



CHUBB®
CHUBB

Speaker Session at Megathon'25

11th October'2025



Who We Are – Chubb at a Glance

What We Do

Chubb is a world leader in insurance. With operations in 54 countries and territories, Chubb provides commercial and personal property and casualty insurance, personal accident and supplemental health insurance, reinsurance and life insurance to a diverse group of clients.



WHY CHUBB

Chubb at a glance

#1

Commercial lines insurer
in the U.S.

#1

Personal lines insurer for
high-net worth families in
the U.S.

54

Countries and territories
where we operate.

> 3 Million

Yearly new claims
handled by Chubb
globally.

Where We Are Today

 1100 offices worldwide	 43000+ employees	 200+ insurance products & services
 Listed on the New York Stock Exchange	 Listed 95 on Forbes Global 2000 List & 215 on Fortune 500	 Rated AA for financial strength from S&P and A++ from AM Best

2024 Annual Results

P&C Combined Ratio: 86.6%	Assets: \$246.5 bn
Gross Written Premiums: \$62 bn	Net Income: \$9.272 bn

Did You Know?

Interesting Facts About Chubb



Titanic

Chubb insured the Titanic's owner, the White Star Line for both the Titanic and its sister ship, the RMS Olympic



Hope Diamond

Chubb insured the Hope Diamond, a famous blue diamond weighing 45+ carats that is now on display at the Smithsonian Institution



Dinosaur Fossil

Chubb insured the fossilized remains of a Tyrannosaurus Rex, nicknamed 'Sue' in 1997, which was later sold at an auction



The Empire State Building

Chubb was one of the insurance companies that provided coverage for the construction of the iconic New York City skyscraper in the 1930s



Race Horse

Chubb insured the thoroughbred racehorse, Cigar in 1996. Cigar was one of the most successful racehorses of all time with 19 consecutive victories



Famous Musician's Guitar

Chubb insured Eric Clapton's Fender Stratocaster Guitar



Olympic Torch

Chubb provided insurance coverage for the Olympic torch used during the 2002 Winter Olympics in Salt Lake City, Utah

Our History

A Legacy of 230+ Years

1792

Investors meet at Independence Hall in Philadelphia to organize the Insurance Company of North America

1882

Thomas and Percy Chubb set up Chubb & Son in New York City

1967

The Chubb Corporation is formed. That year, Chubb acquires the Pacific Indemnity Companies.

1984

Chubb is listed on the New York Stock Exchange

1999

Chubb acquires Cigna P&C, thus entering 50 countries and transforming its business from specialty catastrophic risk to diversified P&C

2016

ACE and Chubb become one and adopt the Chubb name

Our Presence in India



The Chubb Engineering Center in India is part of a global network of CECs driving operational excellence through **engineering, analytics and automation capabilities**



2900+ headcount with a **young work force** that's a mix of Gen Z and Millennials



Global company with a **start-up mind set**, culture of can-do attitude and collaboration



Excellent **referral hiring** of 33% and **diversity hiring** at 33%

When you join Chubb, you will be part of an exciting journey of digital transformation, cutting edge technology, and value creation in the overall Chubb global ecosystem

Life at Chubb India

CHUBB®

In India, we are 2500+ strong



Hyderabad



Bhubaneswar



Bengaluru



Certified as a Great Place to Work® three years in succession!

Select Recognitions

- EY GCC Awards 2023: Best GCC Talent Strategy of the Year Finalist
- Rising 2024 Tech Summit: Best Firms for Diversity & Inclusion
- 3AI ACME Awards 2023: Top GCCs to Work for in AI & Analytics

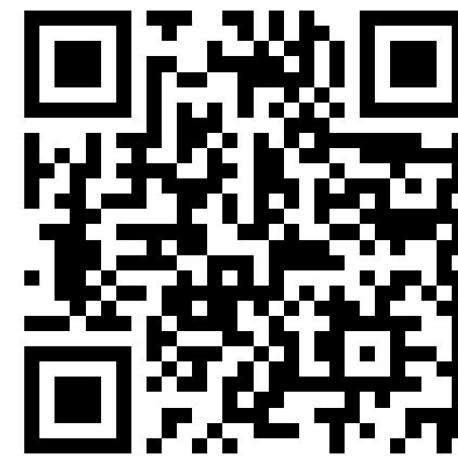
- 3AI's GenAI Max Conclave: Inspiring Leaders in AI and Analytics Awards
- HYSEA: Outstanding contributions and volunteering appreciation award

CHUBB®

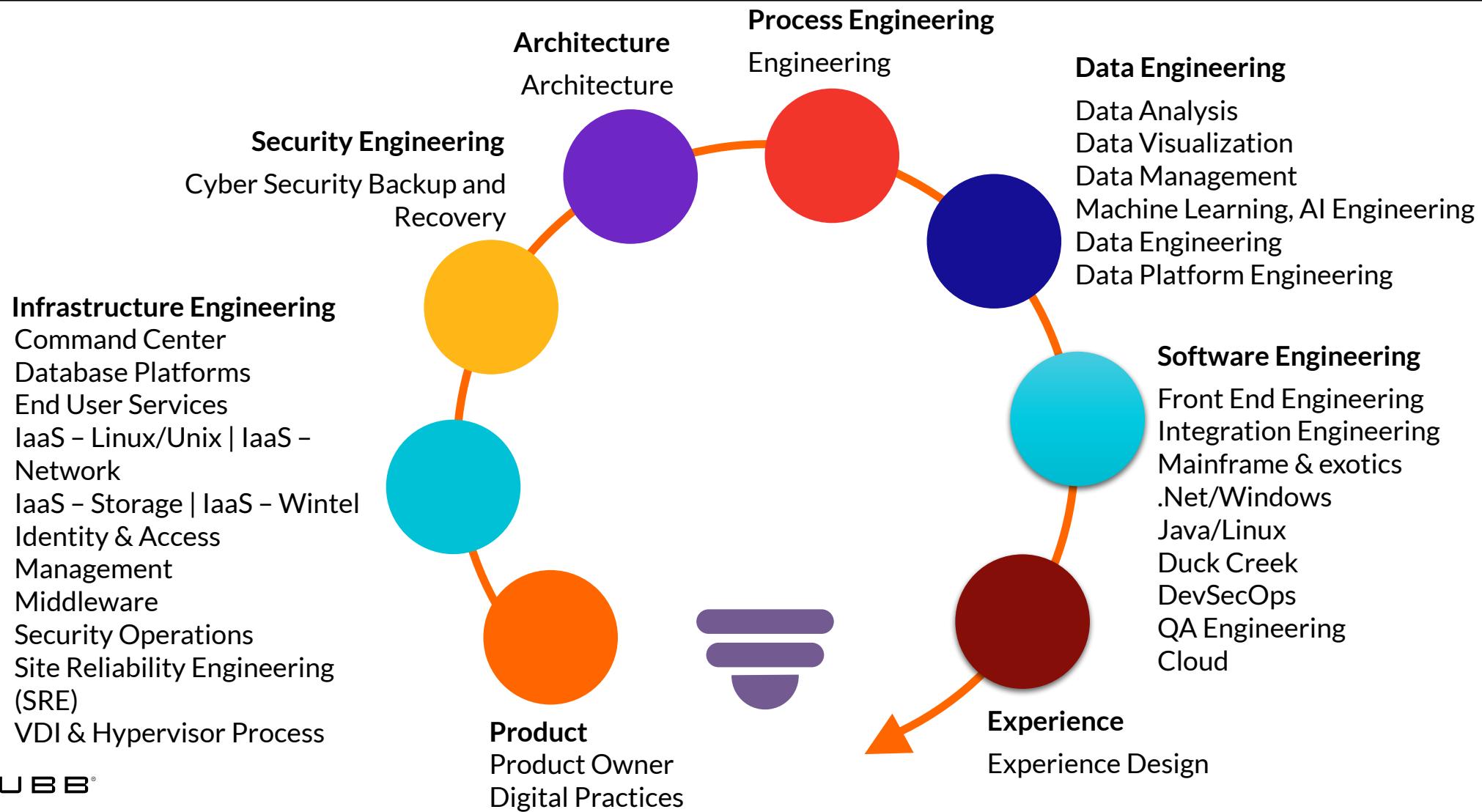
Let's Play !!

MEGATHON QUIZ

Join at
slido.com
#4640 565



Tech Tapestry: Unraveling the Stacks Powering Chubb



The Role of Technology in Insurance

- Let's explore how technology is transforming the insurance industry!
- Check out the articles below:

01 [Trends in tech shaping the future of insurance](#)

02 [Chubb Studio®: Technology to power our partnerships](#)

Chubb is harnessing the power of AI and data analytics to speed up claims processing and pinpoint risks with precision. For example, our cutting-edge digital tools can analyze real-time weather data to evaluate a homeowner's flood risk.

Have a peek into our Global Head, Chubb Engineering Centers, **Rohin Govindarajan's** passion for AI and technology in his [social media posts](#)

CASE STUDY

[How we built a best-in-class digital experience with DBS](#)



Chubb in News !!



[Chubb and Australian Open strengthen ties with expanded partnership](#)



[Healthy Paws Partners with PetSmart to Bring Insurance where Pet Parents Shop](#)



[Chubb to Acquire Liberty Mutual's P&C Insurance Businesses in Thailand and Vietnam](#)



[CHUBB RENEWS AS OFFICIAL INSURANCE SPONSOR OF THE US OPEN WITH MULTI-YEAR AGREEMENT](#)



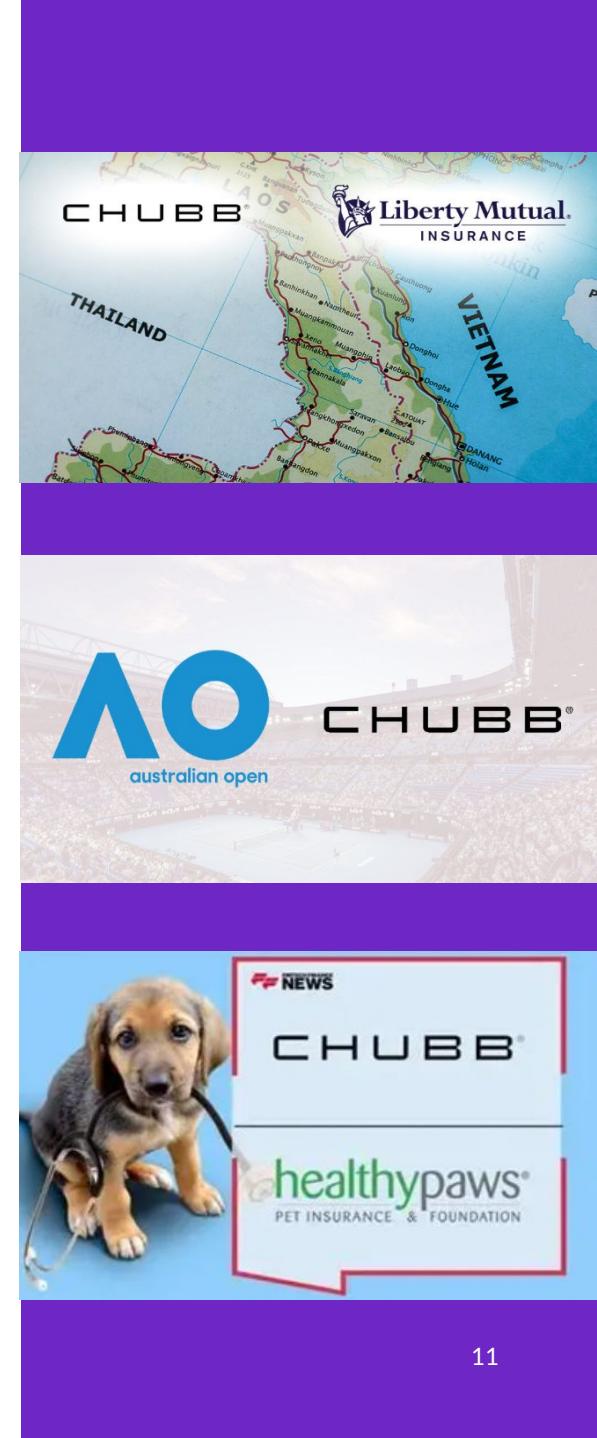
[Chubb to Acquire General Aviation Insurance Specialist in Australia](#)



[Chubb to Acquire Healthy Paws, a Leading Pet Insurance Provider](#)



<https://news.chubb.com/2024-01-23-StreamLabs-Expands-Product-Suite-with-Scout,-a-New-Smart-Water-Sensor>



Can you explain the scene here?



Hackathon Problem Statement: 1

AI-Powered Multimodal Claims Assessment & Fraud Intelligence System



Context/Background: Modern insurance claim workflows face two major challenges: Subjective visual assessment of damages — leading to inconsistency in severity and cost estimation and rising fraud sophistication — image reuse, tampering, or mismatched claim narratives.



The Challenge:

To design an AI system that can “think like an assessor”, combining image understanding, contextual reasoning, and basic fraud intelligence — all without access to private insurer data.

Understanding the Problem and Scope

Use Cases: Participants are expected to build an AI powered prototype by keeping in mind -
Image damage detection + severity + cost band

Recommended Tools/Technologies but not limited to

YOLOv8,
EfficientNet, any
pretrained CNN

OpenCV

Azure Vision APIs

Python

Guidelines

Data and Resources: Use publicly available datasets like:

- Kaggle: Car Damage Detection & Severity Dataset (<https://www.kaggle.com/datasets/anujms/car-damage-detection>)
- Part Identification: CompCars [http://mmlab.ie.cuhk.edu.hk/datasets/comp_cars/]
- Image Tampering: CASIA2 Image Tampering Dataset [<https://www.kaggle.com/datasets/sophatvathana/casia-dataset>]
- Optional Text Claims: Participants can generate synthetic text inputs (e.g., “front left fender damaged”).

Technical and Implementation Details

- Technology Constraints: No internal or proprietary data may be used. Teams can use any public datasets, pretrained models, or open APIs
- Focus on conceptual completeness, not production-grade accuracy.
- Create Modular design with separate components for vision, text, and fraud detection.

Explainability:

- Encourage Grad-CAM / saliency maps (“why this region contributed to prediction.”)

Guidelines contd...

Output	Damage type & severity Fraud confidence score Final Claim Confidence Score (aggregated) Deployment: Local/Colab/AWS/GCP free tier acceptable.
Edge Cases to Test	Blurry or shadowed images. Multiple damage areas. Same image reused with different text claims. AI-generated or altered images.

Expectations and Deliverables

Format

- Both a working prototype (minimum: web app or notebook demo) + a short presentation (pitch deck).

Prototype Deliverable

- Web app (Streamlit/Gradio/Flask) where user uploads an image and outputs severity level, possible damage type, cost band.
- API endpoint or notebook for backend logic.

Minimum Deliverable

- Standalone Jupyter Notebook with inference pipeline using public images.

Turnaround Expectation

- Should be able to return a result within seconds per image.

Accuracy Expectation

- Not benchmarked to production – instead: proof of concept accuracy >70% for classification tasks is acceptable.

Evaluation Criteria

Key criteria:

- Innovation & Creativity – how novel is the approach?
- Technical Implementation – how well is AI/ML applied (model choice, training, reasoning)?
- Accuracy & Reliability – correctness of severity classification, detection of tampering, cost estimation approach.
- Usability & Explainability – clear UI, ability to explain predictions (not just “black box”).
- Scalability & Realism – can the solution scale, and would it make sense in real-world claims?

Judging Rubric:

- Innovation and System Design (20%)
- Technical Depth (25%)
- Accuracy and Robustness (25%)
- Usability and Explainability (15%)
- Presentation and Impact (15%)

Let's Play !!

Megathon 25 Quiz: AI-Powered Multimodal Claims
Assessment & Fraud Intelligence System

Join at

slido.com
#2730 342



Hackathon Problem Statement:2

AI Powered Churn Reasoning and Retention Insights for Auto Insurance



Context/Background: Customer churn is a major challenge for auto insurance companies, leading to significant revenue loss and high acquisition costs. By leveraging historical data such as demographics, income, location, policy tenure, and credit behaviour, machine learning models can identify customers likely to discontinue their policies. Explainable AI (XAI) provides the reasoning behind model outputs, enabling targeted, data-driven retention strategies.



The Challenge:

Develop a model that analyses historical customer, demographic, and policy data to predict the likelihood of churn, using explainability techniques such as SHAP, LIME, or Grad-CAM to highlight which factors most influenced each prediction. Design an interactive dashboard showing customer churn probabilities, reasons for churn, and regional trends that allows non-technical users to explore predictions and insights.

Understanding the Problem and Scope

Use Cases: Participants are expected to build a prototype that can:

- analyses historical customer, demographic, and policy data to predict the likelihood of churn.
- output a churn probability score for each customer, helping identify high-risk individuals before they leave.
- Develop an interactive dashboard showing customer churn probabilities, reasons for churn, and regional trends that allows non-technical users to explore predictions and insights.

Recommended Tools/Technologies but not limited to

Explainability techniques(e.g., SHAP, LIME, or Grad-CAM)

Interactive web app (e.g., Streamlit, Gradio, Flask, Angular)

Azure Vision APIs

Python

Guidelines

Dataset: Use publicly available datasets such as:

- No internal company data will be shared.
- Kaggle: [Auto Insurance churn analysis dataset](#)
- Participants may use publicly available pretrained models or libraries for explainability and visualization.

Technical and Implementation Details

- Train an ML model to compute a churn risk score for each customer.
- Ensure the solution is broken into distinct components (e.g., prediction, explainability, UI).
- Add simple explainability (e.g., feature importance, SHAP, or reason codes).
- The solution should handle large datasets efficiently.
- Ensure model inference runs within seconds.

Explainability:

- Encourage the use of SHAP/LIME to explain pricing decisions and highlight which behavioural factors most influenced the premium

Guidelines contd...

Output

A working prototype (web app, API, or notebook demo).

A model should output a churn probability score for each customer, helping identify high-risk individuals before they leave.

An interactive dashboard showing customer churn probabilities, reasons for churn, and regional trends that allows non-technical users to explore predictions and insights.

Edge Cases to Test

Handling noisy, incomplete, or imbalanced data.

Addressing edge cases, such as customers with very short or very long policy tenures.

Ensure explanation exists even when model abstains or confidence low and works consistently across diverse customer profiles.

Expectations and Deliverables

Format

- Both a working prototype (minimum: web app or notebook demo) + a short presentation (pitch deck).

Prototype Deliverable

- Web app (e.g., Streamlit/Gradio/Flask)

Minimum Deliverable

- Standalone Jupyter Notebook with inference pipeline using public or synthetic data

Turnaround Expectation

- The system should return results within seconds per customer record

Accuracy Expectation

- Not benchmarked to production – instead: proof of concept accuracy >70% for classification tasks is acceptable.

Evaluation Criteria

Key criteria:

- Innovation & Creativity – how novel is the approach?
- Technical Implementation – how well is AI/ML applied (model choice, training, reasoning)?
- Accuracy & Reliability – correctness of risk scoring, premium recommendations, and handling of edge cases.
- Usability & Explainability – clear UI, ability to explain predictions (not just “black box”).
- Scalability & Realism – can the solution scale, and would it make sense in real-world claims?

Judging Rubric:

- Innovation and Explainability (30%)
- Technical Approach (25%)
- Accuracy and Robustness (20%)
- User Experience (15%)
- Presentation and Impact(15%)

Let's Play !!

Megathon 25 Quiz: AI Powered Churn Reasoning and
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#2730 343



Let's hack!

Chubb. Insured.SM

Q&A