

# Claim Prediction for Flood Insurance

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# Table of Content

- Overview
- Data Source
- Key Insights
- Supervised ML Modelling
- Model Prediction Evaluation
- Conclusion

*Presentation Duration : 20 min*

# Overview

- There is a need to process flood insurance claims quickly and efficiently as volume of flood claims rises drastically in the event of natural disaster requiring fast and prudent processing.

<https://www.nytimes.com/2017/11/04/business/a-broke-and-broken-flood-insurance-program.html>

- Based on the patterns of the past claims amount paid, the geographic location of property, property condition, insurance coverage carried etc a reasonable prediction of the claim amount can be made.

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# Data Source: Nation Flood Insurance Policy Dataset

FEMA dataset that includes more than two million flood insurance claims records dating back to 1978.

## [FEMA Publishes NFIP Claims and Policy Data](#)

### FIMA NFIP Redacted Claims Data Set

#### [Federal Insurance & Mitigation Administration National Flood Insurance Program \(FIMA NFIP\) Redacted Claims Dataset](#)

Congress passed the National Flood Insurance Act (NFIA), 42 U.S.C. 4001 in 1968, creating the National Flood Insurance Program (NFIP) in order to reduce future flood losses through flood hazard identification, manage floodplain, and provide insurance protection. The Department of Housing and Urban Development (HUD) originally administered the NFIP, and Congress subsequently transferred the NFIP to FEMA upon its creation in 1979. FEMA and insurance companies participating in FEMA's Write Your Own (WYO) program offer NFIP insurance coverage for building structures as well as for contents and personal property with the building structures to eligible and insurable properties. The WYO program began in 1983 with NFIP operating under Part B of the NFIA and allows FEMA to authorize private insurance companies to issue the Standard Flood Insurance Policy (SFIP) as FEMA's fiduciary and fiscal agent. FEMA administers NFIP by ensuring insurance applications are processed properly; determining correct premiums; renewing, reforming, and cancelling insurance policies; transferring policies from the seller of the property to the purchaser of the property in certain circumstances; and processing insurance claims.



# FEMA

# Data Source: Key Information on Flood Claims

Claim Amount Paid  
on Building Damage

Claim Amount Paid  
on Content Damage

Coverage Amount on Building

Coverage Amount on Content

Time of Flood

Property Age

Start Date of Policy

Zip code

County

State

Flood Zone Code

Condominium (Y/N)

Elevated Building (Y/N)

Primary Residence (Y/N)

House of Worship (Y/N)

Agriculture Structure (Y/N)

Non Profit Structure (Y/N)

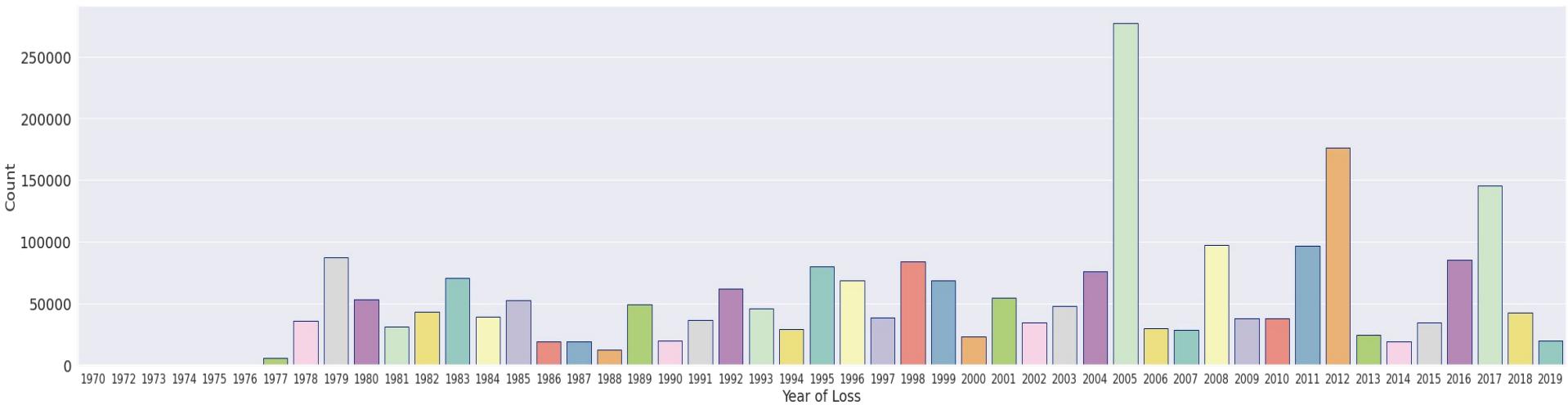
Small Business (Y/N)

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- Key Insights
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# Key Insights: Number of Property Loss with Year



- Recent three instances of heavy property losses occurred in the years of major hurricane events; Hurricane Katrina (2005), Hurricane Sandy (2012) and Hurricane Harvey (2017)

# Key Insights: Number of Property Loss with Year



**Seasonal boundaries**

**First system formed**

**Last system dissipated**  
January 6, 2006  
(record latest, tied with 1954)

**Strongest storm**

**Name** Wilma  
(Most intense hurricane in the Atlantic basin)

- Maximum winds** 185 mph (295 km/h)  
(1-minute sustained)

- Lowest pressure** 882 mbar (hPa; 26.05 inHg)

**Seasonal statistics**

**Total depressions** 31 (record high)

**Total storms** 28 (record high)

**Hurricanes** 15 (record high)

**Major hurricanes (Cat. 3+)** 7 (record high)

**Total fatalities** 3,912 total

**Total damage** \$171.755 billion (2005 USD)  
(Second-costliest tropical cyclone season on record)



**Seasonal boundaries**

**First system formed**

**Last system dissipated**  
October 29, 2012

**Strongest storm**

**Name** Sandy

- Maximum winds** 115 mph (185 km/h)  
(1-minute sustained)

- Lowest pressure** 940 mbar (hPa; 27.76 inHg)

**Seasonal statistics**

**Total depressions** 19

**Total storms** 19

**Hurricanes** 10

**Major hurricanes**

(Cat. 3+) 2

**Total fatalities** 355 total

**Total damage** ≥ \$72.32 billion (2012 USD)  
(Third-costliest tropical cyclone season on record)



**Seasonal boundaries**

**First system formed**

**Last system dissipated**  
November 9, 2017

**Strongest storm**

**Name** Maria<sup>[nb 1]</sup>

- Maximum winds** 175 mph (280 km/h)  
(1-minute sustained)

- Lowest pressure** 908 mbar (hPa; 26.81 inHg)

**Seasonal statistics**

**Total depressions** 18

**Total storms** 17

**Hurricanes** 10

**Major hurricanes**

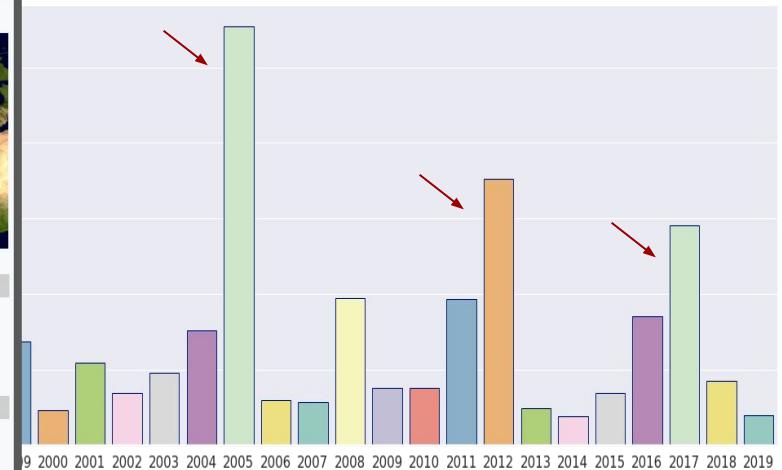
(Cat. 3+) 6

**Total fatalities** 3,364 total

**Total damage** ≥ \$294.92 billion (2017 USD)  
(Costliest tropical cyclone season on record)

**Related articles**

- Timeline of the 2017 Atlantic hurricane season
- 2017 Pacific hurricane season

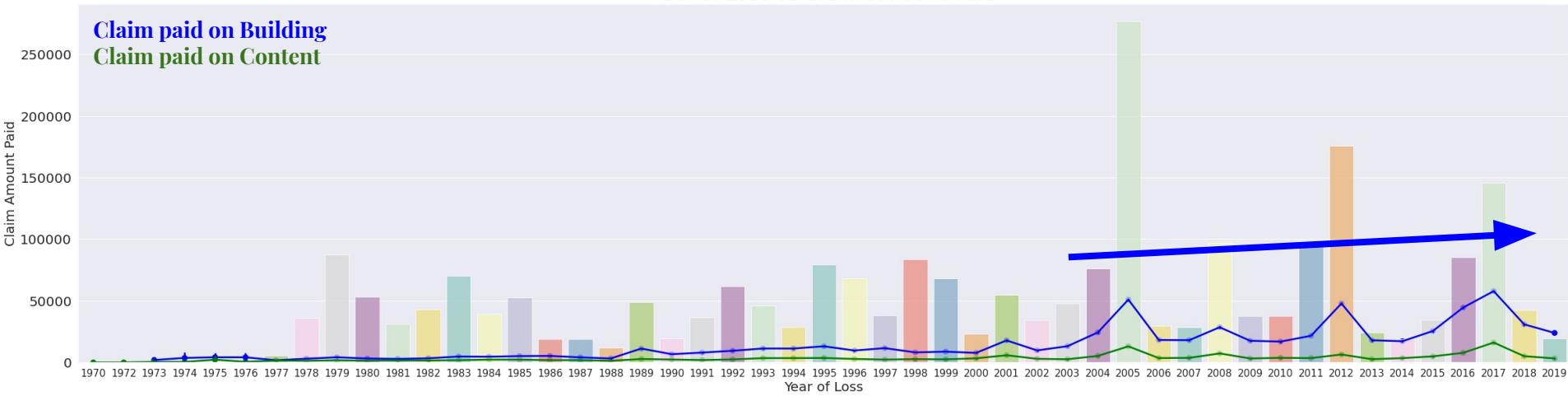


Top three costliest hurricane season

Source: Wikipedia

# Key Insights: Trend of Claim Amount

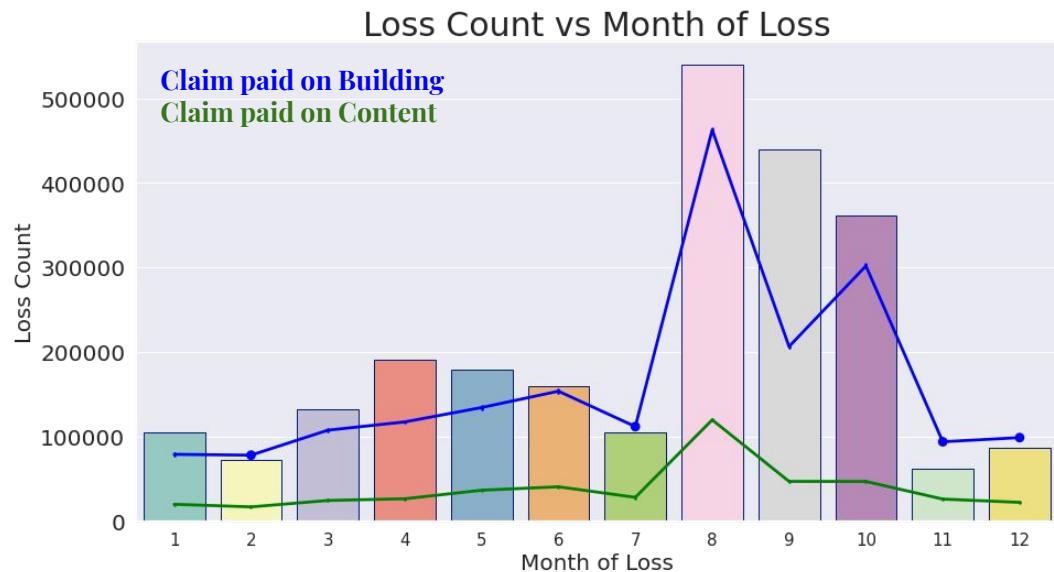
Year of Loss vs Claim Amount Paid



- Upward trend in claim amount paid through flood insurance on damaged properties
- Increase in paid claim amounts correlates with major hurricane seasons

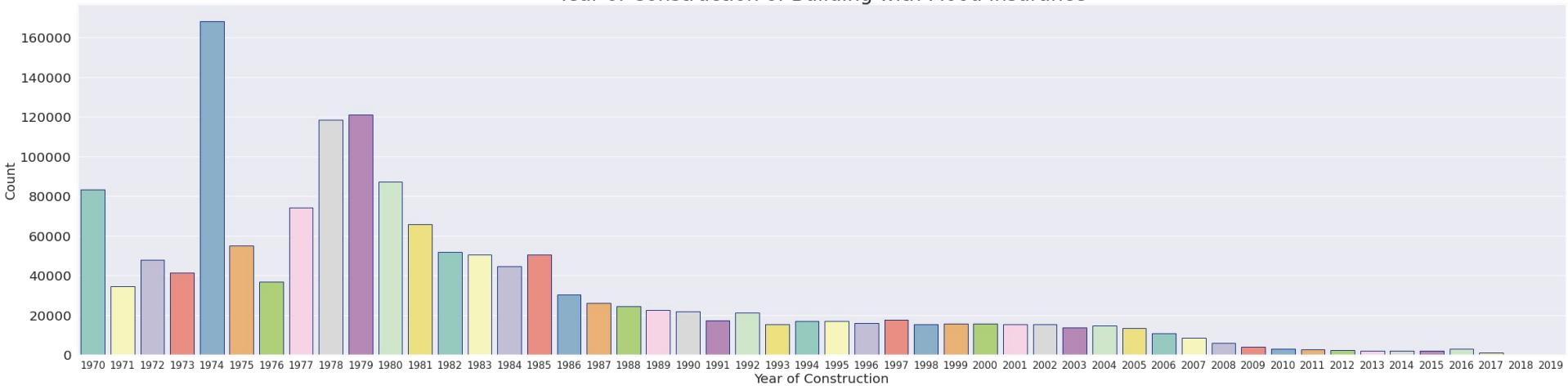
# Key Insights: Trend of Claim Amount (Month)

- Majority of claims come from the month of August, September and October
- Highly correlated with the timing of major hurricane events.



# Key Insights: Year of Construction

Year of Construction of Building with Flood insurance



- Major flood insurance claims made on older property (1975 – 1990)
- Downward trend on claims made on newer property hinting at better community planning to mitigate flood risks

# Key Insights: Year of Construction

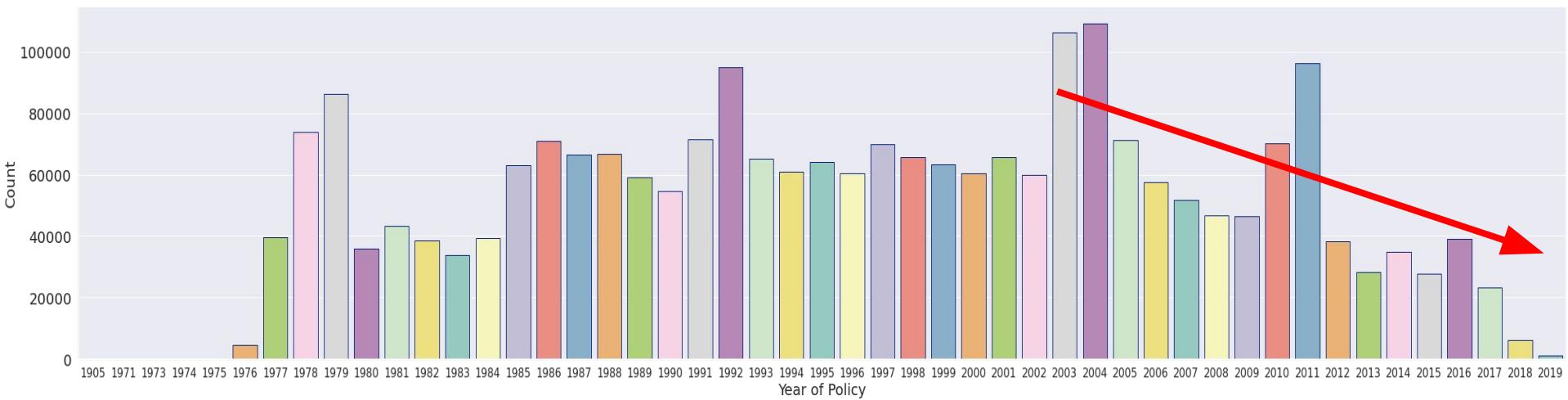


## Biggert–Waters Flood Insurance Reform Act of 2012 [edit]

The Biggert–Waters Flood Insurance Reform Act of 2012 was "designed to allow premiums to rise to reflect the true risk of living in high-flood areas."<sup>[1]</sup> The bill was supposed to deal with the "insolvency" of the National Flood Insurance Program by requiring the premiums to reflect real flood risks.<sup>[3]</sup> The result was a 10 fold increase in premiums.<sup>[1]</sup> At present, \$527 billion worth of property is in the coastal floodplain.<sup>[3]</sup> The federal government heavily underwrites the flood insurance rates for these areas. The law "ordered FEMA to stop subsidizing flood insurance for second homes and businesses, and for properties that had been swamped multiple times."<sup>[6]</sup> These changes were to occur gradually over the course of five years. FEMA was also instructed to do a study on the affordability of this process, a study which it has failed to complete.<sup>[6]</sup>

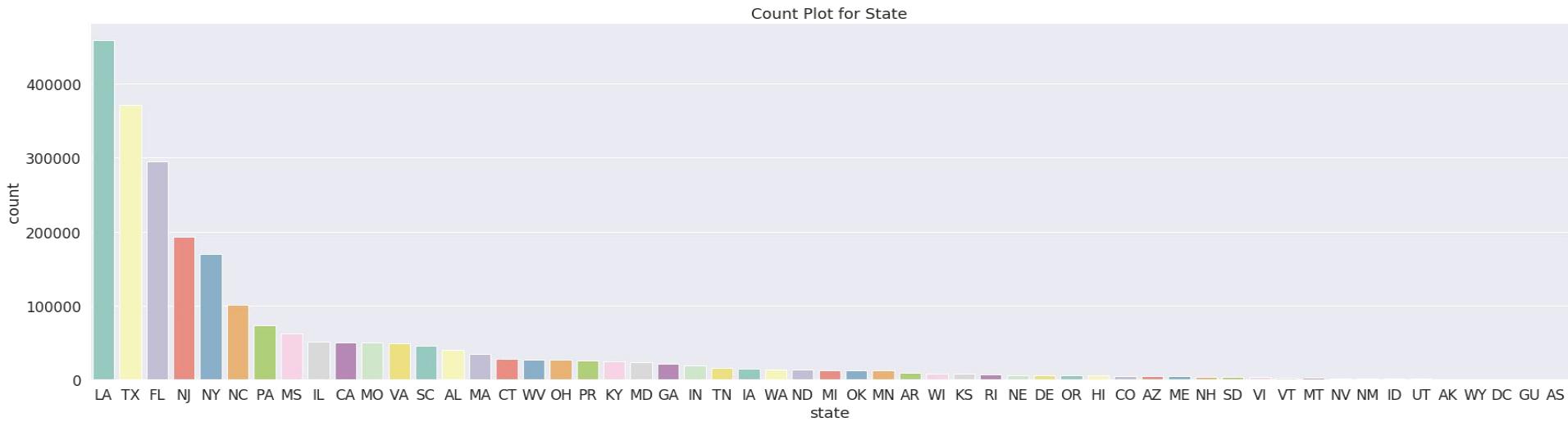
The [Homeowner Flood Insurance Affordability Act of 2014](#) (S. 1926<sup>[2]</sup>) was a [United States Congress bill](#) that would have delayed the increases in [flood insurance](#) premiums that were part of the Biggert–Waters Flood Insurance Reform Act of 2012.<sup>[1][2]</sup> The reforms from that law were meant to require flood insurance premiums to actually reflect the real risk of flooding, which led to an increase in premiums.<sup>[3]</sup> At the time of the bill, the National Flood Insurance Program was \$24 billion in debt.<sup>[3]</sup>

# Key Insights: Insurance Policy Years



- Downward trend in number of flood insurance policies carried through 50 years

# Key Insights: State-wise Number of Claims



- More number of flood claims coming from coastal states. Top six states include Louisiana, Texas, Florida, New Jersey, New York and North Carolina.

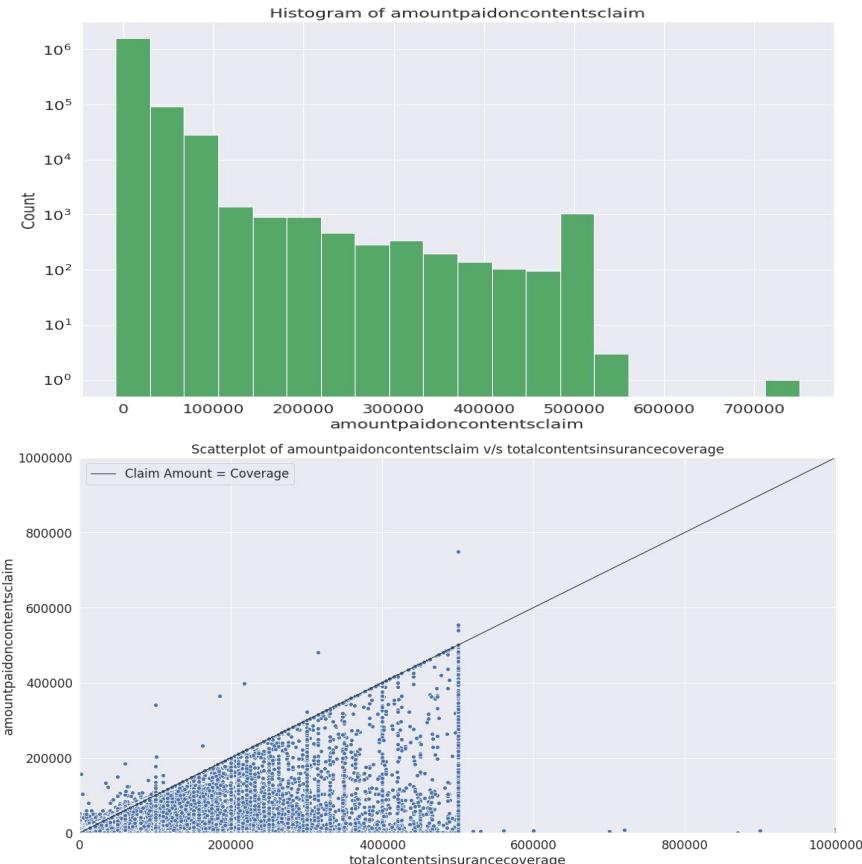
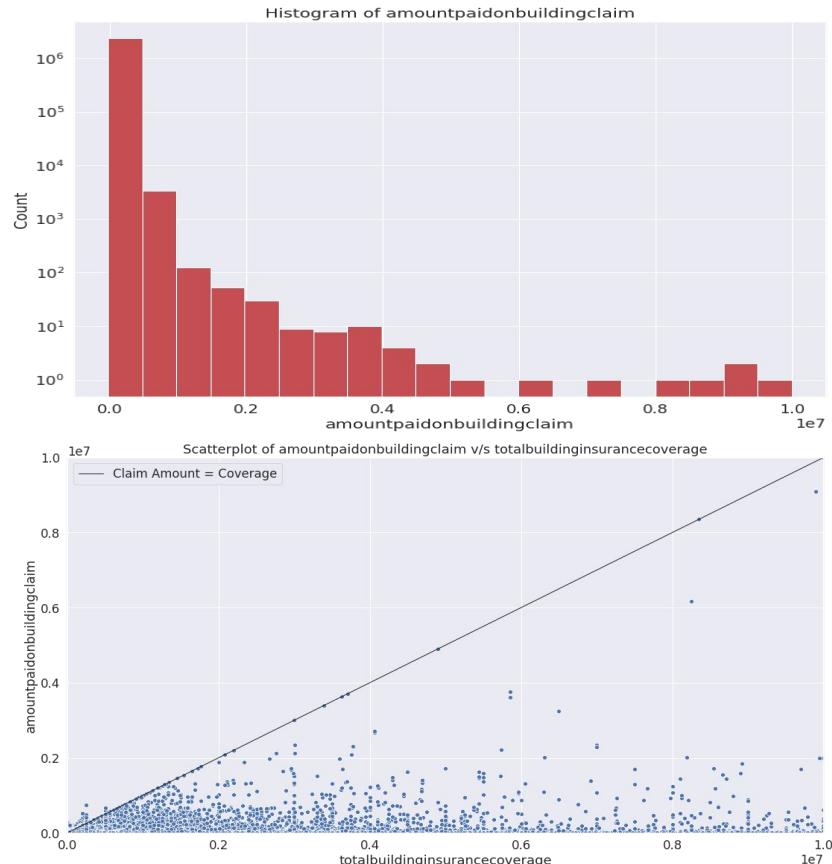
# Key Insights: State-wise Number of Claims

State vs Claim Amount Paid



- Variation among states in paid insurance amount.

# Coverage versus Claims paid



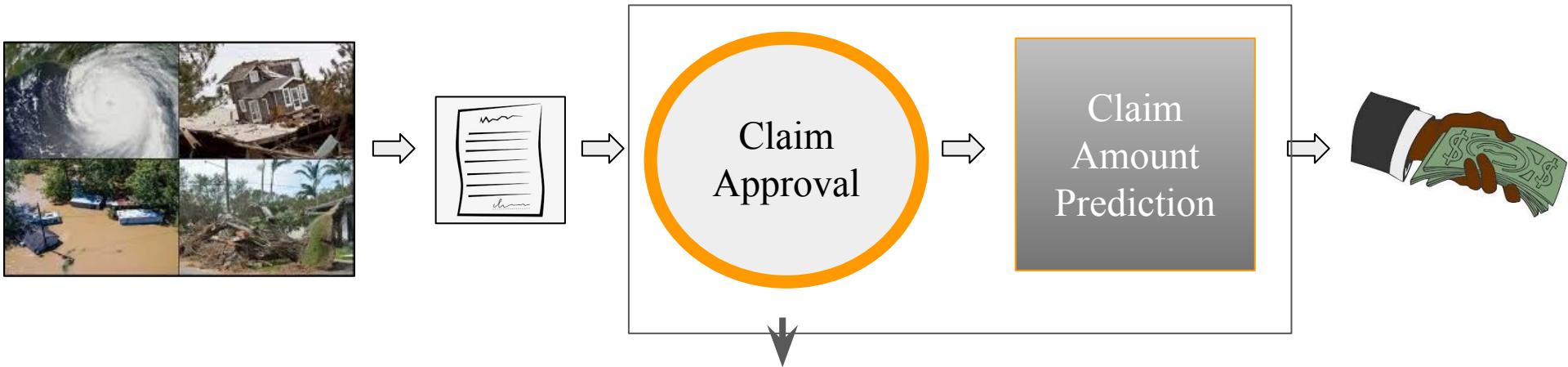
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- Overview
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# Supervised ML Modelling: 2 Step

- Build supervised modelling to predict claim amount using information from FEMA dataset such as year of loss event, property location, property condition etc



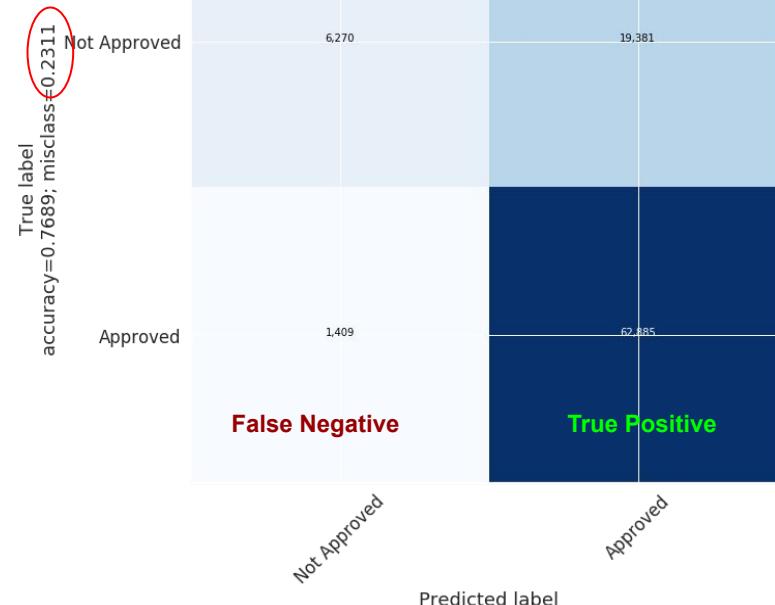
Is a claim acceptable for payout ??

Performed on Individual States

# Random Forest Classification

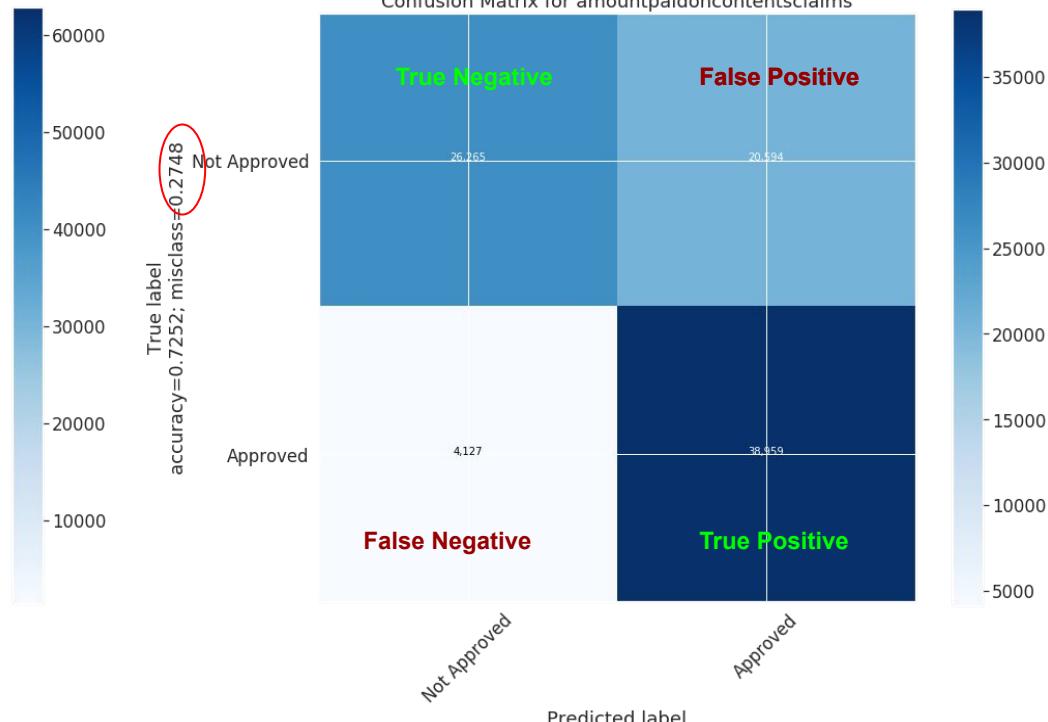
## Building Damage

Confusion Matrix for amountpaidonbuildingclaims



## Content Damage

Confusion Matrix for amountpaidoncontentsclaims



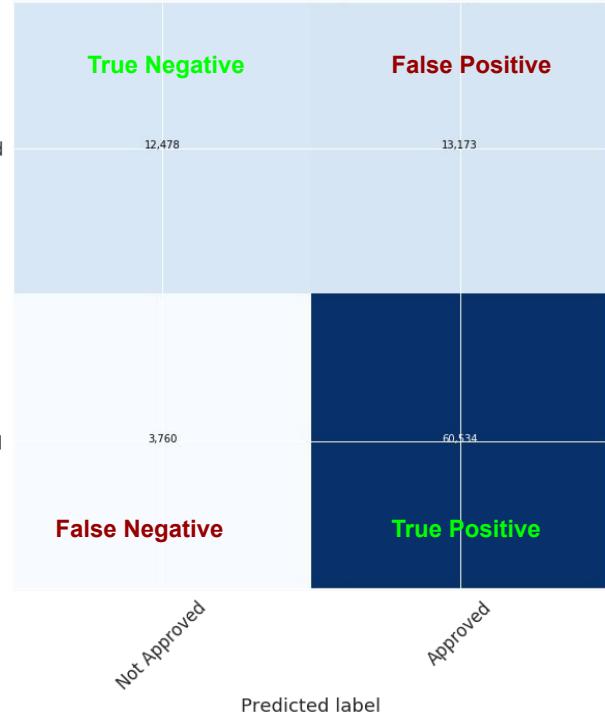
Performed on Individual States

# Gradient Boost Classification

## Building Damage

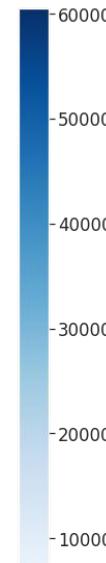
Confusion Matrix for amountpaidonbuildingclaims

True label  
accuracy=0.8117, missclass=0.1883



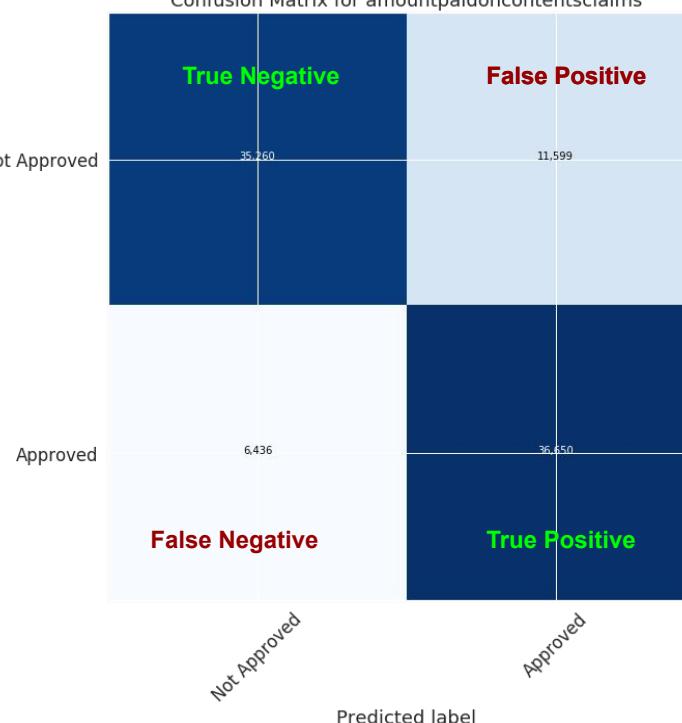
## Reduced False Positive

True label  
accuracy=0.7995, missclass=0.2005



## Content Damage

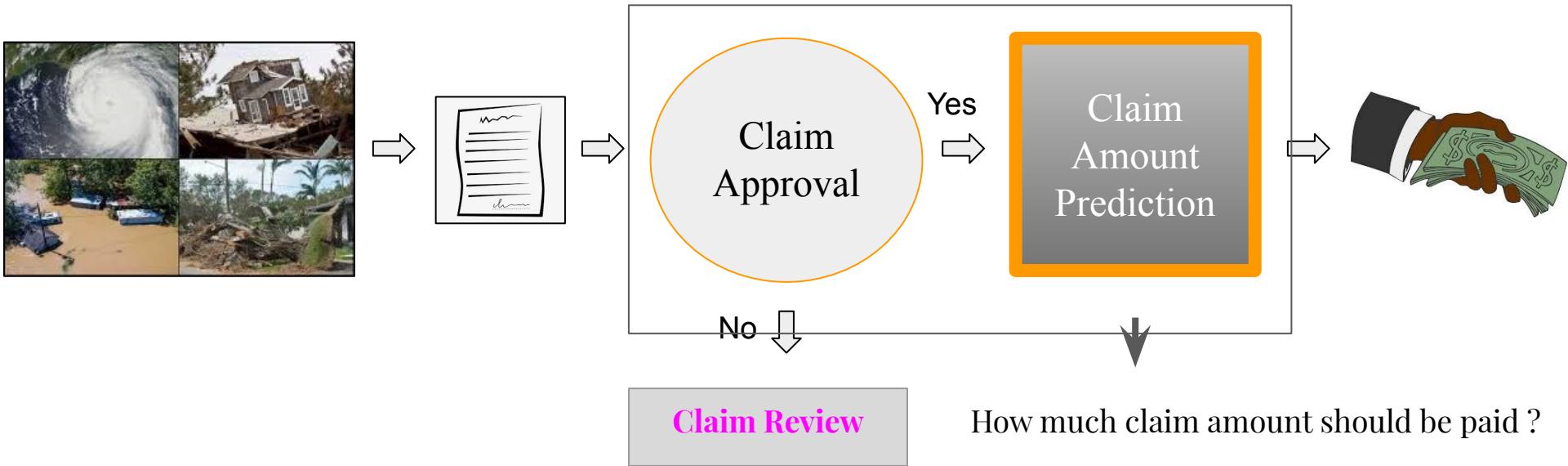
Confusion Matrix for amountpaidoncontentsclaims



Performed on Individual States

# Supervised ML Modelling: 2 Step

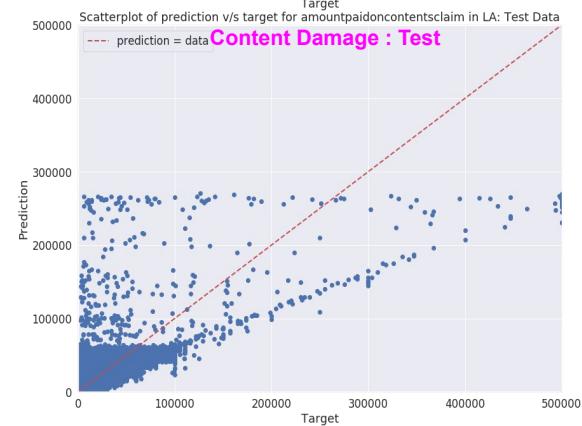
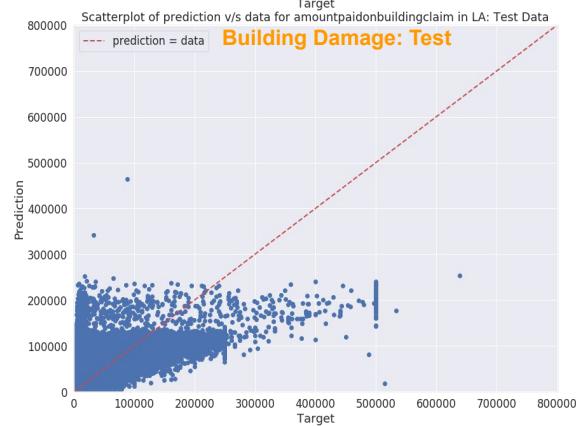
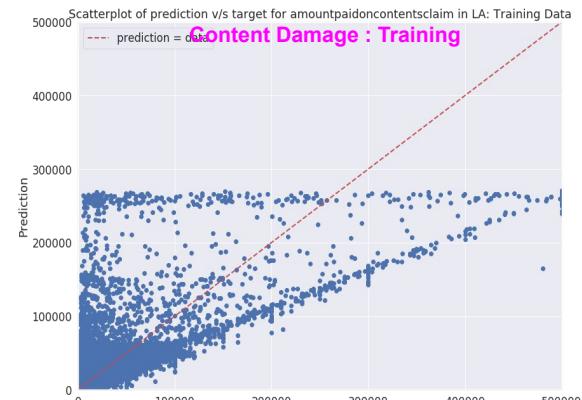
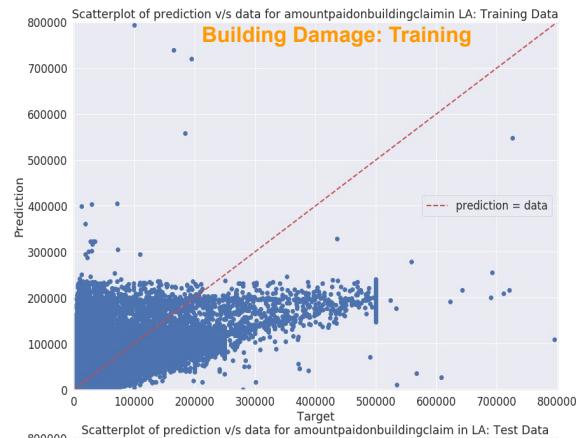
- Build supervised modelling to predict claim amount using information from FEMA dataset such as year of loss event, property location, property condition etc



Performed on Individual States

# Linear Regression Performance

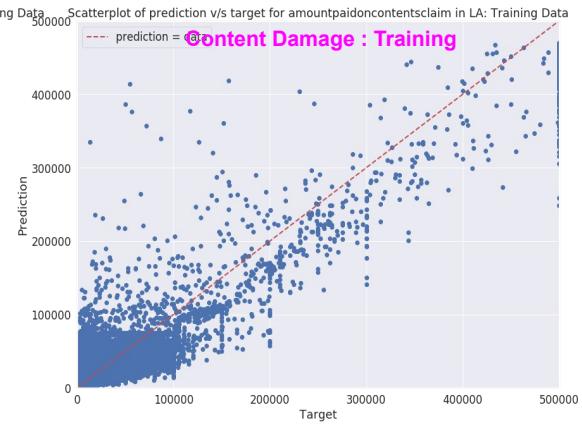
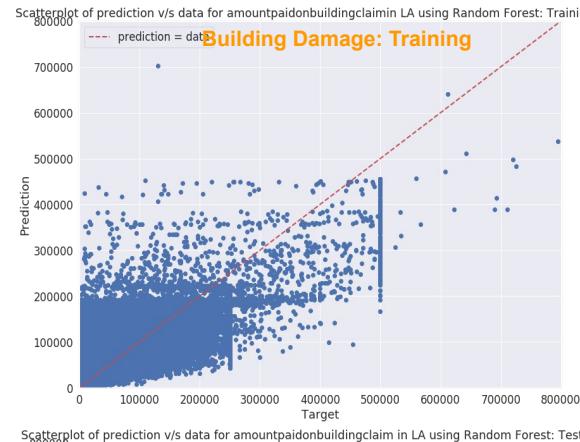
	Training	Test
R Square	0.677	0.631
RMS Error	40789.23	42323.43



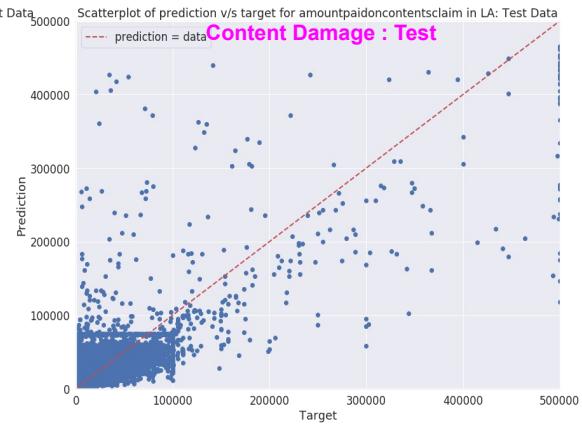
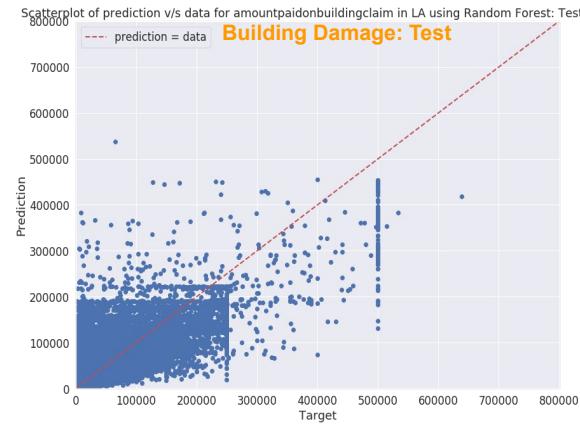
	Training	Test
R Square	0.672	0.612
RMS Error	17819.11	19919.19

# Random Forest Performance

	Training	Test
R Square	0.743	0.713
RMS Error	30032.47	32173.71

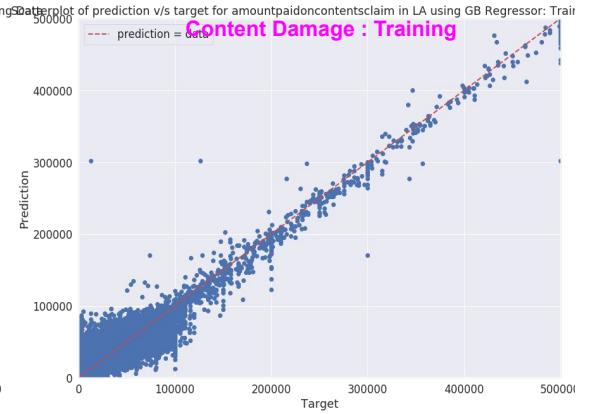
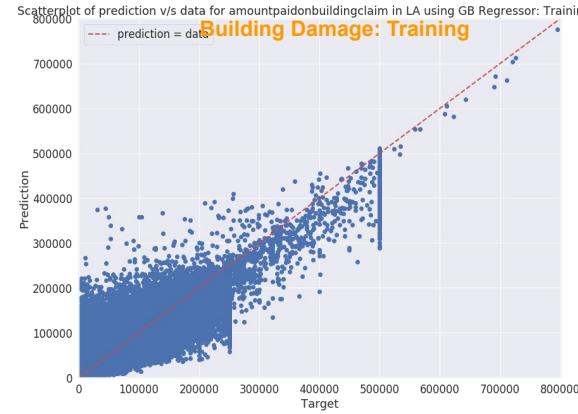


	Training	Test
R Square	0.772	0.658
RMS Error	12922.82	16739.56

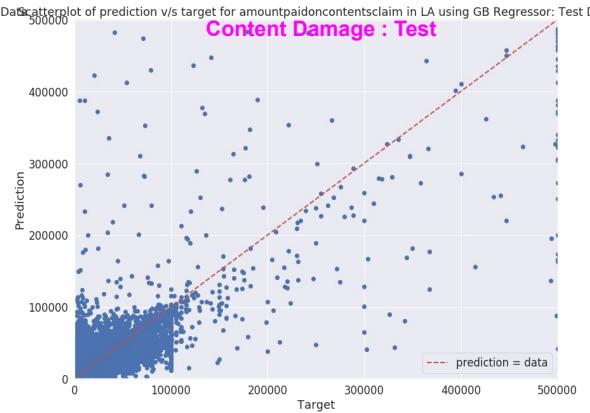
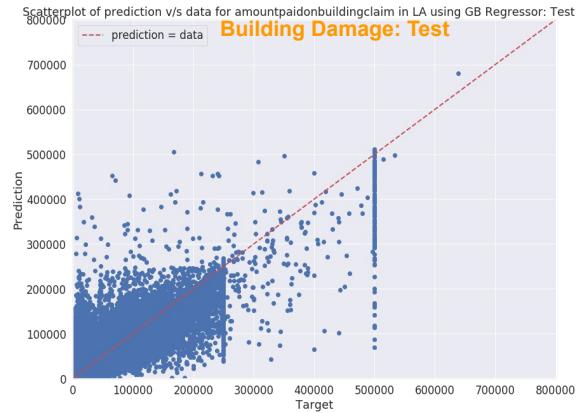


# Gradient Boosting Performance

	Training	Test
R Square	0.859	0.772
RMS Error	21899.32	28708.95



	Training	Test
R Square	0.898	0.688
RMS Error	8834.17	15988.67



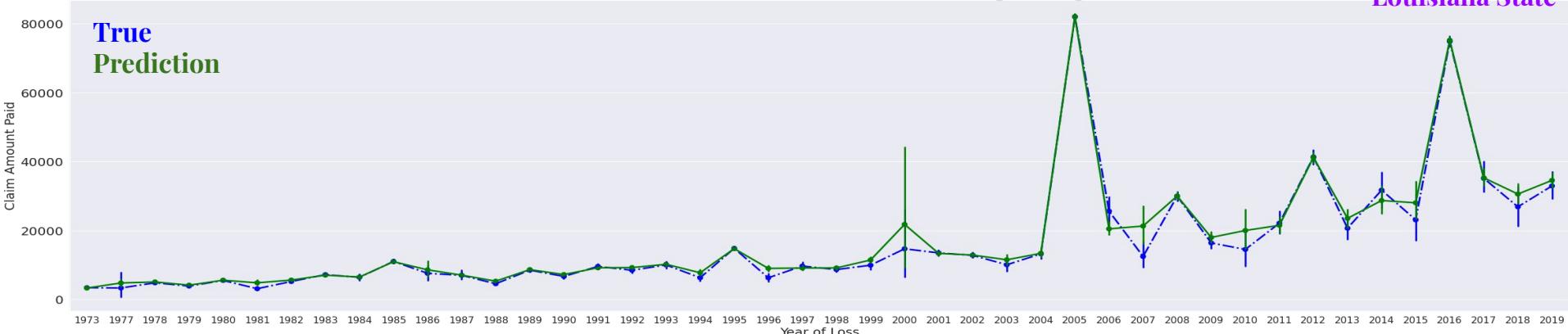
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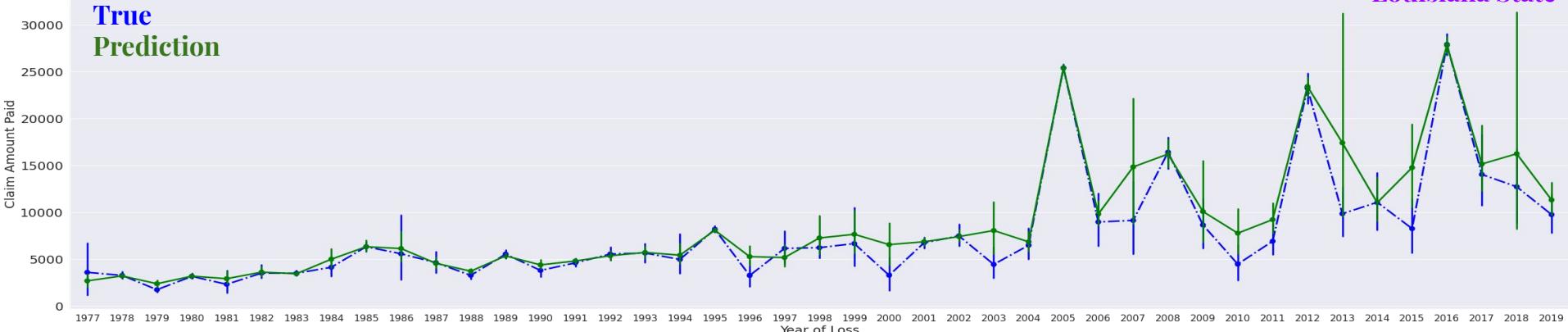
*Presentation Duration : 20 min*

# Model Evaluation (Test Dataset)

Year of Loss vs Claim Amount Paid on Building Damage



Year of Loss vs Claim Amount Paid on Content Damage



# Modelling: Limitation

- Model evaluation suggest overfitting issues with gradient boosting method despite better accuracy.
- Prediction accuracy in content claim is relatively low.

# Modelling: Future Improvements

- Strong correlation between natural disaster events (hurricane) and mean claim amount which is needed to be addressed in input.
- Include premium information of each flood insurance policy.

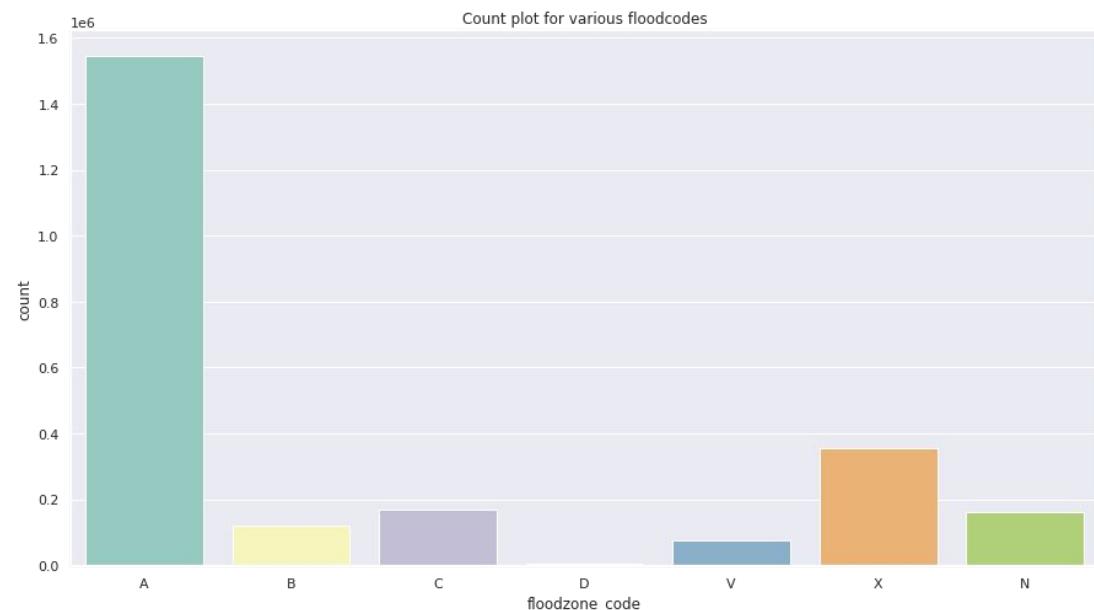
# Conclusion

- Building a prediction model is useful in the process of issuing claim amounts in the event of natural disaster (Hurricane, rainfall etcs) which require many working hours and resource in underwriting and claim evaluation.

Questions ?

# Appendix

# Categorical Variable: Flood-Zone Code



The flood-zone can be broadly classified:

1. Special Flood: A
2. Moderate Flood: B
3. Minimum Flood: C
4. Possible Flood: D
5. Velocity Flood: V
6. Moderate/Minimum Flood: X
7. Not specified: N

# Categorical Variable: Property Attributes

