



Unreported & Unpaid Claims Distribution

- N = total number of claims in Accident Year
 - assume $N \sim \text{Poisson}(\text{Lambda}=100)$
- ${}_R N_t$ = total Reported claims | Reported claims at t
- ${}_P N_t$ = total Paid claims | Paid claims at t
- Then
 - ${}_R N_{t=0} \equiv {}_P N_{t=0} \equiv N$ a Poisson r.v.
 - ${}_R N_{t=\infty} \equiv {}_P N_{t=\infty} \equiv n$ realisation of N

Unreported & Unpaid Claims Distribution

- ${}_R N_t \sim \text{Poisson}(\text{Lambda} \times \Pr[R > t])$
+ Number of Claims Reported by time t
- ${}_P N_t \sim \text{Poisson}(\text{Lambda} \times \Pr[S > t])$
+ Number of Claims Paid by time t
- Now
 - $1 - \Pr[R > t] \equiv \text{Incurred Development Profile}$
 - $1 - \Pr[S > t] \equiv \text{Paid Development Profile}$

Unreported & Unpaid Claims Distribution

- Take expected values for Incurred claims

$$\begin{aligned}
 E[{}_R N_t] &= E[\text{Poisson}(\text{Lambda} \times \Pr[R > t]) \\
 &\quad + \text{Number of Claims Reported by time } t] \\
 &= E[\text{Poisson}(\text{Lambda})] \times \Pr[R > t] \\
 &\quad + \text{Number of Claims Reported by time } t \\
 &= \text{Initial Estimate} \times (1 - \text{IncDevProf}) \\
 &\quad + \text{Number of Claims Reported by time } t
 \end{aligned}$$

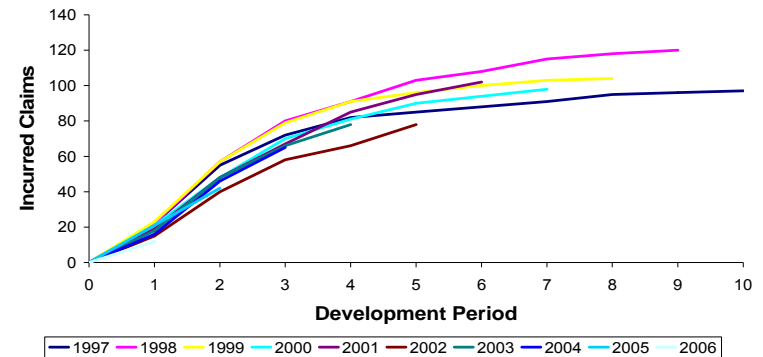
i.e. Bornhuetter-Ferguson Method

- Similar result holds for Paid claims

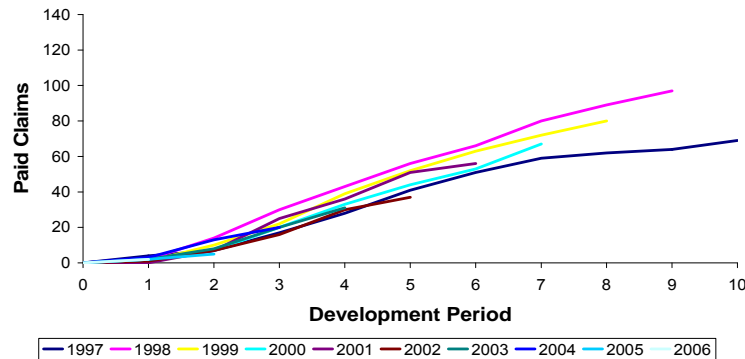
Development Curves

- Claims v. Development Period
- Check consistency of development
- Point towards higher/lower ultimates
- Incurred and Paid should converge

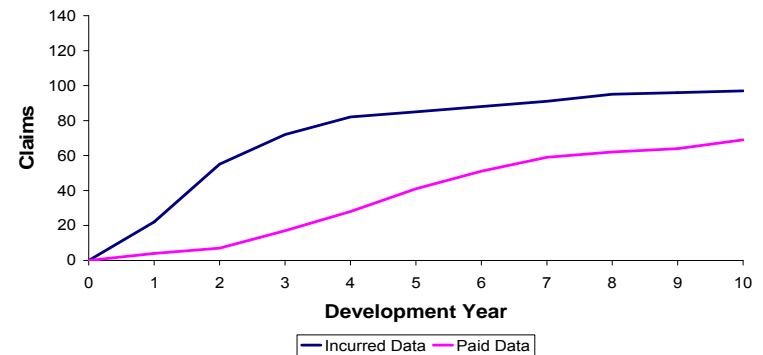
Incurred Claims Development



Paid Claims Development



Accident Year Claims Development



Why Look at Claims Level Approach?

- Linking business processes to reserving process
- Greater computing power
- Ready capture of claims processing data
- Alternative foundation for stochastic reserving and ICAs
- Can allow for known/anticipated changes in processing

What I've Talked About

- Treated claims development as a process
- Looked at simplest possible model
- Derived closed form expressions
- Demonstrated practical extensions to model
- Showed results of initial investigations
- Presented more general model form

Any Questions?