

Data Science Approaches to Assess Suicide Risk in Individuals, Populations, & Communities: Current Practices, Opportunities, and Risks

The National Academies of Sciences,
Engineering, and Medicine

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About Us



Facts and Figures 2015

Cincinnati Children's at a Glance

- Full-service, nonprofit pediatric academic medical center, established in 1883
- Our vision: to be the leader in improving child health
- 629 registered beds, including 95 inpatient behavioral-health beds and 36 residential behavioral-health beds
- Comprises the Department of Pediatrics of the University of Cincinnati College of Medicine
- #3 in pediatric research grants from the National Institutes of Health
- 850 volunteers contributed 90,263 hours at our facilities, an increase of 11% from the previous fiscal year

National Recognition

- Ranked #3 in the nation among all Honor Roll hospitals in *U.S. News & World Report's* 2015-16 Best Children's Hospitals list, including top 10 for all 10 pediatric specialties analyzed.
- Ranked #1 for pulmonology by *U.S. News* for the first time, and earned top-5 honors in cancer, nephrology, gastroenterology and GI surgery, orthopaedics and urology.
- Earned Magnet-designation (2013-2017) by the American Nurses Credentialing Center. Of the nearly 6,000 hospitals in the U.S., fewer than 7% have earned Magnet recognition.

Statistical Highlights (JULY 1, 2014 – JUNE 30, 2015)

Admissions (includes short stay).....	33,692
Emergency Department Visits	102,557
Outpatient Visits	
Primary Care	84,711
Specialty.....	852,899
Test Referral Center.....	76,486
Urgent Care	63,481
Surgical Procedures	
Inpatient	6,529
Outpatient.....	25,810
Surgical Hours.....	44,031
Patient Encounters	1,246,447

Faculty and Staff

Faculty	
Pediatrics	724
Surgery.....	103
Radiology.....	50
Anesthesia.....	59
Clinical Fellows	262
Research Postdoctoral Fellows	162
Residents.....	211
Active Medical Staff.....	1,789
Total Employees	15,260

Operating Revenues and Expenses (dollars in thousands)

Net Hospital Patient Services Revenue	\$1,527,936
Research Grants and Contracts.....	\$198,052
Total Operating Revenues	\$2,206,470
Total Operating Expenses	\$1,996,846
Available to Reinvest in the Mission	\$209,624



The importance of early identification

- Half of all psychiatric disorders that begin by 18 years of age are usually preceded by subsyndromal pathologies that evolve into a major psychiatric disorder.
 - These disorders account for 45% of the global burden of disease for individuals under 25 years 4.
- The cost of doing nothing is staggering—the WHO reports that mental illnesses are the leading cause of disability-adjusted life years (DALYs) worldwide, and \$2.5 trillion global with a projected increase to more than \$6T by 2030.
- We have done nearly fifteen studies, today I'll review two – suicide corpus construction and NLP in the emergency department World Health Organization.
- "Global health estimates: leading causes of DALYs." *World Health Organization. Retrieved 23 (2021).*

Definition: One DALY represents the loss of the equivalent of one year of full health. DALYs for a disease or health condition are the sum of the years of life lost to due to premature mortality (YLLs) and the years lived with a disability (YLDs) due to prevalent cases of the disease or health condition in a population.

The complexity of suicide

- “Suicide is a complex public health problem of global importance.
- Suicidal behavior differs between sexes, age groups, geographic regions, and sociopolitical settings, and variably associates with different risk factors, suggesting etiological heterogeneity.
- Although there is no effective algorithm to predict suicide in clinical practice, improved recognition and understanding of clinical, psychological, sociological, and biological factors might help the detection of high-risk individuals and assist in treatment selection.”
- DOI: [10.1016/S0140-6736\(15\)00234-2](https://doi.org/10.1016/S0140-6736(15)00234-2)

Some requirements for dealing with this complexity



3-D AI Representation of EHR
Estimated combinatorial size ---



High-performance computing for
training models



Highly training multi-dimensional team

Early identification can reduce impulsive suicide



You think you are so wonderful. This is for you. I took the pills for you. She means so much to you.

Mom its all wrong. I took 14 of your pills. They Mother I am so tired. Please forgive me. I can't decide anything any more. Daddy* please hurry home. Mom I'm so tired. Please forgive me for everything.

Source: Pestian, Shneidman, Leenaars Suicide Note Database

Sentiment corpus construction, annotation, and shared task

- Use notes to create a corpus for linguistic models
- Collect 1,319 notes written by people before they died by suicide.
 - 1950 - 2020 by Drs. Shneidman, Leenaars, and John Pestian
- Corpus construction began in 2009
 - CCHMC's Institutional Review Board (#2009-0664).
 - Each note was scanned into the Suicide Note Module (SNM)
 - The scanned notes were then transcribed to a text-based version by a professional transcriptionist.
 - Each note was then reviewed for errors by three separate reviewers.
- Hold a competition to develop machine learning algorithms
- DOI: [10.4137/BII.S10213](https://doi.org/10.4137/BII.S10213)

Annotations read and added emotions

- Annotators were asked to identify the following emotions:
 - abuse, anger, blame, fear, guilt, hopelessness, sorrow, forgiveness, happiness, peacefulness, hopefulness, love, pride, thankfulness, instructions, and information.
 - Emotion tokens were provided from Pubmed abstracts
 - Delphi analysis to rank
 - A special web-based tool was used to collect, monitor, and arbitrate annotator's activity. The tool collects annotation at the word and sentence level.

Upload
Transcription
Notes to review
Suicide note review
Review confirmation
Final check

Suicide note interpretation (Find emotions that are represented by independent sentences not by the whole passage) # 1

Page 1

Jan 1st 01
Dearest Mom . I will try to answer your letter . Thank you for your presents .
It was awful sweet of you to send them to me Mom .
Please forgive me for what I am going to do . I ca n't help it . Do n't feel
bad . I love you very much , dear , do n't ever forget that , and I love John
too , but I ca n't make him happy , and I broke my promise to him .
Good By Mom , and take good care of Jane . All my love
Jane

Click each word in an entire sentence where the writer is saying he or she *

☐ Was abused verbally, physically, mentally ...

☐ Is angry with someone ...

☐ Is blaming someone ...

☐ Is afraid of something ...

☐ Feels guilt ...

☐ Feels hopeless ...

☐ Feels sorrow ...

Click each word in an entire sentence where the writer is saying he or she

☐ Is forgiving someone ...

☐ Is feeling happy or peaceful ...

☐ Has hope for future ...

☐ Feels love for someone ...

☐ Feels pride ...

☐ Is thanking someone ...

Click each word in an entire sentence where the writer is

☐ Giving directions on what to do next (instructions) ...

☐ Giving practical information where things stand (information) ...

Most frequent sentiments found in suicide notes

Table 3

Frequency and example of assigned emotions.

Description	Frequency	Example
Instructions	609	Careful, cyanide gas in the bathroom
Hopelessness	601	I just didn't want to live anymore
Love	472	I love her
Information	430	I have no debts except for what my wife knows
Guilt	423	Forgive me please
Sorrow	342	Oh, how I suffer
Blame	235	I have been pushed around too much
Hopefulness	216	You will a happy and healthy life
Thankfulness	187	You, John have been so good to me and Jane
Anger	183	Well, Jane I hope this makes you happy!
Fear	154	I am terrified
Happiness/ peacefulness	119	I'm ready for the next step with joy and anticipation
Pride	89	We have another sweet little daughter
Forgiveness	61	I do not blame you for anything, my dear
Abuse	53	Life is so cruel when you are persecuted by in-laws and ex-wife

Shared task

- 35 International teams participated
 - Used the corpus to train and test their models for finding

Table 4. Team ranking using micro-average F_1 , precision and recall.

Team	F_1	Precision	Recall
Open university	0.61390	0.58210	0.64937
MSRA	0.58990	0.55915	0.62421
Mayo	0.56404	0.57085	0.55739
Nrcit	0.55216	0.55725	0.54717
Oslo	0.54356	0.60580	0.49292
Limsi	0.53831	0.53810	0.53852
Swatmrc	0.53429	0.57890	0.49607
UMAN	0.53367	0.56614	0.50472
Cardiff	0.53339	0.54962	0.51808
LT3	0.53307	0.54374	0.52280
UTD	0.51589	0.55089	0.48506
OHSU	0.50985	0.53351	0.48821
Wolverine	0.50315	0.45334	0.56525
TPAVACOE	0.50234	0.49922	0.50550
CLIPS	0.50183	0.51889	0.48585
SIP	0.49727	0.67429	0.39387
SRI & UC Davis	0.48003	0.49831	0.46305
DIEGO-ASU	0.47506	0.41791	0.55031
Ebi	0.45636	0.60077	0.36792
Duluth	0.45269	0.45985	0.44575
Columbia	0.43017	0.42125	0.43947
Pxs697	0.40288	0.37192	0.43947
Lassa	0.38194	0.35089	0.41903
Saeed	0.37927	0.37059	0.38836
SNAPS	0.35294	0.58684	0.25236
Senti6	0.29669	0.30532	0.28852

Registered team locations



Table 5. Examples of sentence/label combinations that were misclassified by all systems.

Error type	Text ID	Sentence	Annotator	System
False negative	200909031138 4664	"Goodbye my dear wife Jane."	love	none
False negative	200809091809 2119	"I ask God alone to judge my action."	guilt	none
False negative	200812181837 2227	"I hope something is done to John Johnson, for I do not wish to die in vain."	anger	none
False positive	200908201415 0445	"respectfully Mary P.S. I love you BABY."	none	love
False positive	200812181838 1506	"Dearest Jane I am about to commit suicide. x Please notify police that I am in the deserted garage at the top of Terrace in Cincinnati near the rose bowl."	none	instructions
False positive	200809091735 1923	"John: I can't take your cruel unkind treatment any longer."	none	hopelessness

doi.org/10.4137/BII.S9042

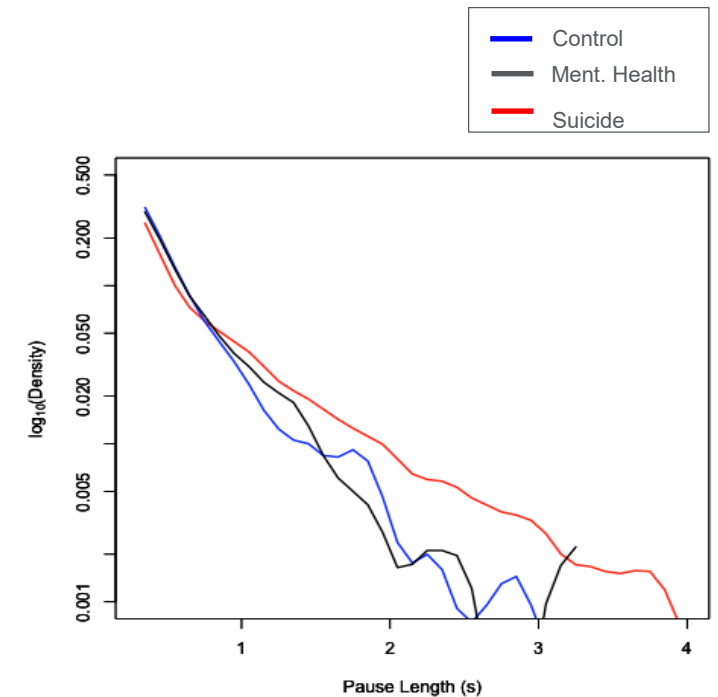
Next study: Linguistic and Acoustic Clinical Trial

- Enroll patients from ED
- collect linguistic, acoustic, video data from EHR
 - Interview with semi-structured ubiquitous questions to extract language for machine learning
 - Administered by clinicians during ED visit
 - *Do you have hope?*
 - *Do you have any fear?*
 - *Do you have any secrets?*
 - *Are you angry?*
 - *Where does it hurt emotionally?*
- Doi: [10.1111/sltb.12180](https://doi.org/10.1111/sltb.12180)



Audio & Facial Features

- Pauses Longer
 - Smaller vowel spacing
 - Smile less
 - Gaze down
-
- DOI: [10.1109/ICASSP.2015.7178880](https://doi.org/10.1109/ICASSP.2015.7178880)
 - DOI: [10.1109/FG.2017.96](https://doi.org/10.1109/FG.2017.96)



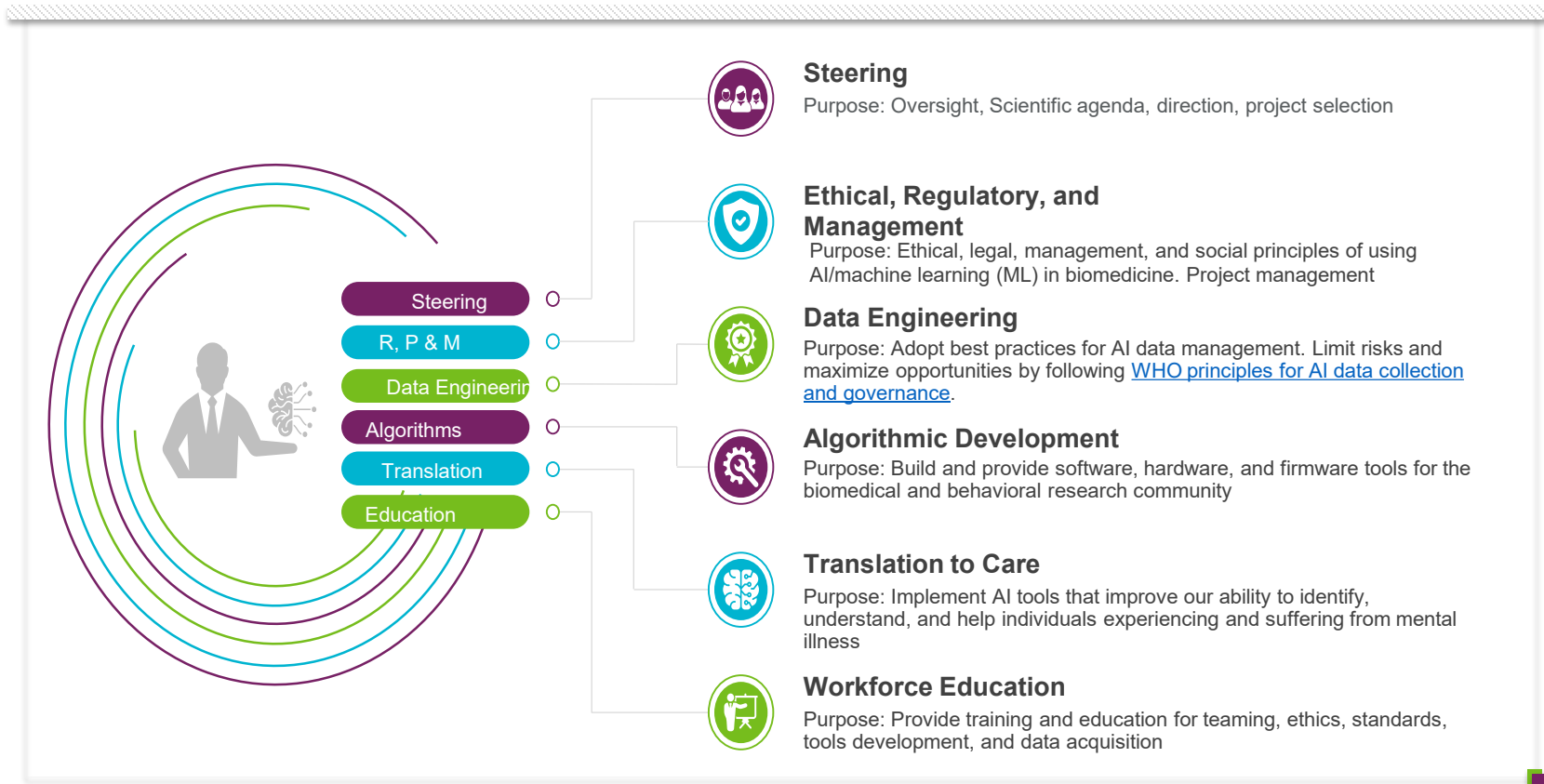
Results: machine learning linguistics

Table 1: The AROC for the machine learning algorithm. The non-suicidal group comprises of either mentally ill and control subjects. Class performances are shown for adolescents, adults, and the combined adolescent and adult cohorts.

	Suicidal vs. Controls			Suicidal vs. Mentally Ill			Suicidal vs. Mentally Ill and Controls		
	Adolescents ROC (SD)	Adults ROC (SD)	Adolescents+Adults ROC (SD)	Adolescents ROC (SD)	Adults ROC (SD)	Adolescents+Adults ROC (SD)	Adolescents ROC (SD)	Adults ROC (SD)	Adolescents+Adults ROC (SD)
Linguistics	0.87 (0.04)	0.91 (0.02)	0.93 (0.02)	0.82 (0.05)	0.77 (0.04)	0.79 (0.03)	0.82 (0.04)	0.84 (0.03)	0.87 (0.02)
Acoustics	0.74 (0.05)	0.82 (0.03)	0.79 (0.03)	0.69 (0.06)	0.74 (0.04)	0.76 (0.03)	0.74 (0.05)	0.80 (0.03)	0.76 (0.03)
Linguistics + Acoustics	0.83 (0.05)	0.93 (0.02)	0.92 (0.02)	0.80 (0.05)	0.77 (0.04)	0.82 (0.03)	0.81 (0.04)	0.84 (0.03)	0.87 (0.02)

- Next Steps: Spreading Activation Mobile (SAM)
- [Real Time Demo](#)

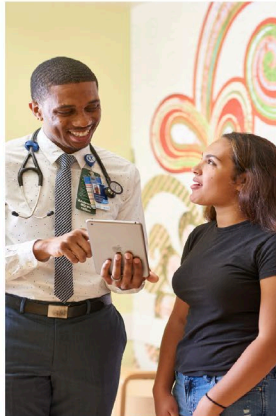
Future approaches to data collection



Data for suicide holistic measures
Clinical
Environmental
Social Determinant
Census
Pubmed
Social media

•
•

Clinical and High-Speed Computing Collaboration



Being in the right place at the right time, with the right tools, and delivering the right care

Scientific Goals

- Create a multimodal data repository for computing trajectories.
- Create methods for extracting all scientifically validated indicators of mental illness and use machine learning to discover new ones.
- Create and validate trajectories of depression, anxiety, and suicide.
- Determine how to operationalize the results to clinical settings.

Administrative Goals

- Establish the framework for data collection and sharing.
- Establish the necessary regulatory and legal framework.
- Establish data-sharing partnerships with other regional stakeholders.
- Establish new strategic partnerships collaboration.



Developing Mental Health Trajectories Using High-Performance Computing

A Collaboration Between Cincinnati Children's Hospital Medical Center and the United States Department of Energy's Oak Ridge National Laboratory

See backside for citation



What inspired this initiative?

Innovations in mental health treatment that use advanced technology lag the pace of innovations in medical care and basic science. We want to use advances in artificial intelligence to help identify children with mental illness earlier and provide families and caregivers with precision information that can be useful for treatment.

What are we hoping to accomplish?

Similar to pediatric growth charts that measure and plot trajectories of growth in height, weight, body mass, and head circumference, we will predict patient trajectories of anxiety, depression, and suicide prevention that are informed by genetic traits, current physical and mental health state, social determinants of health, and environmental exposure to known indicators of risk.

Through our unique national collaboration, we will use our clinical and computational expertise to develop near real-time visualizations of patient-specific mental health trajectories that enable earlier identification of patients with clinically-elevated symptoms and connect them to services, leading to better outcomes and quality of life.

By combining information that is regularly collected during healthcare encounters, we can use advanced artificial intelligence and machine learning methods to identify children at highest risk of developing a mental health concern over time. The first three conditions of focus are depression, anxiety, and suicidal ideation.

For more information, contact John Pestian, PhD, MBA (john.pestian@cchmc.org) or Tracy Glauser, MD (tracy.glauser@cchmc.org)

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Infographic of Statistics Abstracted from *Mental Health: By the Numbers* by the National Alliance on Mental Illness (NAMI). Online: <https://www.nami.org/nami/stats>. Copyright 2021 NAMI based on studies conducted by organizations like Substance Abuse and Mental Health Services Administration (SAMHSA), Centers for Disease Control and Prevention (CDC), and the U.S. Department of Justice.

BRV424008

A Multidisciplinary Collaboration

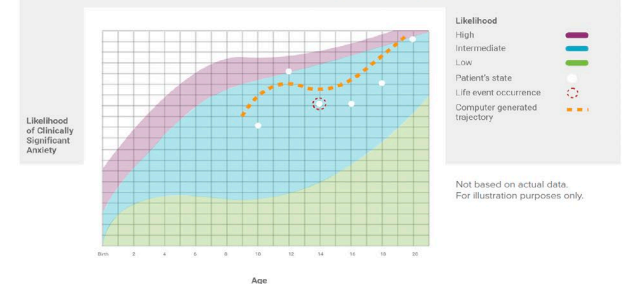


To develop a computation model for early detection of anxiety, depression, and suicide, Cincinnati Children's has partnered with Oak Ridge National Laboratories—the U.S. Department of Energy's largest science and engineering lab and home to the nation's most powerful supercomputer. From its roots in the Manhattan Project, Oak Ridge National Laboratory's mission continues to evolve at the frontiers of science.

Most recently, Oak Ridge National Laboratory has dedicated substantial resources to health sciences biomedical research and has partnered with governmental and non-governmental agencies like Cincinnati Children's, who, as collaborators, can impact the nation's health by using advanced computational tools and methods. Cincinnati

Children's and Oak Ridge National Lab have also established a Diversity Committee of clinicians and scientists who will ensure that the scientific approach, selection of data sources, and interpretation of findings reflect the diversity of experiences for children in our community and leads to improvements in care delivery and health for all youth, regardless of age, race, ethnicity, gender, sexual orientation, or economic background.

Patient's Anxiety Trajectory



What do we need to accomplish our objectives?

Amazing opportunities exist to gain insights into children's mental health trajectories using existing clinical data from the medical record. The lower costs of accumulating and accessing biological and clinical data, the availability of diverse datasets, and advances in artificial intelligence-based algorithms have ushered in a new era of research specifically focused on identifying children and families at high risk for mental health concerns, with opportunities for early intervention, prevention, and treatment.

While we have experience in trajectory computing, this collaboration allows us to scale-up and think about how to help children and families coping with mental health concerns regionally, nationally, and internationally.



Thank you.