



# Exploring the Potential of Image-based Deep Learning in Insurance

Luisa F. Polanía Cabrera



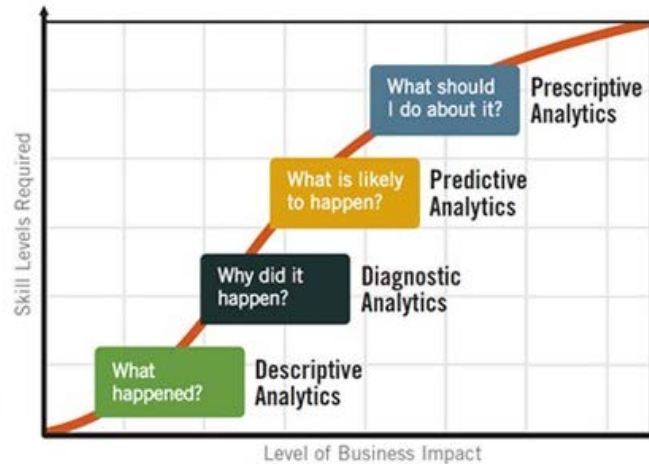
- Madison, Wisconsin – based American Family Insurance is the nation's third-largest mutual property/casualty insurance company .
- The company sells American Family-brand products, including auto, homeowners, life, business and farm/ranch insurance, through its agents in 19 states.
- American Family affiliates (The General, Homesite and AssureStart) also provide options for consumers who want to manage their insurance matters directly over the Internet or by phone.

# Strategic Data & Analytics

In order to accelerate advancement of big data analytics capabilities and business transformation, AmFam created a center of excellence for predictive analytics

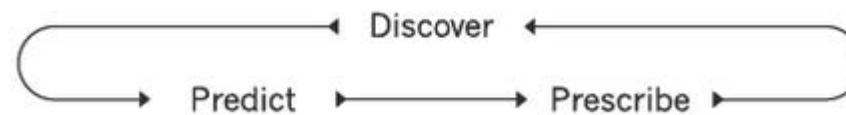
## Mission

To create competitive advantage and economic value by bringing together data innovation, advanced analytics and business acumen to optimize or transform our business models.



## Expertise

- Natural language processing
- Advanced modeling: Random forests, SVMs, etc.
- Probabilistic graphical models and Bayesian techniques
- Big data – parallel processing
- Deep learning



# Deep learning

Near-human  
accuracy speech  
recognition and  
image recognition  
systems



From machines that could  
not beat a serious Go  
player, to beating a world  
champion.



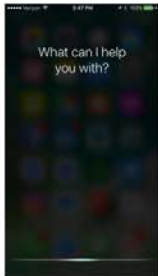
International Conference on Learning Representations



AMAZON'S ALEXA



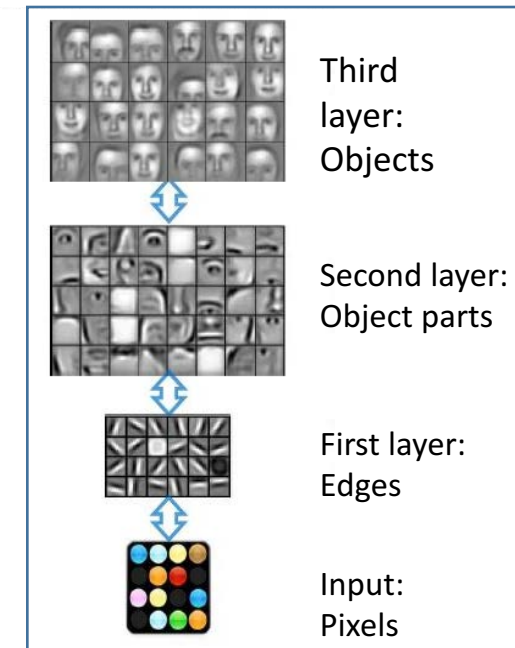
GOOGLE'S ASSISTANT



APPLE'S SIRI



MICROSOFT'S CORTANA





# Image Processing and Computer Vision in Insurance

*Roof Condition Prediction*



BMI Classification using Selfie Images



Automatic odometer reading from images



Blurring of human faces from drone images.



# Roof Condition Prediction

## Use case:

- There is a high likelihood that the roof age captured on the AmFam book of business, for property, is inaccurate (self-reported roof age), which results on risks not entirely assessed when insuring a property.
- There is a latent need to build predictive roof age/condition models.

Data from Millennium Inspection



## Roof Condition

- A: Roof has 16+ years of remaining life.
- B: Roof has 11-15 years of remaining life.
- C: Roof has 6-10 years of remaining life.
- D: Roof has 0-5 years of remaining life.



# Challenges

- Uncontrolled illumination.
- Scale variation.
- Capture angle variation.
- Roof material variation.
- Noise artifacts, shadows, occlusions.





# Roof Condition Prediction

Data from Millennium Inspection: In-the-wild data set (16M images)





# Filtering of Millennium Survey Images

Filtering with VGG: 1% of the total images are classified as roof images by VGG.



# Additional Filtering

- We collected 1000 labels to classify images between good and bad images for roof image analysis. NEXT<sup>(1)</sup> was employed to collect the labels.



Good Image

Bad Image

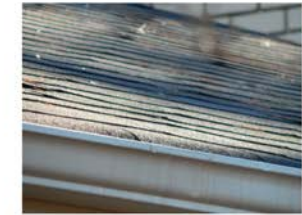
## Examples of bad quality images for roof analysis



Low-illuminated roof area



Roof area is almost completely occluded by branches and leaves



Blurry roof region



Overexposed image



Roof area has low resolution (roof details are not clearly visible)



Small roof region, low-illuminated roof



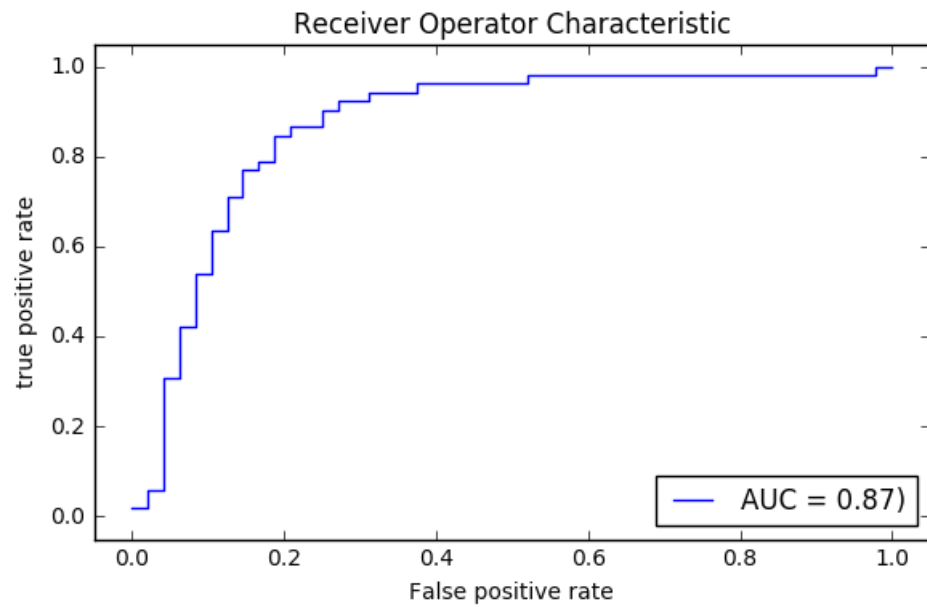
Roof covered by shadows



Non-roof images

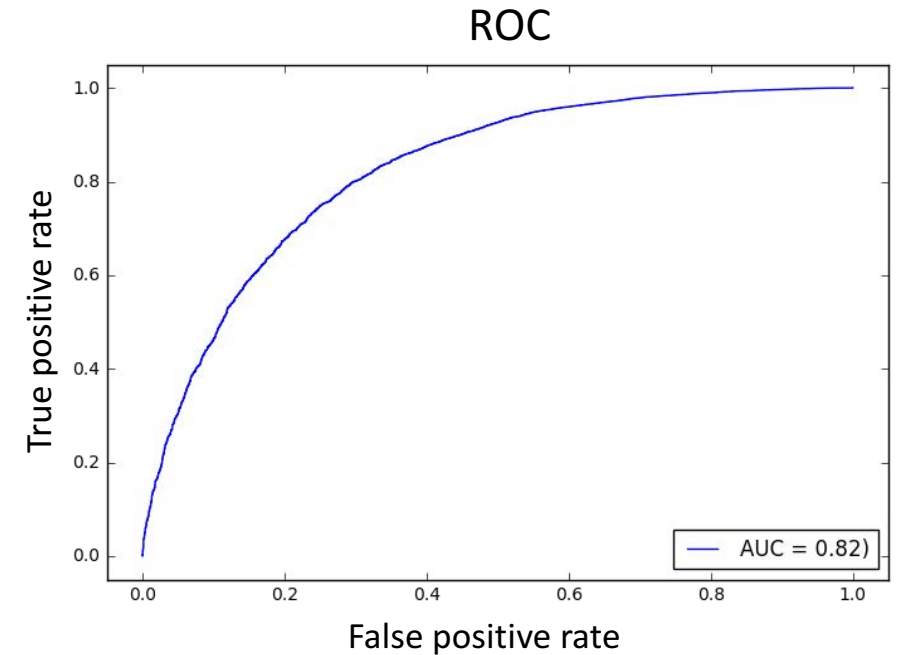
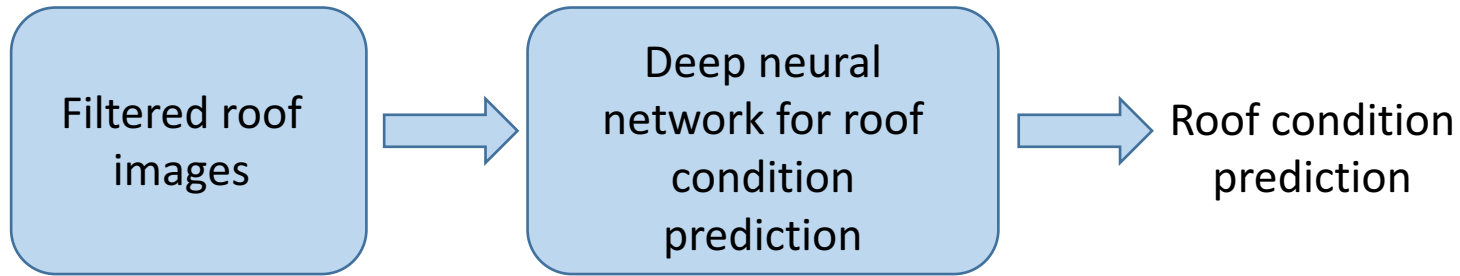


# Additional Filtering





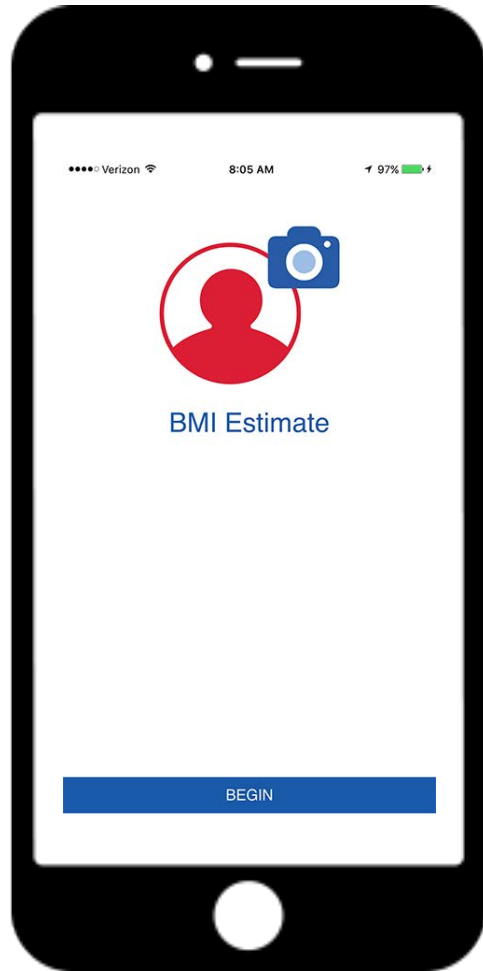
# Roof Condition Prediction



## Next steps:

- Detection of signs of deterioration: missing shingles, curling shingles, lifting shingles, cupping, etc.

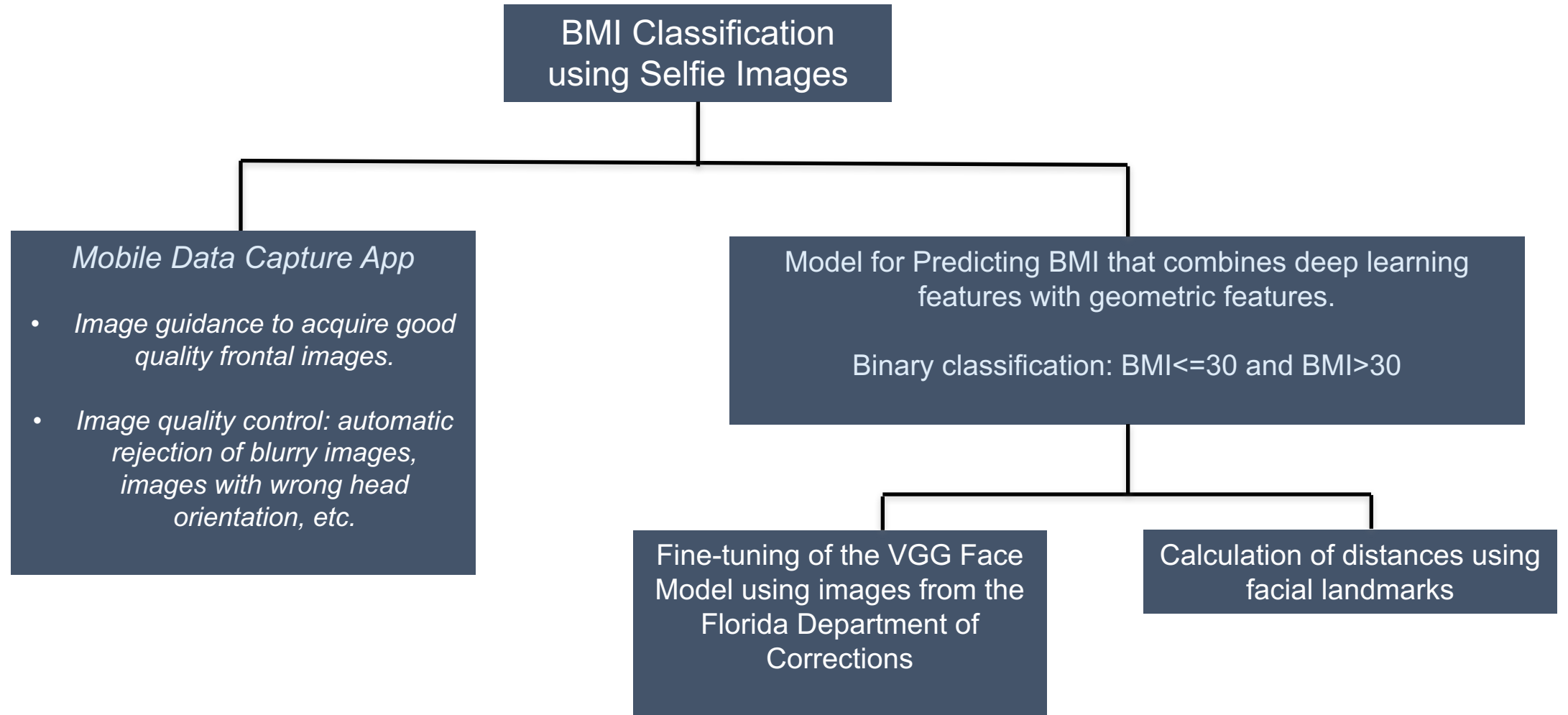
# BMI Classification from Selfie Images



## Use case:

- One of the factors that determine life insurance rates is body mass index (BMI).
- According to the Centers for Disease Control and Prevention, the correlation between a high BMI and the likelihood of developing dangerous health conditions is strong<sup>(1)</sup>.
- Self-reported BMI is inaccurate.
- Challenge: Verify customer BMI without a paramedical exam.

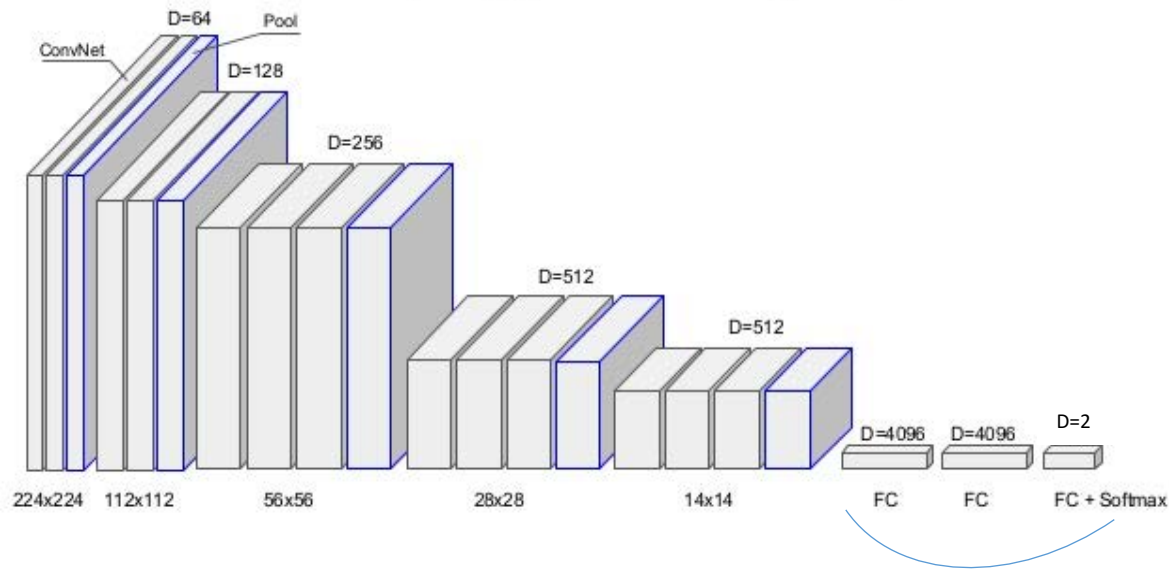
<sup>(1)</sup> [http://www.cdc.gov/healthyweight/assessing/bmi/adult\\_bmi/index.html](http://www.cdc.gov/healthyweight/assessing/bmi/adult_bmi/index.html) - July 11, 2014





# Fine-tuning of the VGG Face Model

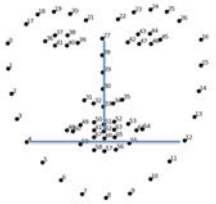
Classical CNN topology - VGGNet (2013)



- Thirteen convolutional layers and three fully-connected layers.
- Pre-trained model: 2.6M images.
- Dataset: 99096 images. BMI<30: 78%, BMI>=30: 22%.
- Training/Validation partitions: 70% / 30%

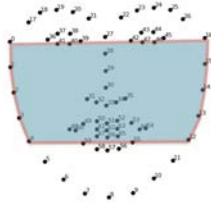
# Geometric features

1



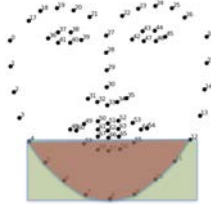
Eyes-to-Lips/Lower Width

4



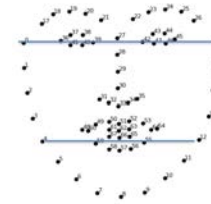
Perimeter Red/Area Blue

3



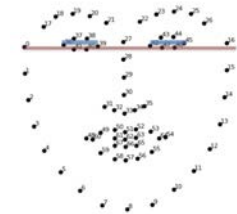
Area Green/Area Brown

0



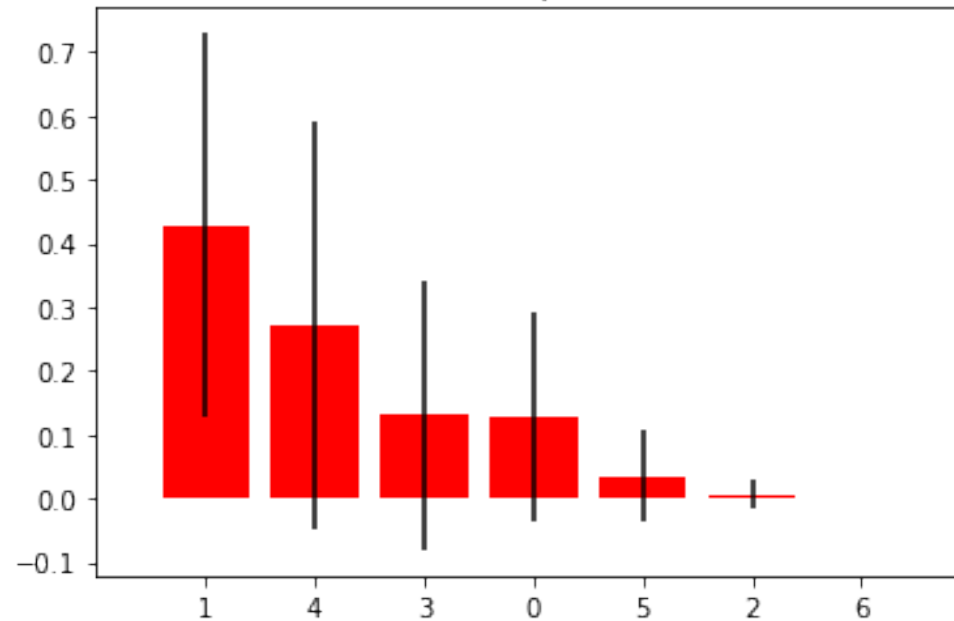
Upper Width/Lower Width

5

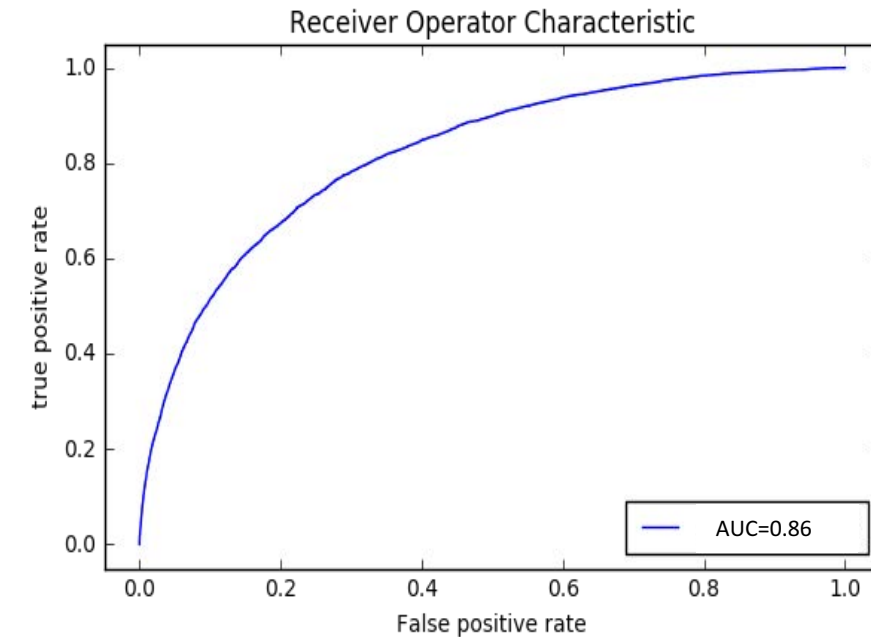
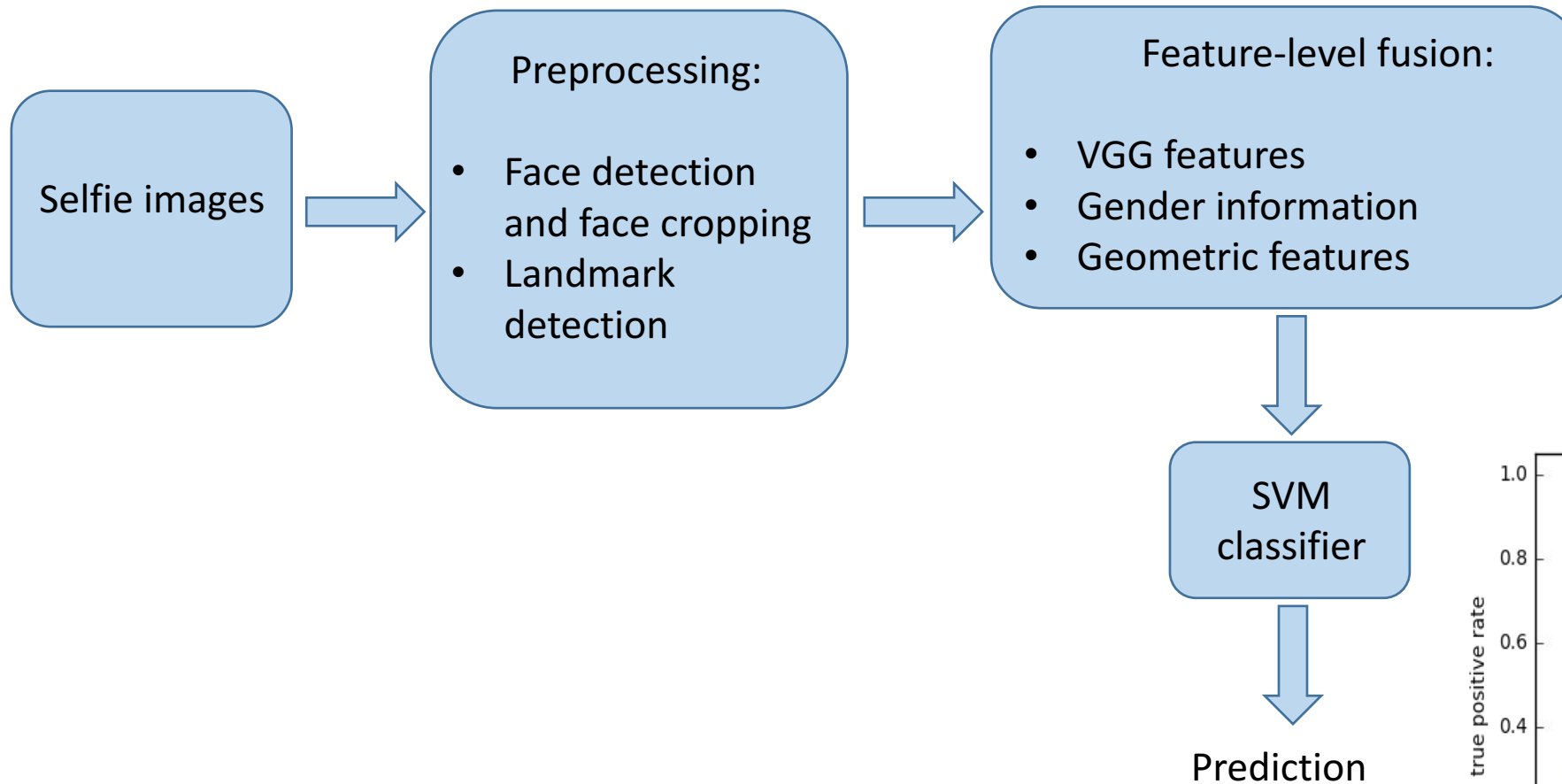


Average Eye Width/Upper Width

Feature importances



# BMI Classification





# Automatic Odometer Reading

## Use case:

- After a car accident, in addition to the pictures of the damaged areas, the adjuster also needs to take pictures of the license plate, VIN, and odometer. Then, they need to manually insert the information from these pictures into the claim system to receive a damage estimate.
- Automatic odometer reading would lead to time savings in the claim process and reduction in manual error annotations.



Output:  
151018

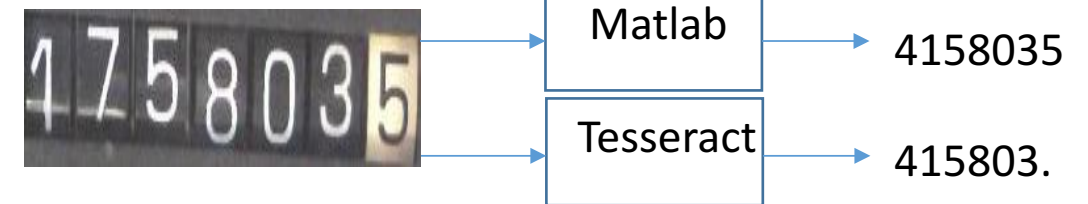
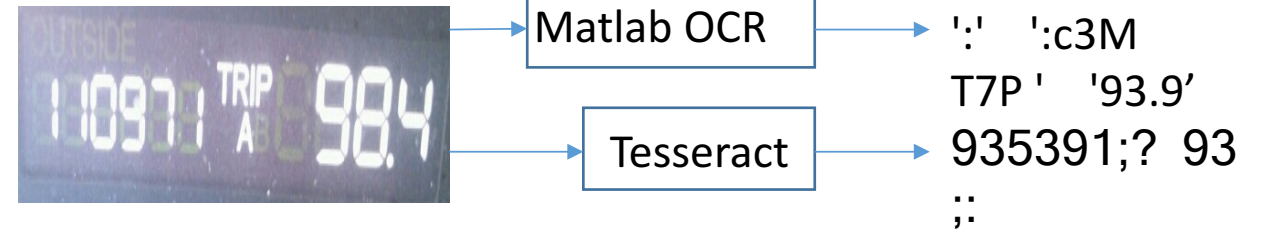
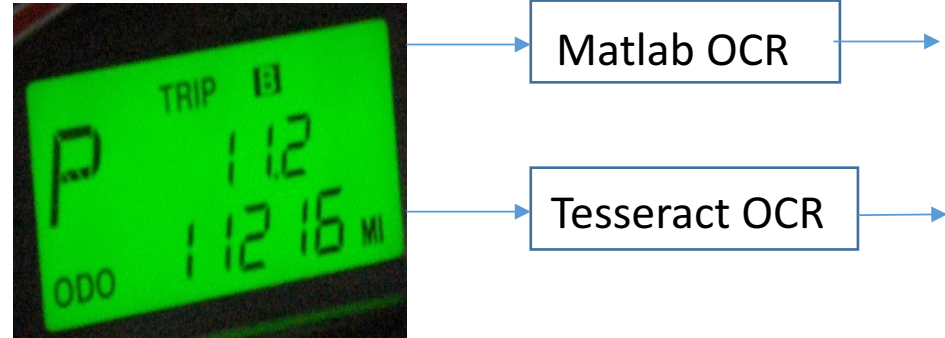
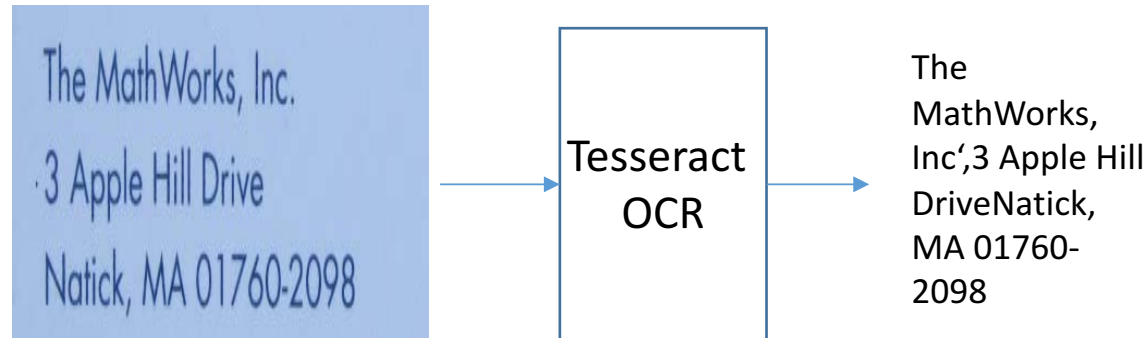
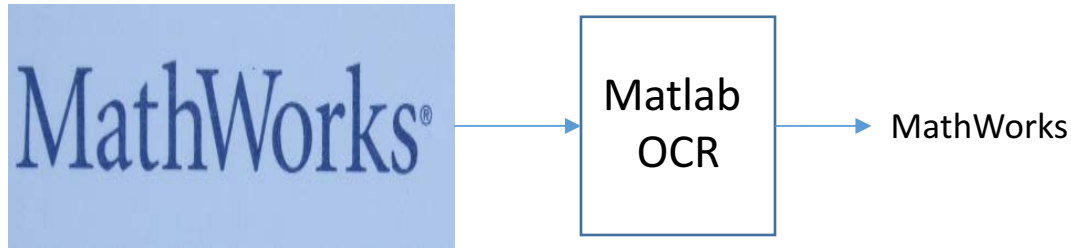
Dataset size: 6300 images

## Examples of images in the dataset:

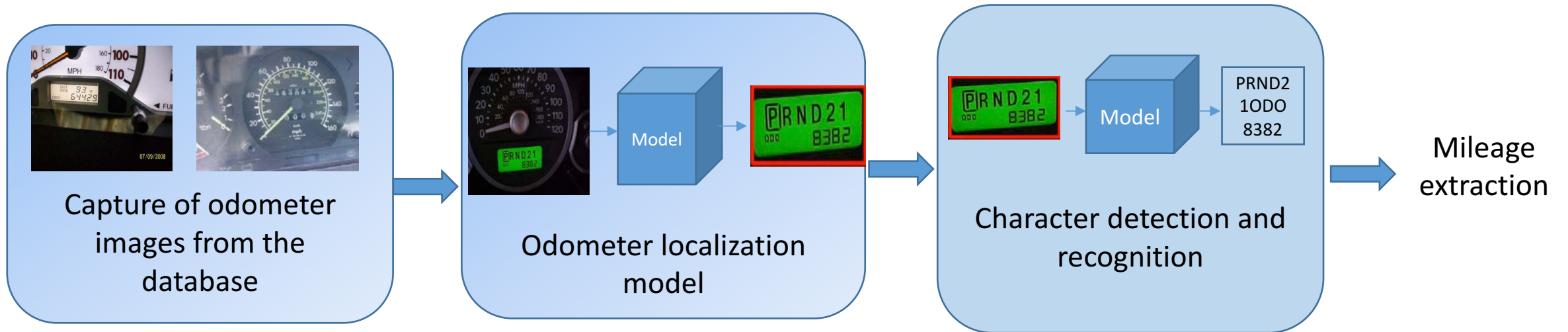


Challenges: Noisy images, non-uniform illumination, random noise, motion blur, odometer type variability (analog and digital).

# Character Recognition



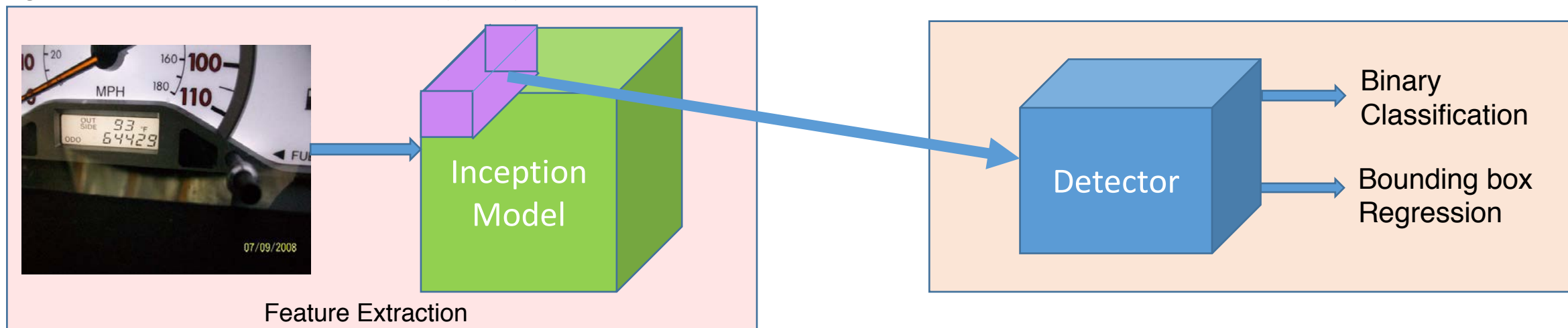
# Project Pipeline



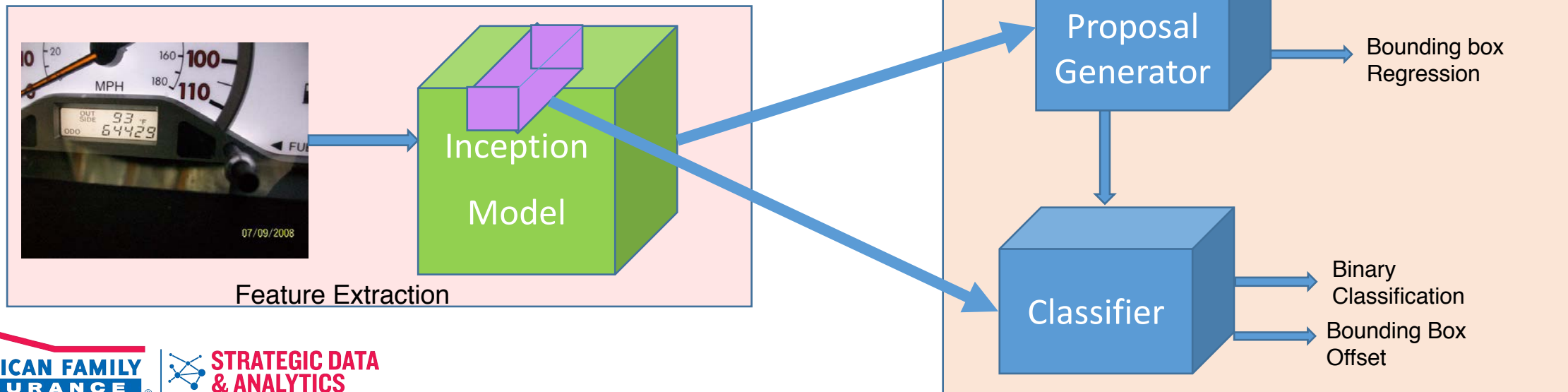


# Models for Object Localization

## Single Shot Multi Box Detector (SSD)



## Faster R-CNN Detector



# Odometer Localization Results

Results: 90% of the odometers in the images were correctly localized using Faster R-CNN.



True Positives



False Positives

# Final Remarks

Deep learning is driving computing innovation as the insurance industry sets its sights on artificial intelligence.

In this talk, we described different insurance use cases for deep learning:

- Roof condition prediction from images for property underwriting.
- BMI classification from selfie images for life insurance.
- Automatic odometer reading from images to accelerate claim processing.

