### The Promise, Pitfalls and Process of Al

Stephen J. Mildenhall October 24, 2019



Predictive Analytics <sup>-</sup>erence I Representation Deep Learning Data ScienceExpert System Data Visualization Statistics Exploratory Data Analysis Neural Network Big Data p > n Artificial Intelligence Data Represenatation Machine Learning





### Facts and Features





Year Built 1928



Heating Radiant





7,400 saft

### Central INTERIOR FEATURES

#### **Bedrooms**

### Beds: 3

#### **Bathrooms**

### Baths: 1 full, 1 half

### **Heating and Cooling**

Heating: Radiant Heating: Gas

#### **Appliances**

Dishwasher, Dryer, Garbage disposal, Microwave, Range / Oven, Refrigerator, Trash compactor, Washer

#### Flooring Floor size: 2,346 sqft

Flooring: Hardwood

See More Facts and Features V

#### Home Value

Zestimate @ \$565,863

## **Nearby Similar Sales**

Or call 773-974-9996 for more info

1010 Woodbine Ave, Oak Park,

**Contact Agent** 

#### SOLD: \$565.000 Sold on 05/23/18

3 bds, 2 ba, 1,932 sqft 846 Linden Ave, Oak Park, IL 60302

#### SOLD: \$560.000 Sold on 05/29/18

3 bds, 2.5 ba, 2,660 sqft

### 830 Belleforte Ave. Oak Park, IL 60302

SOLD: \$570,000 Sold on 04/26/19 4 bds, 2.5 ba, 2.332 sqft

913 Linden Ave. Oak Park, IL 60302

#### SOLD: \$550,000 Sold on 11/13/18

3 bds, 2.5 ba, 1,696 sqft

### 1047 N Grove Ave. Oak Park, IL 60302

SOLD: \$540.000 Sold on 11/30/18

4 bds, 2.5 ba, 2,204 sqft

1015 Woodbine Ave, Oak Park, IL 60302

See sales similar to 1010 Woodbine Ave

### Comparables

- Statistics: learning from data
- Descriptive statistics: summarize with no interpretation
- Exploratory data analysis (EDA): claims about current sample

### **Generalizing** Comparables

- Relevant comparables may not be available
- Extend limited data
- Predictive modeling: estimate value for an out of sample unit
- Buyer's perspective

### **Understanding** Comparables

- Seller's perspective: does an investment to
  - remodel the kitchen or
  - paint or
  - landscape or
  - add a bathroom or
  - make life easier for my broker

make sense?

Inferential statistics

 Machine learning: an algorithm allowing computer to build and improve a model from more data

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- GLMs
- Logistic regression
- Decision trees
- iCAS CSPA material

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#### **ML Problems**

- Classification
- Regression
- Transcription
- Machine translation
- Anomaly detection
- Synthesis
- Impute missing values
- De-noise
- Density estimation

## Comparables...

### Comparables. . .

- Encapsulate considerable domain expertise
- "Raw data is both an oxymoron and a bad idea; to the contrary, data should be cooked with care." (Geoffrey Bowker)

...and are a powerful data representation















- 1) A < B < C
- 2) B < A < C
- 3) B < C < A
- 4) C < B < A
- 5) None of the above

FYI, values are x:3x:9x

Home Facts by Data Source ALL SOURCES COMBINED COUNTY RECORDS County website Beds: 6 - Baths: 7.5 4.5 Soft: 12.000 7.914 Lot: 39,622 sq ft / 0.91 acres 39,621 sq ft / 0.91 acres Type: Single Family Single Family Year Built: 1929 1929 Last Sold: Aug 2011 for \$2,100,000 - Parking: Attached Garage Garage - Attached Cooling: Central Central Heating: Gas - Fireplace: Yes Yes Days on Zillow: 47 -MLS #: 10322977 - Other facts Additional Features: Air Conditioning: Zoned, Space Pac, Appliances: All Stainless Steel Kitchen Appliances, 2nd Bedroom Level: 2nd Level, Built Before 1978 (Y/N): Yes, Electricity: Circuit Breakers, 400 Amp Service or Greater, Family Room Level: Not Applicable, Heat/Fuel: Hot Water/Steam, 2+ Sep Heating Systems, Zoned, Sewer: Sewer-Public, Listing Type: Exclusive Right To Sell, Master Bedroom Level: 2nd Level, Parking Type: Garage, Water: Lake Michigan, Kitchen Level: Main Level, Living Room Level: Main Level, Exposure: W (West), S (South), E (East), N (North), Sale Terms: Conventional, 3rd Bedroom Level; 2nd Level, Equipment; Sump Pump. Sprinkler-Lawn, Power Generator, Kitchen Type: Pantry-Closet, Eating Area-Table Space, Island, Basement Bathrooms (Y/N): Yes, Addtl Room 2 Level: 2nd Level. Bath Amenities: Double Sink. Steam Shower. Lot Description: Corner, Addtl Room 4 Level: Main Level, Addtl Room 1 Level: 3rd Level, Addtl Room 3 Level: 3rd Level, 4th Bedroom Level: 2nd Level Addtl Room 3 Name: Office Addtl Room 5 Level: 3rd Level Addtl Room 1 Name: 5th Bdrm, Addtl Room 2 Name: 6th Bdrm, Frequency: Not Applicable, Attic: Full, Interior Stair, Finished, Addtl Room 6 Level: Basement, Addtl Room 7 Level: 2nd Level, Addtl Room 9 Level: Main Level Addtl Room 8 Level: 2nd Level Status: New Addtl Room 4 Name: Library, Addtl Room 5 Name: Attic, Addtl Room 9 Name: Gallery, Square Feet Source: Estimated, Master Bedroom Bath (Y/N): Full. Age: 81-90 Years, Area Amenities: Street Lights, Street Paved, Sidewalks, Pool, Garage On-Site: Yes, Additional Rooms: Recreation Room, 5th Bedroom, Sitting Room, 6th Bedroom, Recent Rehab (Y/N): Yes, Additional Sales Information: List Agent Must Accompany, 2nd Floor Laundry, In-Law Arrangement, Heated Floors, Breezeway, Dog Run, Fireplace Location: Other, Is Parking Included in Price: Yes Type of House 2: 2 Stories Garage Ownership: Owned Laundry Level: Not Applicable, Additional Rooms: Terrace, Attic, Office, Gallery, Library, Addtl Room 6 Name: Recreation Rm, Addtl Room 7 Name: Sitting, Addtl Room 10 Level: Main Level, Addtl Room 8 Name: Sitting, Addtl Room 10 Name: Terrace, Aprox. Total Finished Sq Ft: 0. Total Sq Ft: 0, Lot Dimensions: 215 X 184, Tax Year: 2017, Parcel Identification Number: 15013170210000 - Appliances Included: Dishwasher, Dryer, Garbage disposal, Microwave, Range / Oven, Refrigerator, Washer - County Name: Cook County Cook Covered Parking Spaces: 4 4 Exterior Material: Brick, Stone Stone Floor Covering: Carpet, Hardwood - Heating Type: Forced air - Laundry: In Unit - Parcel #: 1501317021 1501317021 Pets: Contact manager -Roof Type: Slate Slate Room Count: - 0 Room Types: Dining room -Stories: 0.0 2.0 Structure Type: Tudor Split level Unit Count: 0 0 Zillow Home ID: 123561615 -

From https://www.zillow.com/homes/for\_sale/Oak-Park-IL-60302/123561615\_zpid/84453\_rid/globalrelevanceex\_sort/41.897302.87.808603, 41.886408, -97.826799\_rect/15\_zm/?

Home Facts by Data Source ALL SOURCES COMBINED COUNTY RECORDS County website Beds: 5 4 Baths: 4.0 3.5 Soft: 3.360 3.360 Lot: 13,560 sq ft / 0.31 acres 13,562 sq ft / 0.31 acres Type: Single Family Single Family Year Built: 1899 1898 Last Sold: -- Parking: Detached Garage Garage - Detached Cooling: Central - Heating: Gas -Fireplace: Yes Yes Days on Zillow: 117 - MLS #: 10344749 - Other facts Additional Features: Addtl Room 5 Level: Not Applicable. Appliances: Oven Double, All Stainless Steel Kitchen Appliances, 2nd Bedroom Level: 2nd Level, Built Before 1978 (Y/N): Yes, Sewer: Sewer-Public, Listing Type: Exclusive Right To Sell, Master Bedroom Level: 2nd Level, Parking Type: Garage, Space/s, Tax Exemptions: Homeowner, Equipment: Ceiling Fan, CO Detectors, Water: Lake Michigan, Kitchen Type: Eating Area-Breakfast Bar, Eating Area-Table Space Island Kitchen Level: Main Level Living Room Level: Main Level, 3rd Bedroom Level: 2nd Level, Other Information: School Bus Service, Commuter Train, Historical District, Exposure: E (East). Electricity: 200+ Amp Service, Exterior Building Type: Frame, Addtl Room 2 Level: Main Level, Addtl Room 2 Name: Foyer, Addtl Room 3 Level: 2nd Level. Addtl Room 1 Level: 3rd Level. Addtl Room 4 Level: 2nd Level, 4th Bedroom Level: 2nd Level, Driveway: Side Drive. Concrete, Addtl Room 4 Name: Balcony, Addtl Room 1 Name: 5th Bdrm, Family Room Level: 3rd Level, Age: 100+ Years, Addtl Room 10 Level: Not Applicable, Addtl Room 6 Level: Not Applicable, Addtl Room 7 Level: Not Applicable, Addtl Room 8 Level: Not Applicable, Addtl Room 9 Level: Not Applicable, Frequency: Not Applicable, Attic: Full. Interior Stair Finished Foundation: Stone Status: New Souare Feet Source: Assessor, Master Bedroom Bath (Y/N): Full, Bath Amenities: Soaking Tub. Area Amenities: Curbs/Gutters. Street Lights. Street Paved, Sidewalks, Garage On-Site: Yes, Parking On-Site: Yes, Additional Rooms: Balcony, 5th Bedroom, Additional Sales Information: List Agent Must Accompany, Heated Floors, Balcony, Parking Ownership: Owned, Fireplace Location: Other, Is Parking Included in Price: Yes, Parking: Driveway, Type of House 2: 3 Stories, Lot Dimensions: 78X175. Additional Rooms: Fover, Garage Ownership: Owned, Laundry Level: Not Applicable Aprox. Total Finished So Et: 0. Total So Et: 0. Basement Sq Ft: 0, Tax Year: 2017, Parcel Identification Number: 16071030160000 - Appliances Included: Dishwasher, Dryer, Garbage disposal, Microwave, Range / Oven, Refrigerator, Trash compactor, Washer - Assisted Living Community: No - Attic: Yes - Barbecue Area: No - Basement Type: Partial - Basketball Court: No - Cable Ready: No - Ceiling Fan: Yes - Controlled Access: No - County Name: Cook County Cook Covered Parking Spaces: 7 2 Deck: No - Disability Access: No - Dock: No - Doorman: No - Double Pane/Storm Windows: No -Elementary School: Oliver W Holmes Elementary Schoo - Elevator: No - Exterior Material: Stone, Wood Wood Fenced Yard: Yes - Floor Covering: Hardwood, Slate - Garden: Yes - Gated Entry: No -Greenhouse: No - Fitness Center: No - Heating Type: Baseboard. Radiant - High School: Oak Park & River Forest High Sch - Hot Tub/Spa: No - Intercom: No - Jetted Tub: No - Laundry: In Unit -Lawn: No - Legal Description: - (WORWICKS) SUB OF LT1 EX N7FT & ALL OF LT2 IN BLK2 IN KETTLESTRINGS ADD TO HARLEM SUB OF PT OF NW SEC 07-39-13 Middle School: GWENDOLYN BROOKS MIDDLE SCHOOL - Mother-in-Law Apartment: No - Near Transportation: No - Over 55 Active Community: No - Parcel # 1607103016 1607103016 Patio: Yes - Pets: Contact manager - Pond. No - Pool: Yes - Porch: Yes - Roof Type: Asphalt Shake/Shingle Room Count: 9 0 Room Types: Dining room, Family room, Laundry room. Master bath, Walk-in closet - RV Parking: No - Sauna: No - Security System: No - Skylight: Yes - Sports Court: No - Sprinkler System: No - Storage: No - Stories: 3.0 2.0 Structure Type: Victorian Other Tennis Court: No - Unit Count: 0 0 Vaulted Ceiling: No - Waterfront: No -Wet Bar: No - Wired: No - Zillow Home ID: 3803093 -

Home Facts by Data Source ALL SOURCES COMBINED COUNTY RECORDS County website Beds: 3 3 Baths: 1.0 1.0 Soft: 2.080 1.376 Lot: 4,159 sq ft / 0.10 acres 4,158 sq ft / 0.10 acres Type: Single Family Single Family Year Built: 1914 1913 Last Sold: - - Parking Detached Garage Garage - Detached Cooling: Central Central Heating: Gas - Fireplace: - - Days on Zillow: 104 - MLS #: 10260750 - Other facts Additional Features: Addtl Room 4 Level: Not Applicable. Addtl Room 5 Level: Not Applicable, 4th Redroom Level: Not Applicable Built Before 1978 (Y/N): Yes, Sewer: Sewer-Public, Listing Type: Exclusive Right To Sell. Parking Type: Garage. Tax Exemptions: Homeowner, Senior, Senior Freeze, Addtl Room 1 Level: Main Level. 2nd Bedroom Level: Main Level, Water: Lake Michigan, Kitchen Level: Main Level, Living Room Level: Main Level, Master Bedroom Level: Main Level, Other Information: School Bus Service, Commuter Bus, Addtl Room 3 Level: Main Level, 3rd Bedroom Level: Main Level, Addtl Room 2 Name: Recreation Rm. Addtl Room 1 Name: Breakfast Rm. Family Room Level: Rasement Addtl Room 2 Level: Rasement Age 100+ Years, Addtl Room 10 Level: Not Applicable, Addtl Room 6 Level: Not Applicable, Addtl Room 7 Level: Not Applicable, Addtl Room 8 Level: Not Applicable, Addtl Room 9 Level: Not Applicable, Frequency: Not Applicable, Status: New, Square Feet Source: Plans, Master Bedroom Window Treatments (Y/N): Curtains/Drapes, 3rd Bedroom Window Treatments (Y/N): Shades, Exterior Building Type: Aluminum Siding, Addtl Room 2 Window Treatments (Y/N): Blinds, Family Room Window Treatments (Y/N): Blinds, Area Amenities: Curbs/Gutters, Street Lights Street Paved Sidewalks Park/Playground Addtl Room 3 Window Treatments (Y/N): Curtains/Drapes, Garage On-Site: Yes, 1st Floor Bedroom, 1st Floor Full Bath, Additional Rooms: Recreation Room, Breakfast Room, Sun/Florida Room Heated, Is Parking Included in Price: Yes, Addtl Room 3 Name: Sun/Florida Room Heated, Type of House 2: 1 Story, Lot Dimensions: 125 X 33, Garage Ownership: Owned, Laundry Level: Not Applicable, Aprox. Total Finished Sq Ft: 2080. Main So Ft: 1376. Total So Ft: 1376. Tax Year: 2016. Parcel Identification Number: 16053110300000 - Appliances Included: Dishwasher, Dryer, Range / Oven, Refrigerator, Washer - County Name: Cook County Cook Covered Parking Spaces: 2 2 Elementary School: Whittier Elementary School - Exterior Material: Wood Wood Floor Covering: Carpet, Hardwood, Linoleum / Vinyl - Heating Type: Forced air - High School: Oak Park & River Forest High Sch - Laundry: In Unit - Legal Description: - (HOOKERS) SUB OF NE SW SEC 05.39.13 Middle School: GWENDOLYN BROOKS MIDDLE SCHOOL - Parcel #: 1605311030 1605311030 Pets: Contact manager - Roof Type: Tile Shake/Shingle Room Count: - 0 Room Types: Dining room - Stories: 0.0 1.0 Structure Type: Bungalow Bungalow Unit Count: 0 0 Zillow Home ID: 3799641 -

From https://www.zillow.com/homes/tor\_sale/Oak-Park-IL-60302/3799641\_zpid/84453\_rid/globalrelevanceex\_sort/41.90225: 87.774485.41.89136.-87.792681\_rect/15\_zm/?

### Humans excel at image recognition...

- Humans can process images very quickly
  - The three images contain several **megabytes** of data,  $p \gg n$
- Computers can process structured text very quickly
  - The raw text contains a few kilobytes of information
- Which system would you rather try to program?

...and many other problems computers find hard

## Artificial Intelligence (AI)

- Artificial Intelligence: solving problems people find easy but computers find hard
  - Image interpretation
  - Conversation
  - Reading emotions
  - Catching frisbees
- Skills shared with other animals having neuron-based brains
- Al is a problem domain



### The Premise of Al...

Appears people have gotten a long way with a single trick

- Recent studies suggests the mammalian brain uses a single algorithm to solve most tasks
- Previously machine learning research fragmented: distinct approaches for natural language processing, vision, motion planning, speech recognition

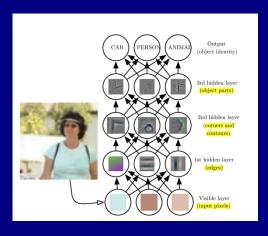
...today AI leverages a single general purpose algorithm inspired by, but not simulating, brain function to solve a wide range of problems

### Al General Purpose Algorithm

- Learn model from experience
  - Use machine learning
  - Show don't tell background information
- Learn what is important
  - Data representation created simultaneously with model
- Understand the world as a hierarchy of concepts
  - Complicated concepts built out of simpler ones
- Hierarchy of concepts has many layers
  - Deep learning

### Al General Purpose Algorithm...

- Hierarchy layers built using simple neuron nodes
- Neural networks build new covariates from non-linear hinge functions of existing covariates and the output of other layers
- Optimize overall fit with least squares



...algorithm inspired by our understanding of brain function

### Al General Purpose Algorithm: A Naïve Description

- Neural networks and deep learning are fancy least squares
  - Solve using gradient descent = walking downhill
  - Stochastic gradient descent = probably walking downhill
- Approximate complex functions by compositions of simple ones
- Build and combine specialized problem solving nodes
  - recognize color...
  - recognize car vs truck vs bike vs motorcycle vs ....
  - can combine to recognize red car, blue truck, ...
- Pool, share and feedback information between nodes
- Nodes simultaneously learn model and data representation

### Al Has a Long History...

Old Approach	New Approach
Knowledge base: hard-coded knowledge about world	Machine Learning: extract model from training data
<b>Expert system</b> : humans decide important features	Representation Learning: determine important factors of variation
Neural networks	Deep learning neural networks

...but hasn't always performed as expected

### Why Is AI Working Now?

- Better algorithms
- More data
- Bigger models

#### 'Godfathers of AI' honored with Turing Award, the Nobel Prize of computing

Yoshua Bengio, Geoffrey Hinton, and Yann LeCun laid the foundations for modern Al By James Vincent | Mar 27, 2019, 6:02am EDT











From left to right: Yann LeCun | Photo: Facebook; Geoffrey Hinton | Photo: Google; Yoshua Bengio | Photo: Botler Al

### Why AI Is Working Now: Better Algorithms

Pre-2006 deep networks were believed to be very difficult to train

- Better algorithms devised by Hinton in 2006 allowed deeper models by making them faster to train
- Beginning of the deep learning era
- Tens of thousands of researchers working in AI/ML

#### A fast learning algorithm for deep belief nets

<u>GE Hinton</u>, <u>S Osindero</u>, <u>YW Teh</u> - Neural computation, **2006** - MIT Press

We show how to use "complementary priors" to eliminate the explaining-away effects that make inference difficult in densely connected belief nets that have many hidden layers.

Using complementary priors, we derive a fast, greedy algorithm that can learn deep, directed ...



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Using complementary priors, we derive a fast, greedy algorithm that can learn deep, directed ...

☆ ワワ Cited by 10337 Related articles All 61 versions

## [PDF] A systematic relationship between **minimum bias** and generalized linear models

SJ Mildenhall - Proceedings of the Casualty Actuarial Society, 1999 - casact.org
The minimum bias method is a natural tool to use in parameterizing classification
ratemaking plans. Such plans build rates for a large, heterogeneous group of insureds using
arithmetic operations to combine a small set of parameters in many different ways. Since the ...

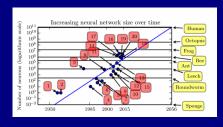
□□ Cited by 77 Related articles All 5 versions 
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### Why Al Is Working Now: More Data

- More data: IOT, Big Data, digitization of society
- More data = less skill required to train model
- Data rules of thumb: a supervised deep learning algorithm will generally
  - achieve acceptable performance with around 5,000 labeled examples per category
  - will match or exceed human performance when trained with a dataset containing at least 10 million labeled examples

### Why Al Is Working Now: Bigger Models

- Bigger = deeper models
  - More computer power
  - Better algorithms
- Model size: connections and number of neurons or nodes
- Biological neurons not densely connected
  - Models within order of magnitude of mammal brains
- Number of neurons: current models are very small
  - Doubling every 2.4 years
  - Match human brain by 2050

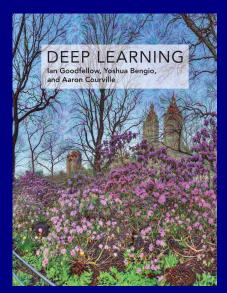


...can't expect much from a leech-brain sized network

### Deep Learning Summary

**Deep learning** is an approach to AI. Specifically, it is a type of machine learning, a technique that allows computer systems to improve with experience and data.

- ... machine learning is the only viable approach to building Al systems that can operate in complicated, real-world environments
- Deep learning...achieves great power and flexibility by learning to represent the world as a nested hierarchy of concepts



Central Park Dreamscape CGI

### When Does Al Work Well?

#### Stunning results

- Static, rules based environment
- Clean, direct observations
- Essentially limitless data
- Definitive right answer

#### **Examples**

- Classification problems
- Image, speech recognition
- Pedestrian detection
- Traffic sign classification
- Simple dynamic control
- Reinforcement learning, robotics
- Synthetic data, alphaZero
- GANs

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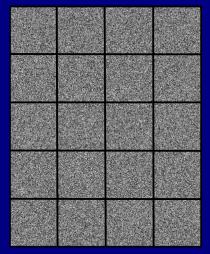
#### Characteristics of insurance

- Behavioral feedbacks
- Dynamic: reacts as we learn
- Proxy data, not direct
- Inherently small classes
- Uncertain data: development
- Latent data

...Al works well on certain types of real world data

### How is *Real Data* Special?

- Real data often appears very high dimensional (p ≫ n) but really contains hidden
   structure
- Photographs do not look like random images
  - Continuous: image near a photograph recognizable
  - No jumps or cliffs
  - Not fractal-like



Real images occupy a **negligible proportion** of the total volume of image
space

### How is *Real Data* Special?

#### Helpful Al data rules of thumb

#### Representation learning: A review and new perspectives

Y Bengio, A Courville, P Vincent - IEEE transactions on pattern ..., 2013 - ieeexplore.ieee.org
The success of machine learning algorithms generally depends on data representation, and
we hypothesize that this is because different representations can entangle and hide more or
less the different explanatory factors of variation behind the data. Although specific domain
knowledge can be used to help design representations, learning with generic priors can
also be used, and the quest for AI is motivating the design of more powerful representationlearning algorithms implementing such priors. This paper reviews recent work in the area of ...

☆ ワワ Cited by 4662 Related articles All 40 versions

### Al Data Rules of Thumb

Assumption	Concerns
Smoothness	Weather, butterfly effect; singularities
Multiple explanatory factors  Common material drivers	Sudden change in relationship or sentiment
Shared factors across tasks  Hierarchical organization of factors	Phase change or market crash
Simple linear factor dependencies	Discontinuities, exponential
Manifolds: real data is low <i>d</i> Natural clustering Sparsity, 80/20 explanatory rule	
Temporal and spatial coherence Causal factors	Feedback loops, latency State dependent causality

### Insurance-Related AI Applications

#### Good fit general business

- Marketing
- Customer service
- Underwriting homogeneous classes
- Fraud detection

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#### Potential?

- Claims process management, reserves
- Data evaluation, anomalies, completion
- Risk management in high frequency lines
- Underwriting accounts: pricing or ranking?

### Insurance-Related AI Applications

#### **Good fit** general business

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#### More exposure to see rarer events

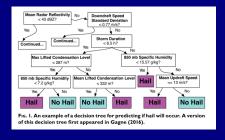
- Good performance = **5000** observations
- 50,000 exposures at 10% claim rate
- 500,000 exposures at 1% claim rate
- 5 million exposures at 0.1% claim rate

 For approx. 1% claim rate see examples on r/CatastrophicFailure, r/IdiotsInCars, r/nononono, r/OSHA, r/Roadcam

Al needs lots of data... ...can you collect enough?

### Al Compared to Catastrophe and Weather Models

- Is enough data practically possible?
- Weather and catastrophe models are physics-based models
- Physical models better than a learned phenomenological models
- Re-learning physics inefficient
- Al can work effectively in conjunction with physics-based models



...Al already being applied to forecast high impact weather events

## Al is Not a Universal Panacea—No Singularity

Problems	Computers find Easy	Computers find Hard
		<b>B</b> . II
	Clustering	Reading emotion
People find	Consistency	Conversation
Easy	Inconsistency	Unstructured interaction
	Outliers	Physical interactions
	Big data	Underwriting
People find	Complex computations	Investing
Hard	Simulations	Middle East Peace Plan
	Dispassionate	Identifying fake news
		Gödelization, thinking outside the box

### Al Bias and Sources of Modeling Failure

# Statisticians know training data must be a **random** cross section of population

- Extend tools to include new characteristic picked up by ML
  - Computers do not see images with semantic understanding
  - Computers see things the eye cannot detect: lighting, type of camera
  - Computers don't filter irrelevant information
- Enhance skills for the AI modeling era
  - Enhanced methodological rigor managing training data
  - Technical tools to analyze and diagnose the behavior of the model
  - Training, education and caution in the deployment of ML in products

### Al Bias and Sources of Modeling Failure

- Model-in-use can fail for three reasons
  - Algorithm
  - Implementation
  - Bureaucracy
- Understanding decisions in big organizations is very complex
  - Like ML models, people complex and opaque

...Al is just a machine—think washing machine not HAL9000—powerful but limited

Benedict Evans

### Three Things to Remember...

- Al = solving problems people find easy but computers find hard
- Spectacular progress: bigger models; better algorithms; more data
  - Show don't tell
  - Derive data representation and model form
- Al relies on special characteristics of data: applied to non-Al data it can disappoint

Thank you!

### References

'p > n'. 'Statistics']

Slide(s)	Reference
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26	Bengio, Yoshua, Aaron Courville, and Pascal Vincent. "Representation learning: A review and new perspectives." IEEE transactions on pattern analysis and machine intelligence 35.8 (2013): 1798-1828. https://arxiv.org/pdf/1206.5538.pdf
29	McGovern, Amy, et al. "Using artificial intelligence to improve real-time decision-making for high-impact weather." Bulletin of the American Meteorological Society 98.10 (2017): 2073-2090. https://journals.ametsoc.org/doi/pdf/10.1175/BAMS-D-16-0123.1
31-32	Benedict Evans, Notes on Al Bias (2019) https://www.ben-evans.com/benedictevans/2019/4/15/notes-on-ai-bias

```
## AI talk buzzword bingo image, Python Code
                                                                                  word_freqs = { w: (0.4 + 0.6 * np.random.rand()) for w in words}
from imageio import imread
                                                                                  word fregs['Artificial Intelligence'] = 2.5
import matplotlib.pyplot as plt
                                                                                  word_freqs['Deep Learning'] = 1.5
from wordcloud import WordCloud
                                                                                  word fregs['Neural Network'] = 1.5
import numpy as np
                                                                                  wordcloud = WordCloud(colormap=plt.cm.magma. **args)
args = {'width':3600, 'height':2400, 'max words': 200,
        'max_font_size':240, 'relative_scaling': 0.4,
                                                                                  for i in range(30):
        'background color' : 'white', 'prefer horizontal':0.75}
                                                                                          wordcloud.generate_from_frequencies(word_freqs)
                                                                                         plt.figure(figsize=(11.8.5))
words = ['Artificial Intelligence', 'Big Data',
                                                                                         plt.imshow(wordcloud)
 'Causal Analysis', 'Data Representation', 'Data Visualization',
                                                                                         plt.axis("off")
 'Deep Learning', 'Data Representation', 'Data Science'
                                                                                         plt.savefig(f'word_cloud_{i}.png', dpi=300)
 'Expert System', 'Exploratory Data Analysis', 'Inference',
                                                                                         plt.show()
 'Machine Learning', 'Neural Network', 'Predictive Analytics',
```