



Dr. Detective

combining gamification techniques and crowdsourcing to create a gold standard in medical text

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- Watson = natural language question answering system
- training data:
 - databases
 - ontologies
 - taxonomies
 - i.e. ground truth



• goal: adaptation of Watson for other domains

(e.g. medical domain)

• problem: how to acquire ground truth?

Training Watson for the medical domain:

- Answer questions about diagnosing
- Find synonym phrases
- Identify negation (and its variations)
- Identify different term types
- Identify relations between term types

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Problem: language ambiguity

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Problem: language ambiguity

Patients exhibiting *acute tailbone pain* should be examined for extra bone nodules.

- traditional approach: guidelines for consistently choosing one answer (e.g. tailbone pain is the primary term)
- **Crowd Truth** approach: capture and measure diversity of opinion (e.g. by counting votes for each variation of a term)

Research goals

- Investigate the feasibility of a game for niche (expert) sourcing that captures a diversity of opinions
- Measure quality of *Crowd Truth* through metrics
- Evaluate effectiveness of game for engaging the experts

Crowd-Watson Workflow



available at: http://crowd-watson.nl

Crowd-Watson Workflow: Game Aspects



available at: http://crowd-watson.nl/dr-detective-game/

Crowd-Watson Workflow: Input Data

CASE RECORDS OF THE MASSACHUSETTS GENERAL HOSPITAL

Richard C. Cabot, Founder, Eric S. Rosenberg, M.D., Editor, Nancy Lee Harris, M.D., Editor, Jo-Anne O. Shepard, M.D., Associate Editor, Alice M. Cort, M.D., Associate Editor, Sally H. Ebeling, Assistant Editor, Emily K. McDonald, Assistant Editor

Case 23-2013 — A 54-Year-Old Woman with Abdominal Pain, Vomiting, and Confusion

Kamyar Kalantar-Zadeh, M.D., M.P.H., Ph.D., Raul N. Uppot, M.D., and Kent B. Lewandrowski, M.D. N Engl J Med 2013; 369:374-382 | July 25, 2013 | DOI: 10.1056/NEJMcpc1208154

Comments open through July 31, 2013

Poll open through July 23, 2013

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- *input:* patient case reports
- *source:* New England Journal of Medicine

Article References Comments (221)

The description of this case was presented as a Case Challenge. Readers were invited to review the case description, vote on the diagnosis, and submit comments. The full case discussion and final diagnosis now appear below, along with the poll results.

PRESENTATION OF CASE

Dr. Sara R. Schoenfeld (Medicine): A 54-year-old woman was admitted to this hospital because of abdominal pain, vomiting, and confusion.

The patient was in her usual health until approximately 3 days before admission, when she reportedly began to feel unwell, with weakness, chills, and skin that was abnormally warm to the touch. She self-administered aspirin, without improvement. During the next 2 days, her oral intake decreased. Approximately 22 hours before presentation, vomiting occurred. Nine hours before presentation, she began to travel home to Italy from the eastern United States. During the next 2 hours a presentation and shorthouss of breath, and

notification from her primary care doctor 1 week later that she was doing well and had resumed her normal daily activities.

FINAL DIAGNOSIS

Toxic effects of metformin.



Crowd-Watson Workflow: Pre-processing

- input data filtering
- micro-task template setting
- target crowd setting



Crowd-Watson Workflow: Pre-processing

- input data filtering
- micro-task template setting
- target crowd setting

- segment text into paragraphs
- extract diagnosis, specialization domains
- evaluate length of paragraphs, sentences
- evaluate readability (SMOG)
- medical named entity recognition (NER)



Dr. Watson	Home	Game	High Scores	About	Hi android! You scored: 63.0 points	1 🖡	(logout)	
Level: Hard, Domain: Primary Care/Hospitalist/Clinical Practice								

In the following text, find all the clues that could help diagnose 2009 influenza A (H1N1) virus infection.

0

Step 1: Select the type of clue you are looking for.

Step 2: To pick a clue, highlight all the words that describe it by clicking on them. 2

On the second day, hypoxemia (Table 2) and renal failure (Table 1) developed and urine output fell to 20 to 30 ml per hour. Transthoracic echocardiography showed an ejection fraction of 50% and was otherwise normal. Microscopical examination of the urine sediment revealed white-cell casts and granular casts, with tubular cells and nondysmorphic red cells. Continuous venovenous hemofiltration was begun, complicated by catheter-related thrombosis. Heparin was administered.

Show clues by others						
Save clue per hour						
On the second day ×						

Time/Duration -

Step 3: After all the words in the clue are highlighted, save the clue.

Step 4: After you found all the clues for Time/Duration, submit them.

Submit your clues for Time/Duration

Step 5: Go back to step 1 and select another clue type, or move on to the next diagnosis.

Next diagnosis >>

2

Gaming elements:

- difficulty
- scoring
- immersion
- playing options
- others' answers

Features in the difficulty vector of a paragraph:

- number of words in the paragraph
- number of sentences in the paragraph
- average sentence length
- SMOG readability index of text
- number of medical terms



Gaming elements:

- difficulty
- scoring
- immersion
- playing options
- others' answers

Users receive points for:

- paragraph difficulty
- consecutive answers
- popular answers
- new answers
- wrong answers



Hi android! You scored: 409.3 points

+354.3 points gained from other users agreeing with your answers

-162 points deducted by other users disagreeing with your answers

Gaming elements:

- difficulty
- scoring
- immersion
- playing options
- others' answers



- influences the next document selection mechanism
- next document needs to have increased difficulty, but minimum differing features



Gaming elements:

- difficulty
- scoring
- immersion
- playing options
- others' answers

Pick your domain: 🚱

- Hematology/Oncology
- Nephrology
- Primary Care/Hospitalist/Clinical Practice
- Viral Infections
- **domains:** selected from the most popular categories in NEJoM

Pick the level you want to play: 😧

Quick Game	Normal Game	Hard Game

 levels: based on the time it takes to solve a paragraph



Gaming elements:

- difficulty
- scoring
- immersion
- playing options
- others' answers

Q: Does having access to the answers of other users stimulate diversity of opinion?

- game v1: option to view others' answers
- game v2: no such option

The patient returned the next afternoon because of persistent fever, cough, myalgias, low back pain, and new scrotal pain. The temperature was 39.0°C, and the other vital signs were normal. There were rhonchi in the left lower lung field, and the remainder of the examination was normal. A test for Lyme disease, sent the day before, was negative. Other test results are shown in Table 1. A chest radiograph showed incomplete segmental consolidation of the apical posterior segment of the right upper lobe and right hilar prominence, features suggestive of pneumonia and lymphadenopathy, respectively. Levofloxacin was prescribed, and he was sent home.



Q: Does having access to the answers of other users stimulate diversity of opinion?



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Pilot run results

Q: How do the answers annotated by the crowd compare to those found by an NER?

- NER: UMLS MetaMap
- full term match = the crowd found at least one term that has all the words in the term found by the NER
- partial term match = the crowd found at least one term that has any of the words in the term found by the NER

Pilot run results

Q: How do the answers annotated by the crowd compare to those found by an NER?



words distribution

terms distribution

Pilot run results

Q: How do the answers annotated by the crowd compare to those found by an NER?



words distribution

terms distribution

Crowd-Watson Workflow: Disagreement Analytics

Metrics:

- expert metrics
- sentence metrics
- term metrics
- relation metrics
- domain metrics



Conclusions

- crowd answers are at least as accurate as an NER for term extraction and categorization
- to capture disagreement, access to the answers of others should be limited
- most users enjoyed playing the *Dr. Detective* game

Future Work

- *user interaction:* score reports, more challenging tasks
- *data analysis:* specialized disagreement analytics for game
- *integration:* combine gaming and micro-task crowdsourcing
- running experiments with more participants

Try it out:

- CrowdWatson: http://crowd-watson.nl
- Dr. Detective: http://crowd-watson.nl/dr-detective-game

Related talks:

- Content and Behavior-Based Metrics for Crowd Truth @CrowdSem 14:45
- Domain-Independent Quality Measure for Crowd Truth Disagreement @DeRiVE 14:50

Read more:

- Measuring Crowd Truth for Medical Relation Extraction, Aroyo & Welty, AAAI Fall Symposium SPD '13
- Crowd Truth: Harnessing disagreement in crowdsourcing a relation extraction gold standard, Aroyo & Welty, WebSci '13
- *Dr. Detective*, Dumitache, MSc thesis: http://goo.gl/doAERZ