

# Industry Standard Claims Reporting Data and Claims Management

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# **Discussion topics**

- Predictive Analytics and Claims
- Data Issues
- Industry Standard Claims Submission Data for Property/Casualty
- An application to claims management (fraud)

## **Predictive Analytics and Claims**

- Predictive Analytics less prevalent in the Claims area relative to pricing/underwriting.
- Claims Management emerging as an increasingly critical competitive differentiator
  - fierce price competition, need for tighter cost control, increasing claims inflation, decreasing margins and returns
  - increasing recognition that a "positive claims experience" by customers creates opportunities for customer retention and cross-sells.
- Predictive modeling enables companies to identify claims that are likely to represent the greatest loss exposure and apply claim practices, business rules and experienced claims resources to manage these claims.

#### **The Claims Process**

- Claims notification and assignment
- Claims Investigation and evaluation
- Negotiation and settlement

#### **Decisions at Different Stages in the Claims Process**

- Assignment of claims to reps, units
- Fraud referrals, Subrogation referrals
- Complex claims identification
- Claims likely to go to litigation
- Determine claims settlement amount
- Models can be designed and structured to address each of these issues.

# **Claims Handling**

- Old way
  - Claim comes in; supervisor/adjuster examines a limited set of indicators to gauge severity of exposure
  - Claims are assigned
- New way
  - Claims are evaluated based on large set of claim characteristics
  - Scores are assigned to each claim: higher scores indicative of higher risk
  - Low risk claims can be settled quickly, while claims with high exposure can be routed for special attention. Claims with intermediate exposure can be routed through the standard process.
  - Claims are assigned based on better understanding of claim => better claimadjuster matching => better workload balance => higher productivity => better claim outcome => low leakage

### **Claims Predictive Modeling: Data Issues**

- "Level of success carriers acheive with analytics greatly depends on the quality of the data"
- ".... putting it (the data )nto a common format so you can use it consistently is a challenge..."
- Perhaps the biggest challenge is working across traditional information silos to integrate and consolidate data into a single, consistent format – which is the foundation of sound analytics "

#### **Industry Standard Claims Data**

- 98% of Auto insurers, 95% of Home insurers participate
- Data is produced and reported at regular intervals
- One of the most valuable information resources the insurance industry has collectively and collaboratively produced

#### **Current application of the Data**

- Current major use is for a comprehensive search of loss histories of applicants for pricing/ underwriting purposes.
- Rating factor relativities

# What Data is Reported

- Policy Information
  - Policy Type
  - Inception date, expiration date
  - Renewal indicator
  - Others
- Claims Information
  - Claim type
  - Number of parties involved
  - Date, time of Loss
  - Loss report date
  - Others
- Claimant Information
  - Name , Address, Date of Birth/ Age
  - Relationship to Insured
  - OTHERS

# Data (cont'd)

- Loss Information
  - Loss description
  - Loss location
  - Nature of injury
  - Body part
  - CPT, ICD-9 codes
  - Medical amount paid
  - Indemnity amount paid
  - Witness
  - Police Report
  - Hit and run
  - Single vehicle
  - Phantom vehicle
  - OTHERS

# Data (cont'd)

- Property Information
  - VIN, Year, Make
  - Anti-theft device
  - Air bag status
  - Odometer reading
  - Residential/ Commercial/ off-premise
  - Fire indicator
- Service Provider Information
  - Names, Addresses, Date of Birth
  - Professional License information
  - Individual/ Business Indicator
  - OTHERS

# **Applications other than current uses**

- Can this data, collected for submission, be used internally ?
- Why ask the question?
  - Level of information rich enough for other uses
  - Data is produced regularly
  - Data is available and accessible
  - Data is accurate, reliable and tested
  - Data is well understood

# **Claims Fraud**

- Claims fraud is a significant and costly concern
  - According to the Insurance Information Institute and the NICB, it adds about 10% to Loss and loss adjustment expenses, resulting in \$24 billion of losses each year
  - Of this, \$5 billion are attributable to WC claims
  - Costs policy holders an estimated amount of \$200-\$300 a year in increased premium

- P/C insurers detect less than 20% of fraud
- Life/disability insurers detect less than 10% of their fraud
- Health care insurers detect only about 1% of their fraud

# Claims Fraud (cont'd)

- Current shape of Fraud Detection
  - Most use manual reviews and industry standard 'red flags'
  - Moving towards predictive systems and technology used by credit card companies and banks to detect fraud

# **A Claims Fraud Model**

- A definition of fraud
  - Independent medical examination
  - Referrals to SIU (Special Investigative Unit)
  - Verified/ Mitigated fraud
- Information on
  - Claim
  - Claimant
  - Policy
  - Bills
  - Payments
  - Others

- Probability of Fraud (BI) = f (Policy Inception date, Policy Expiration date, Age of Claimant, Injury date time, Injury report date, Hire Date, Injury Type, Nature of Injury, Body Part, ICD-9 codes, CPT codes, Witness, Police Report, ER, AWW, Job class, First payment date, number of distinct medical providers, distance provider-claimant, attorney, date of attorney introduction, ...)
- The algorithm processes and analyzes the data. Weights are calculated for each variable. Weighted values are combined to produce a model score. This model score expresses the likelihood of a claim being fraudulent
- Models are calibrated prior to implementation. Depending on business needs, threshold values of scores are selected. Claims within the threshold scores are routed for review and validation.
- The model score can change over the life of the claim

# **Claims Fraud Model (cont'd)**

#### Variables known to be predictive

- Report lag, Treatment lag
- Provider type
- Number of providers
- Age of vehicle
- CPT codes, ICD-9 codes
- Limits, deductibles
- Distance claimant-provider
- First Claim payment date
- Years of employment
- Witness
- Policy Start date Claim date
- Policy End Date Claim date
- OTHERS

# **PROS and CONS**

- Most, if not all of these are available from standard industry claims submission data efforts.
  - Additionally, the Loss Description field provides significant scope for knowledge derivation using Text mining
  - External data can be combined with available fields (zip code)
- Not all desired input/ predictor variables are available or can be derived off the Industry Standard claims data
  - a number of reporting fields are optional, and therefore not reported
  - Information is at an aggregated level granular data provides deeper insights
  - Bills and payments details are not available
  - Lag between event and data collection for reporting

# RECAP

Data prepared for industry standard claims submission contains Claimant, Policy, Loss and other data that have useful predictive value to support decision making

Data is tested, reliable, well understood, available and accessible

- Fraud example shown. Many other applications for claims management
  - Identification of severe medical only claims
  - Identification of claims likely to go to litigation
  - Identification of claims with recovery/ subrogation potential
  - Claim duration analysis
  - Estimating claims settlement values