Al for Social Good: Key Techniques, Applications, and Results

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Mission Statement: Advancing Al research driven by...



Grand Challenges of Social Work

- Ensure healthy development for all youth
- Close the health gap
- Stop family violence
- Advance long and productive lives
- Eradicate social isolation
- End homelessness





Overview of CAIS Project Areas

AI for Assisting Low Resource Communities Social networks: Spread HIV information

- Maximize influence under uncertainty
- Real-world pilot tests: Big improvements

AI for Protecting Endangered Wildlife



- Machine learning/planning: Anti poaching
- Scale, boundedly rational poachers,...
- Real-world: Uganda, South Asia...

Overview of CAIS Project Areas



- Game theory: security optimization
- Solve massive "security games"
- Real-world: US Coast Guard, FAMS...

Gangs, Substance abuse, Veterans mental health









- Social networks, robust optimization,...
- Behavioral models...
- Real-world: Research in progress



- Introduction
- HIV Information among homeless youth



Wildlife Conservation

AI Program: HEALER

Outline: HIV Information & Homeless Youth

- Domain of homeless youth and HIV information dissemination
- Real World Challenges in Influence Maximization



- POMDP Model and algorithms
- Pilot Study

Influence Maximization Background

- Input:
 - ➢ Graph G
 - Influence Model I
 - Choose K nodes per time step
 - Number of time steps for influence spread T
- Output:
 - K nodes per time step maximizing expected # influenced nodes

Independent Cascade Model

$$G = (V, E)$$

Propagation Probability (for each edge)



- Uncertain network state
- Uncertainty in network structure
- Adaptive selection

Challenge 1: Uncertain Network State



Challenge 2: Uncertain Network Structure



Independent Cascade Model

$$G = (V, E)$$
 $E = E_{cert} \cup E_{uncert}$

Propagation Probability (for each edge)



Existence Probability (for uncertain edges only)



HIV Prevention Programs: Using Social Networks to Spread HIV Information



Challenge: Adaptive selection in Uncertain Network



Challenge: Adaptive selection in Uncertain Network



Challenge 3 : Adaptive selection



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Pilot Study

POMDP Model

- Sequential decision making under uncertainty
 - Homeless shelters sequentially select nodes
 - Homeless shelters network state not known



Real world scale: Why is it hard to solve?



Real world scale: Why is it hard to solve?



Real world networks have community structure



Graph Partitioning



HEALER : Hierarchical Ensembling



Graph Partitioning: Why partition?



Real Networks - Solution Quality



Robustness & Parameter Uncertainty

HEALER: fixed propagation and existence probability



- Want policies robust to different possible values of P(A,B) and U(A,B)
 - Express as ranges of values, e.g., U(A,B) is in [0.4, 0.8]

HEALER++ Robustness & Parameter Uncertainty

Worst case parameters: a zero-sum game against nature



Payoffs: (performance of algorithm)/OPT

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Pilot Tests with 170 Youth in LA Area

Recruited youths:

HEALER	HEALER++	DEGREE CENTRALITY		
62	56	55		



Preliminary network —> HEALER

Bring 4 youth for training, get edge data —> HEALER

Bring 4 youth for training, get edge data —> HEALER

Bring 4 youth for training



Informed Non-Peer Leaders Who Started Testing for HIV

Testing Non-Testing



AI Program: HEALER





- 900 youth study begun at three locations in Los Angeles
 - > 300 enrolled in HEALER/HEALER++
 - > 300 enrolled in no condition
 - > 300 in Degree centrality
- IRB approvals
- Presenting video updates every few months



- Introduction
- HIV Information among homeless youth
- Wildlife Conservation



Protecting Wildlife in Uganda

Uganda Wildlife Authority & Wildlife Conservation Society



PAWS: Applying AI for protecting wildlife

Poacher Behavior Prediction

Predicting Poaching from Past Crime Data



Poacher behavior prediction

Queen Elizabeth National Park, Uganda 1300 targets, 12 years of patrol data



Initial Attempt Using Dynamic Bayes Net: Time Dependency & Imperfect Observation



Poacher behavior prediction

Poacher Behavior Prediction

Ensemble of Decision Trees





Poacher attack prediction

Poacher Behavior Prediction



Results from 2015



Real-world Deployment: Results

- Two 3 sq km patrol areas: Predicted hot spots with infrequent patrols
- Trespassing: 19 signs of litter etc
- Snaring: 1 active snare
- Poached Animals: Poached elephant
- Snaring: 1 elephant snare roll
- Snaring: 10 Antelope snares
- Hit rates (per month)
 - Ours outperforms 91% of months



Historical Base Hit Rate	Our Hit Rate		
Average: 0.73	3		

On-Going Experiments: Queen Elizabeth National Park

- Red: Group 1 (highest attack prediction rate)
- Yellow: Group 2
- Green: Group 3





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On-Going Experiments: Queen Elizabeth National Park

- Rangers followed poachers' trail; ambushed camp
 - > Arrested one (of 7) poachers
 - Confiscated 10 wire snares, cooking pot, hippo meat, timber harvesting tools.
- Indirect poaching signs; pursuit of poachers
- Signs of road building, fires, illegal fishing



PAWS: Applying AI for protecting wildlife

Game Theory + Poacher Behavior Prediction



Predicting Poaching from Past Crime Data

Towards the Future

- Significant potential: AI for low resource communities, emerging markets
 - Direction of AI research in our hands
- Not just applications; novel research challenges:
 - Fundamental computational challenges from use-inspired research
 - > Designing AI systems in society:
 - Interpretability
 - Maintaining human autonomy

- Methodological challenges:
 - Encourage interdisciplinary research: measures impact in society

AI for Social Good





THANK YOU

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POMDP Model: Challenges

- Number of states (node & edge uncertainty): ~ 2³⁰⁰
- Number of actions (N-choose-K): > 1,000,000,000
- Number of observations (Edges exist or not): Exponential



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Divide and Conquer

- Main idea:
 - Divide POMDP into smaller POMDPs
 - Combine solutions of smaller POMDPs
- Two different ways of dividing POMDPs
 - Using uncertain edges: PSINET
 - Using graph partitioning and sampling: HEALER

...And the Past



"...prize every invention of science made for the benefit of all"

Intermediate POMDP



Intermediate POMDP



HEALER++ Algorithm

- Computes an equilibrium strategy for this game
- Exponentially large strategy space: incremental generation, double oracle
- Under some conditions, provably converges to approx. equilibrium

	Params #1	Params #2	Params #3		Nature's oracle			
Policy #1	0.8, -0.8	0.3, -0.3	0.4, -0.4			Params #1	Params #2	
Policy #2	0.7, -0.7	0.5, -0.5	0.6, -0.6		Policy #1	0.8, -0.8	0.3, -0.3	
Policy #3	0.6, -0.6	0.4, -0.4	0.7, -0.7		Policy #2	0.7, -0.7	0.5, -0.5	
Influencer's oracle								
	Params #1	Params #2	Params #3					
Policy #1	0.8, -0.8	0.3, -0.3	0.4, -0.4		1			
Policy #2	0.7, -0.7	0.5, -0.5	0.6, -0.6					

Real networks - robustness

Worst case % of optimal influence

