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**The demise of physician authority and the de-mystification of the clinician – patient relationship**

- Health care becomes commodified and competitive in the interests of increased revenues and efficiency;
- Media reports of medical errors, corporate greed, seemingly heartless denials of insurance coverage increase distrust of organized medicine and their personnel;
- Competition among physician groups and their physician extenders for securing reimbursement: who should be doing what?
- The rise of upper and mid-level non-physician management and their usurping the power of physicians in the interests of corporate's concerns about achieving "productivity targets"



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But health professionals still did OK reputation-wise: Maybe we distrusted organized medicine, but we continued to trust our own private doctors and nurses.



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Why?

- Because clinicians still exerted control over clinical interventions
- The technology kept getting increasingly sophisticated but it didn't replace the physician, nurse or therapist from performing their clinical, interventionist functions.



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What is AI?

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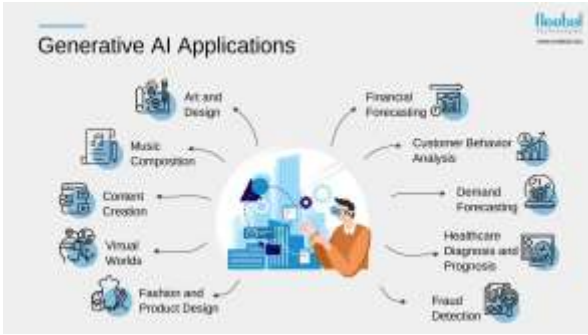
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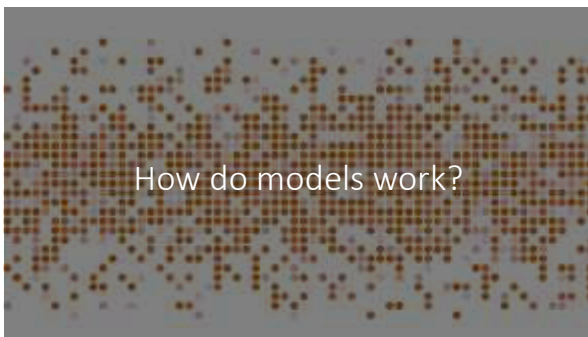
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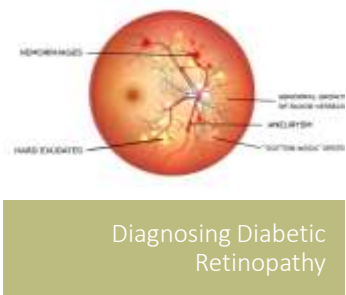
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- Numerous researchers have validated MLCs on DR;
- Ting et al used 484,661 retinal images with 76,370 DR images and the rest spread across glaucoma, AMD and normal; Gulshan et al used 118,419 images: 53,759 normals and 64,660 DR;
- Gulshan's accuracy was 0.990-0.991 for detecting referable DR with sensitivity and specificity for referable DR over the two data sets at 0.870-0.903 and 0.981-0.985

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"Deep Learning at Chest Radiography: Automated Classification of Pulmonary Tuberculosis by Using Convolutional Neural Networks," Paras Lakhani and Baskaran Sundaram, Radiology, 2017; 284(2):574-582.



- "The best performing classifier had an AUC of 0.99, which was an ensemble of the AlexNet and GoogleNet DCNNs...Deep learning with DCNNs can accurately classify TB at chest radiography with an AUC of 0.99. A radiologist augmented approach for cases where there was disagreement among the classifiers further improved accuracy."

Figure 4. (A) Posteroanterior chest radiograph shows upper lobe opacities with pathologic analysis-proven active TB. (B) Same posteroanterior chest radiograph, with a heat map overlay of one of the strongest activations obtained from the fifth convolutional layer after it was passed through the GoogleNet-76 classifier. The red and light blue regions in the upper lobes represent areas activated by the deep neural network. The dark purple background represents areas that are not activated. This shows that the network is focusing on parts of the image where the disease is present (both upper lobes).

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## DEEP PATIENT

- Miotto, R. *et al.* Deep Patient: An Unsupervised Representation to Predict the Future of Patients from the Electronic Health Records. *Sci. Rep.* 6, 26094; doi: 10.1038/srep26094 (2016).




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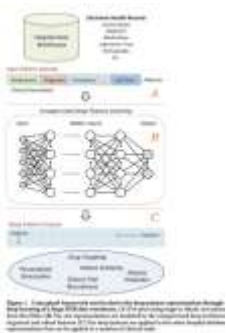
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### What's Deep Patient?

- Using patient data from the hospital data warehouse to predict diseases that might develop
- The training set included ~1.6 million patients treated between 1980 and 2013 (Mt. Sinai, NY)
- The test set included 100,000 patients in 2014 evaluating 79 diseases based on ICD-9 codes, e.g., endocrinology, cardiology, and oncology

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Table 1. Area under the curve (AUC) for 79 diseases

| Disease  | Area under the ROC curve |       |              |
|--|--------------------------|-------|--------------|
|  | Baseline                 | PCA   | Deep Patient |
| Diabetes mellitus with complications                     | 0.798                    | 0.803 | 0.797        |
| Cancer of esophagus and esoph                            | 0.607                    | 0.611 | 0.607        |
| Cancer of liver and intrahepatic bile duct               | 0.600                    | 0.607 | 0.606        |
| Diagnosed acute fibrill abdominal colitis                | 0.614                    | 0.643 | 0.679        |
| Congestive heart failure (non-hypertensive)              | 0.608                    | 0.608 | 0.608        |
| Intentional self-harm of depressive behavioral disorders | 0.780                    | 0.787 | 0.803        |
| Cancer of pancreas                                       | 0.605                    | 0.630 | 0.679        |
| Schizophrenia  | 0.791                    | 0.796 | 0.807        |
| Multiple sclerosis                                       | 0.781                    | 0.789 | 0.808        |
| Acute myocardial infarction                              | 0.771                    | 0.776 | 0.807        |

- “The results were compared with other (predictive) techniques such as raw EHR, principal component analysis (PCA), k-means, and generalized method of moments (GMM) analysis – and it outperformed them all, producing an area under the curve (AUC) of 0.79 versus the other methods.”

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## Results



### A Primary Care (AI) Office Visit in the Future (10 - 20 years?)

- SYMPTOMS + TESTS
- FAMILY HISTORY + PATIENT HISTORY
- PATIENT'S DNA
- ELECTRONIC MEDICAL RECORDS
- REVIEW DIFFERENTIAL DIAGNOSES AND RECOMMENDATIONS DECISION(S) THROUGH PUB MED DATABASES




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Machine learning applications will only be limited by the imaginations of programmers, their knowledge base (which might be enhanced by AI itself!), and the physical/chemical limitations of their technologies.

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So what does this mean for YOU?

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### #1: The De-skilling Threat



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AI and all the Imaging Sciences: The Threat of "Deskilling" from an automation bias.

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### #2: Legal Liability?

- In the instance of disagreement with an AI model, how should the human clinician proceed, especially with an eye to protecting his or her liability?
- Notice: If you follow it or if you override it and there's a bad outcome, you might be sued either way.
- How to understand your obligation to the patient under such threat?

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#3: Healing as a derivative of the geographical and psychological proximity of clinician to patient. Will digital technologies encourage that loss?

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#4: Integration of AI and humans: Would AI have even prevented the Vaught/Murphey Tragedy?

- Very likely, but the hospital was having trouble with their medication ordering system, which is why Vaught had to go to another computer to get the medicine and then override its alarms. Nurses at Vaught's hospital were commonly overriding alarms. What happens when systems go down?

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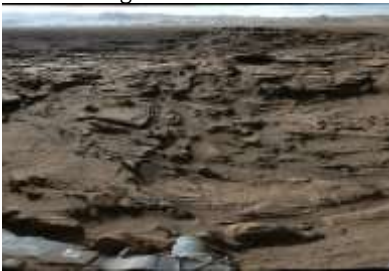
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#6: The Transition to AI Technologies Will Not be Smooth




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### Preparing for AI

- The learning curve will be fraught; the work-flow integration/implementation curve will be fraught; the maintenance curve will be never-ending;
- Discrete systems or tightly coupled?
- The ubiquity of tech teams
- The necessity of feedback loops and highly networked collaborations
- Planning will be lengthy and involve an unprecedented number of multi-disciplinary teams




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### #7: Medical Ethics Today Versus 50 Years Ago

- Honor patients' right to information and decisionmaking
- Practice honesty with patients
- Protect confidentiality
- Disclose errors
- Don't be intimidating, rude, disrespectful, offensive, or sexually provocative with patients
- Guard against personal bias
- Respect and honor teamwork
- Stay abreast with medical science and the contemporary standard of care




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**Implication: Professionalism is a Moving Target**

- Professionalism is shaped by a host of factors:
  - Scientific knowledge
  - New technology and the ethical challenges that accompany their use
  - Economic resources of providers and consumers
  - Cultural and valuatve diversity
  - Legal liability
  - Consumer expectations
  - Insurance trends

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Ethical Practices

- We experiment with and create them
- Let's hope they're good
- If not, let's hope we have the freedom to change them



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