Actuaries Putting Al into Practice in General Insurance

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- How P&C (General Insurance) Lends Itself to AI/ML Applications
- Real-World Applications by CAS Actuaries
- Four Futures for Actuaries in the Wake of AI
- CAS Actuaries Leading the Profession in Data Science and the Application of AI/ML to Insurance
- Ethical AI for Actuaries

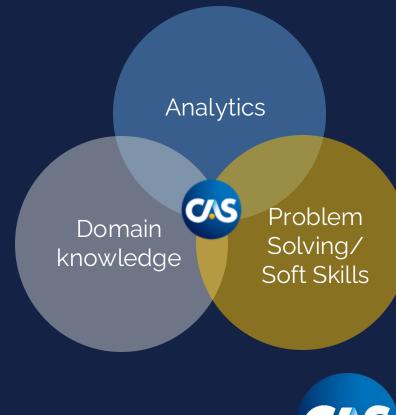


How P&C (General Insurance) Lends Itself to AI/ML Applications

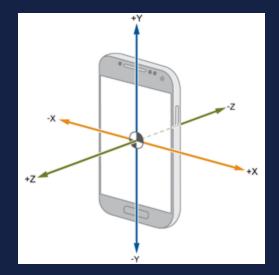


General insurance thinking used in many areas

- How to Use Data
- Emerging risks
- Climate risk reporting
- Risk Mitigation
- Weather volatility & insurability
- Population & demographic shifts
- Automated vehicles
- Cyber liability and hacking



Telematics from Mobile Phones: Advanced Modeling in Action





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Consumer Cost Savings
Gas optimization reduces costs,
pay-as-you-drive saves
infrequent drivers money
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Driving Dashboard

View driving insights, see opportunities to potentially reduce risk and premium



Crash Detection and Support Be able to reach help quickly if you're in an accident



Accurate Insurance Premiums Evaluate premiums based on more accurate and unbiased factors, including distracted or dangerous driving



Real-World Applications by CAS Actuaries



Aerial Imagery and Computer Vision Damage Estimation enables rapid claims servicing during catastrophes

Production

Analytical Solution

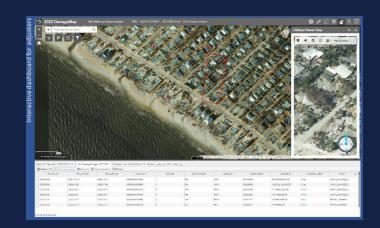
- During large scale wind and fire catastrophes, many members are impacted simultaneously - debris, flood, and evacuation orders often limit access for both member and USAA
- This solution incorporates aerial imagery, member parcel and building intelligence, and advanced computer vision to estimate the level of damage to members property

Benefits

- Allows members to receive specialized loss settlement for severe loss by a virtual adjuster; payouts occur within hours
- Provides rapid triage of a damage area for field adjuster deployment and claim resolution time

2022 Examples

- Marshal Fire in Superior, CO. 79 of 85 total losses were settled with this system.
- Hurricane Ian, FL CV system evaluated 26,000
 members in under an hour, identifying all total losses





Aerial Imagery and Computer Vision helps ensure adequate and appropriate coverage for homeowners

In Development

Analytical Solution

- Over time, the accuracy of dwelling characteristics can change due to home additions, and home improvements
- Created the ability identify home characteristics and identify changes to structural characteristics of a home from high-resolution aerial imagery over time

Benefits

- Unfortunately, in our busy lives there are situations where we make improvements to our home and do not notify USAA
- This solution benefits members by notifying them when a change is detected, providing an opportunity to ensure their full property is insured in case of a loss

Prior Image

Current Image



Exterior Home Characteristics:

- Solar
- Shed
- In-law Suite
- Detached Garage
- Pools
- Decks
- Additional stories



Structural Footprint Change:

- Carport
- Patio/ Porch
- Additional Living Space



Leverage voice analytics to determine if Insurance Professionals (IPs) are meeting member's needs and delivering need-based product/service options

In Development

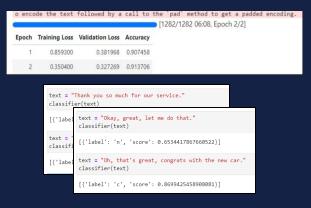
Analytical Solution

- Series of models that identify when IPs achieve the pillars of Mission Accomplishment on voice calls
 - Champion for the member by handling requested business
 - Provide empathetic and exceptional service
 - Deliver needs-based options and ask for the business

Benefits

- Ability to evaluate calls and determine how we are serving members – increased efficiency and coverage since managers will not need to listen to calls
- Data can be used to identify areas for improvement which can be incorporated into coaching and training – increasing IP performance
- Positive improvements to member satisfaction scores

Sentence Transformer – converting language into numbers



e-label	conversation_id	pretty_transcript	label	score
	ac35b1d47fa04162	I'm, so sorry to hear about that Sir is everybody okay.	s	0.791589
	23aba25f58d0475b	Ram so sorry to hear that I'll be happy to get you transferred over to roadside assistance I do have to pull your.	s	0.789897
	4ff1fe28d7d145aa	Oh, my gosh I'm so sorry.	\$	0.793214
	9e8c0c01217e4f14	Oh My God. This is so terrible I had it here saved.	5	0.716318
	095a40d348c2498a	Oh, My Gosh, Oh that stinks it does it stinks.	s	0.74549
	8b465b070ba14eec	Oh, Ashley I'm, so sorry.	s	0.791256
	808a5372fbea450d	Okay, Alright, so you just went through that.	s	0.715035
	ba786e5c2e7c4876	And what I'm so sorry.	5	0.786342
	38f7ae12d7e34195	I'm sorry, she made a decision to do that gosh.	5	0.747715
	b948842f0163463c	Okay, well, that's unfortunate like you said uhm.	\$	0.577184
	1b888767b67540a5	Ah Okay, I'm so sorry.	s	0.794186
	2b6e98437b784221	So I'm. So sorry are you safe do you feel are you in a.	5	0.788876
	de9da3a194cd47bf	Oh I'm. So sorry go ahead please.	5	0.513325
	410fd8f4bb6640ec	No I'm sorry, you went through that because really they should've removed this vehicle.	5	0.774337
	72a9922a9b3e4ebd	[redacted-number] My God are you okay.	s	0.75587
	a5315fa659a04597	I'm, sorry to hear that happened to you like that.	s	0.793953
	2b1ffefddd324978	Alright, I'm, sorry about the news on that.	5	0.795218
	8b7b66748b9a4c2b	[redacted-number] and I'm, so sorry to hear that I know pneumonia is one of those it it's terrible.	5	0.791384
	3b68f69e9b7b4200	Sorry that you're dealing with this totaled vehicle I hope that it all gets resolved for you soon okay.	s	0.785467
	2b6e98437b784221	Okay, Alright, let me get you right over okay, I'm, so sorry about that and be careful.	s	0.503282
	bdc6376c00c94c9d	I mean, I'm so sorry.	s	0.79145
	29faaa8216db46c7	I'm, so sorry about that.	5	0.791859
	55728e605498486c	Oh this poor guy.	5	0.731863







Actuaries in an Al World: Four Futures



Impact to actuarial roles



Names borrowed from Jim Weiss' article in the Actuarial Review, a CAS bi-monthly





Arguments For:

- Routine automatable work:
 e.g.rate or reserve reviews, or predictive model refreshes
- 2. Al is better than actuaries at finding data
- 3. If we've created it, it's out there
- 4. AutoGPT: self-correcting, selflearning, can decompose tasks

What Does This Look Like?

First, AI starts taking mundane tasks, including data gathering.

Hallucination is solved for regulated industries.

Next, we train AI on more complex actuarial work, and it learns from the large body of existing work (4, 5).

Finally, a few actuaries may be needed for work that absolutely cannot be performed by AI.



Judgement Day



Arguments For:

Doomsday plus:

- AI is widespread and accelerating

 issues outpace our ability to
 manage them
- 2. Bias is an unsolved problem
- 3. Many regulations (e.g. Al Act) specify principles that are hard to put in practice

What Does This Look Like?

First, adoption and use are too attractive for us to go slowly

Regulations, testing, and auditing AI fall behind advancing sophistication.

Experts are needed to think about bias and consequences of AI.

Actuaries, who are used to making technical work meet regulations, evolve to fill the gap of managing and detecting issues.



Groundhog Day



Arguments For:

- 1. Good actuaries have originality, negotiation, and persuasion.
- 2. Employers are still investing in actuarial exams, training, and leadership development
- 3. Humans are unpredicatable
- 4. Al start-ups, even famous ones are outspending their revenues

What Does This Look Like?

Similar to DFA, the blockchain, interest in AI will peak, then it will reach a new norm.

Actuaries will integrate new technology, possibly via exams or via education, but AI will not displace large amounts of jobs.

In particular, the use of AI will become commonplace the same way we use current tools, but the applicability will be limited in scope.



Training Day



Arguments For:

- 1. "Doomsday lite": many jobs are doable via AI, but some aren't
- There are professionals who specialize in prompt engineering, RAGs, and tuning
- AI start-ups, even famous ones are outspending their revenues we could be reaching a peak
- 4. Again, humans are unpredicatable

What Does This Look Like?

Actuaries begin adopting AI to help with low-risk, automatable tasks.

As AI continues to develop, actuarial judgment will determine whether or not new methods are applicable.

Actuaries will continue to balancing stakeholder needs, including regulators.

In order to perform this job task, the actuary of the future will need to understand enough about AI and its power and limitations.



CAS Actuaries Leading the Profession in Data Science and the Application of AI/ML to Insurance





CAS credential examinations cover cutting-edge topics and modeling techniques that prepare students for the present and future, including:

- 1. Introduction to Credibility
- 2. Linear Mixed Models
- 3. Statistical Learning
- 4. Time Series with Constant Variance
- 5. Advanced Ratemaking
- 6. Predictive Analytics (PCPA) requirement



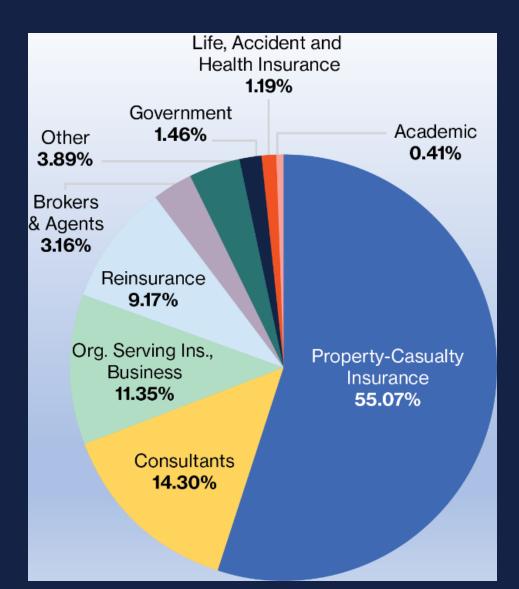
New Predictive Analytics Requirement

The latest addition to the credentialing pathway features a hands-on modeling project using a real-world general insurance business case:

- Incorporates machine learning algorithms and statistical models to help user identify trends, make accurate predictions, and optimize decision making processes.
- Provides candidates with access to a wide range of data sources and analytical tools, including historical data sets and advanced visualization tools.
- Enables candidates to gain a more comprehensive understanding of the factors that influence risk and to develop more effective strategies for managing it.



Where CAS Actuaries Work



Waymo Airbnb CyberCube Google DataRobot Extend CYENCE Doordash Just Root Insurance Hotels.com Lemonade SURE. Cambridge Mobile Telematics Hertz

Uber Expedia Zendrive Hippo HiRoad Mindstronglft Moter **Technologies** Slice Next Insurance Pie Tesla **OpenMRS** Jamstack GoHenry Backbase



How the CAS is preparing actuaries for future AI advancements

Continuing education & professional development opportunities

- 2024 Virtual Predictive Analytics Bootcamp
- The CAS Machine Learning Working Party
- Webinars including special topics in GLM, Reserving with Machine Learning, Ratemaking using Auction Theory, Latent Dirichlet Allocation (LDA) Topic Modeling in Python, and Large Language Models and Applications
- Online courses such as Introduction to Predictive Modeling, P&C Artificial Intelligence bundle
- AI Fast Track series coming this fall!

Research

- Call for Essays: Exploring the Intersection of Actuarial Science and Artificial Intelligence
- Call for Monographs Big Data, Machine Learning and Beyond



Join Us in Exploring Al

Are you passionate about artificial intelligence and eager to contribute?

The CAS and iCAS are looking for volunteers interested in joining future Al initiatives and discussions.

Your expertise can make a significant impact on our work and the broader field of actuarial science, data science and catastrophe risk management.

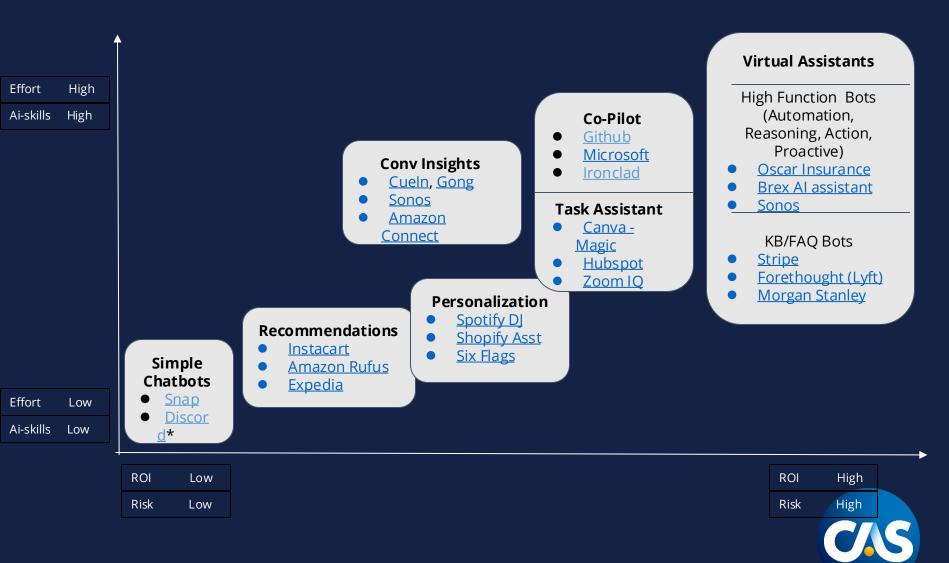
Interested volunteers, please email us at info@thecasinstitute.org to learn more and get involved.



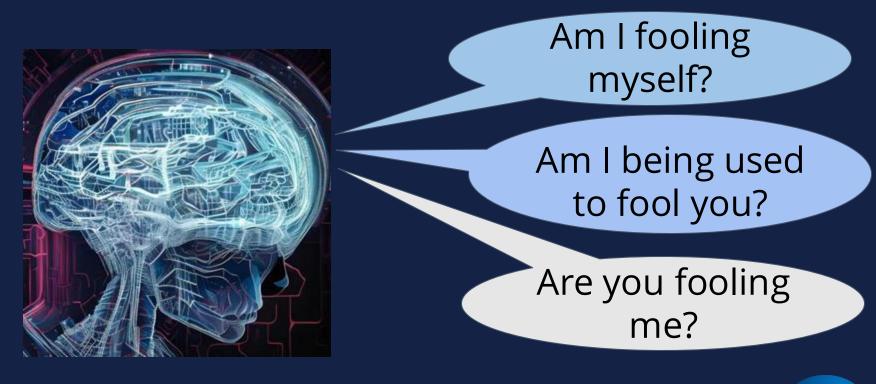
Ethical AI for Actuaries







Real World Concerns





How Can We Address Fairness?

Fairness Enforcement: Two ends of the spectrum

1. Omission



Fairness through "unawareness" - Model is ok, for example, if prohibited rating variables are not used.

Considerations

Prohibited variables are not explicitly used in the model

Adverse Consumer Outcomes may still exist with Omission intentionally or unintentionally through other variables or red-lining



How Can We Address Fairness?

Fairness Enforcement: Two ends of the spectrum

Considerations

The risk of "hidden" or unintended use of a variable is completely eliminated through Equal (average) Rates

However, this could create massive subsidies which have other adverse outcomes



Demographic Parity - For example, the overall average premium across the demographic must the same.



How Can We Address Fairness?

3. Equalized Odds¹

Prediction is equally accurate for all demographics.

Blue / Gold represent different demographics



A square represents a risky driver

The **red rectangle** represents the model's prediction of risky drivers.

Illustration of Equalized Odds: The model correctly identifies ½ of risky **Blue** and ½ of risky **Gold** drivers. Also, the model incorrectly captures ¼ of safe **Blue** and ¼ of safe **Gold** drivers and mis-identifies them as risky.

¹ References:

- 1. Equality of Opportunity in Supervised Learning, Hardt, Price, Srebro, 2016, NIPS
- 2. <u>A Reductions Approach to Fair Classification</u>, Agarwal, Beygelzimer, Dudık, Langford, Wallach, 2018, PMLR
- 3. METHODS FOR QUANTIFYING DISCRIMINATORY EFFECTS ON PROTECTED CLASSES IN INSURANCE Mosley, Weman, 2022, CAS



Is the model accurate, validation methods, testing?

Is the model predictive?

Cross-validation

Was there model "parsimony"?

- Complex models need more data than simple ones
- Testing, validation, and ongoing monitoring becomes harder for more complex models





Key Takeaways

- **Innovating Insurance:** Harnessing AI/ML to transform pricing, claims, and operations for a smarter future.
- **Real-World Impact:** Pioneering solutions like advancements leveraging aerial imagery, computer vision, and voice analytics
- **Empowering Actuaries:** Preparing for a dynamic future where AI augments our expertise and enhances decision-making.
- Leading with Data: CAS champions cutting-edge predictive analytics and advanced modeling techniques.
- **Future-Ready:** Embracing continuous learning and adaptation to thrive in the evolving landscape of insurance.
- Ethical Excellence: Committing to fairness, transparency, and integrity in all AI applications.

Fun Fact: Generative AI was used to brainstorm the outline for this presentation.







Thank you



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