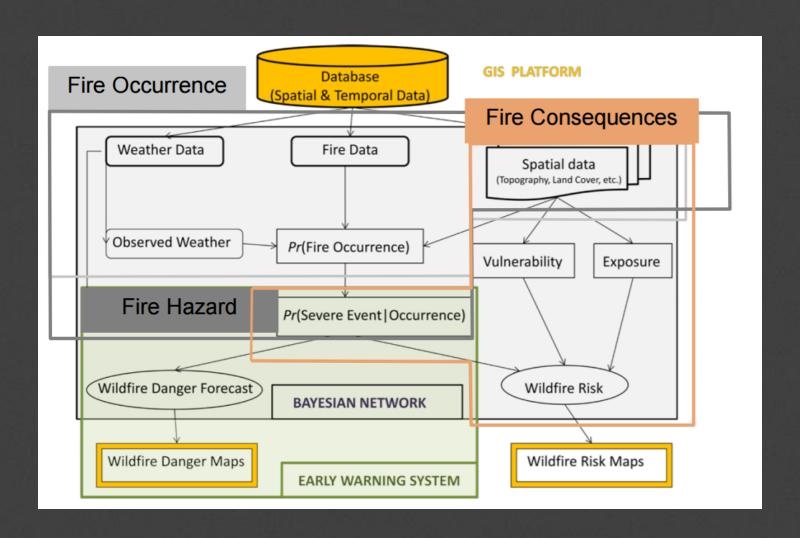


組員 | 鄭余玄 謝昀佐 陳令原

Real case



Workflow

- Find a dataset
- Modal each node
- Form BBN
- MCMC train parameters
- Validation

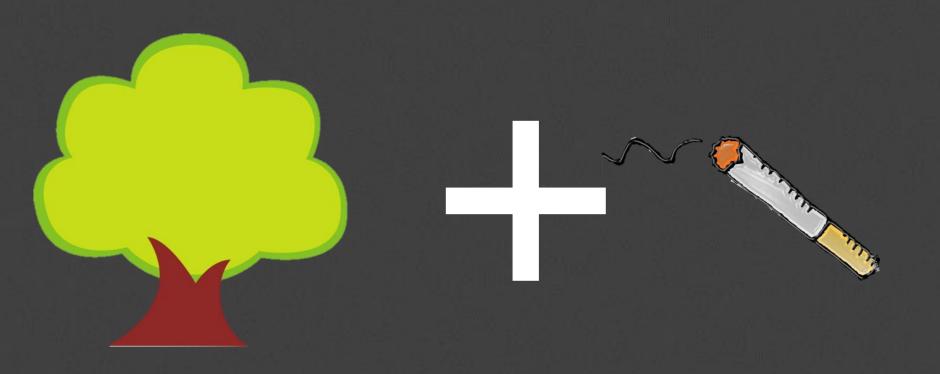
Dataset

- UCI Machine Learning Repository
- Forest Fires Data Set
- About 500 data

	Е	F	G	Н	1	J	K	L	М	
	FFMC	DMC	DC	ISI	temp	RH	wind	rain	area	
5	86.2	26.2	94.3	5.1	8.2	51	6.7	0	0	
2	90.6	35.4	669.1	6.7	18	33	0.9	0	0	
5	90.6	43.7	686.9	6.7	14.6	33	1.3	0	0	
5	91.7	33.3	77.5	9	8.3	97	4	0.2	0	
7	89.3	51.3	102.2	9.6	11.4	99	1.8	0	0	
7	92.3	85.3	488	14.7	22.2	29	5.4	0	0	

Reason?

Most people thinks Forest fire is due to this~



Result



But, actually in our real world, that scene like this~



How about the long-distance view?



 Actually, this picture is what we need to discuss and analysis well.

What do u have in mind?

• And what is our aim?



Definition

for some measurement

- FFMC FFMC index from the FWI system: $(18.7 \sim 96.20)$
- DMC DMC index from the FWI system: $(1.1 \sim 291.3)$
- DC DC index from the FWI system: $(7.9 \sim 860)$
- ISI ISI index from the FWI system: $(0.0 \sim 56.10)$
- temp temperature in Celsius degrees: $(2.2 \sim 33.30)$
- RH relative humidity in %: $(15.0 \sim 100)$
- wind wind speed in km/h: $(0.40 \sim 9.40)$
- rain outside rain in mm/ m^2 : (0.0 \sim 6.4)

Some words are hard for U?

Definition

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Fine Fuel Moisture Code(FFMC)

- Numeric rating of the moisture content of litter and other cured fine fuels.
- This code is an indicator of the <u>relative ease of</u> <u>ignition</u> and the <u>flammability of fine fuel</u>.

Duff Moisture Code(DMC)

 Numeric rating of the average moisture content of loosely compacted organic layers of moderate depth.

 Gives an indication of <u>fuel consumption</u> in moderate duff layers and medium-size woody material.

Drought Code(DC)

 Numeric rating of the average moisture content of deep, compact organic layers.

 A useful indicator of <u>seasonal drought effects on</u> forest fuels and the amount of <u>smoldering in deep</u> duff layers and large logs.

Initial Spread Index(ISI)

Numeric rating of the expected rate of fire spread.

 Combines the effects of wind and the FFMC on rate of spread without the influence of variable quantities of fuel.

Aim

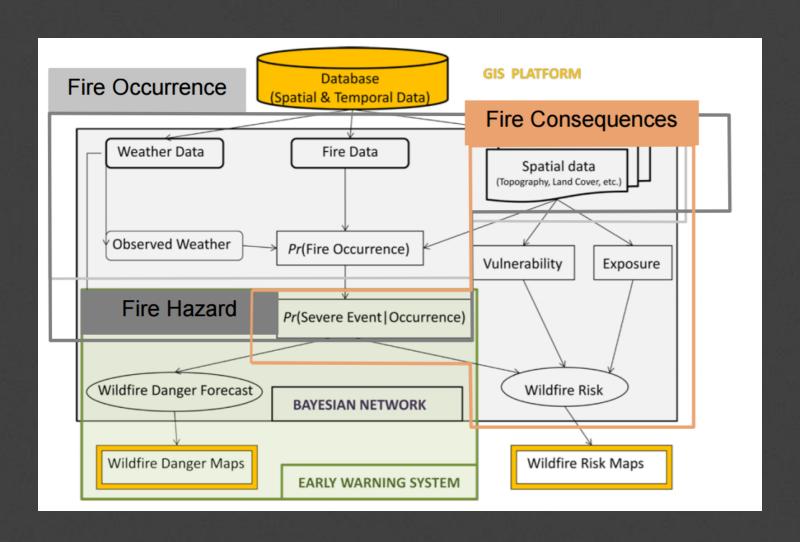
 Using those indicators to estimate how big & large of the "Burnt Area."

 Once we can approach the estimation of the burnt area, then we can assign and make best use of our human power (ex: firefighter) and natural resource like (ex: fire engine, water and helicopter...etc)

Observations

- Each factors are a continuous random variable.
- How are those factors correlated?
- What is the casual network?

Real case



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Dataset

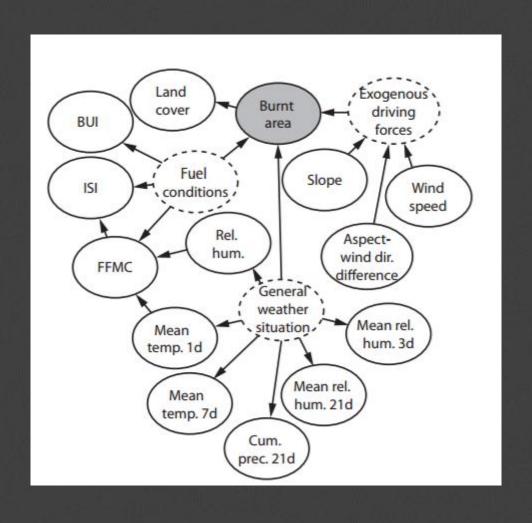
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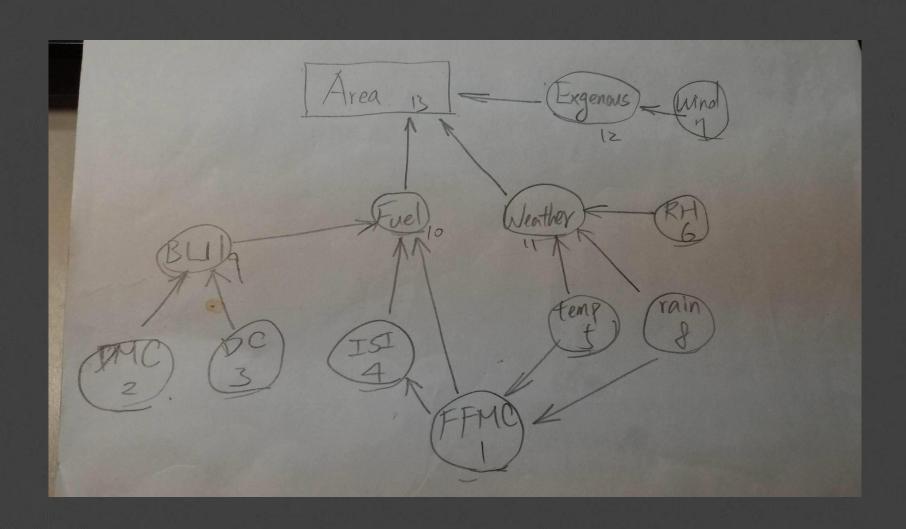
Modal each node

- Continuous Random Variable
- Quantization / Discretize
- CPD -> CPT

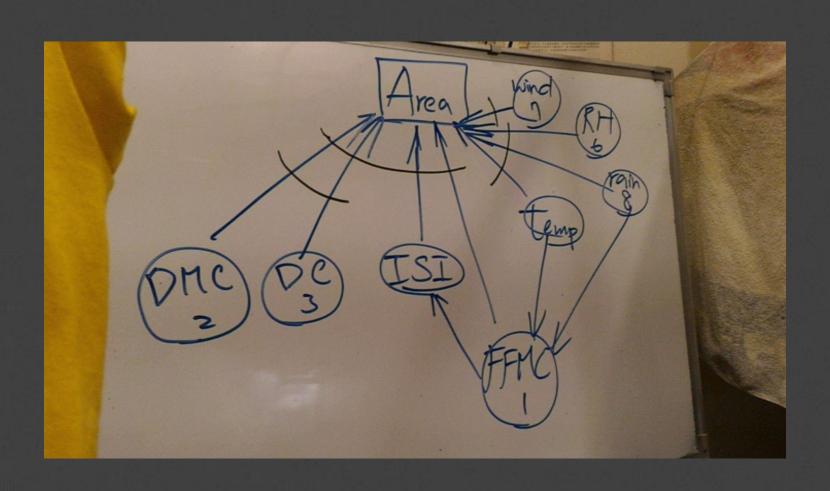
BBN theoretically



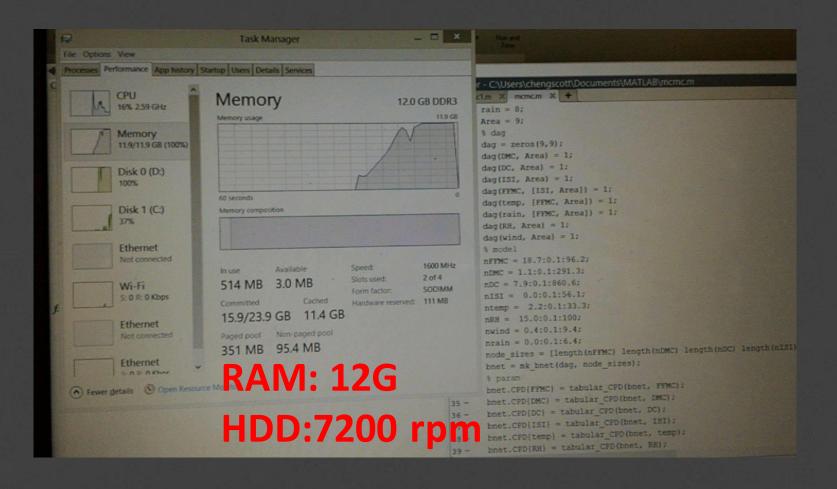
BBN Formulation (I)



BBN Formulation (II)



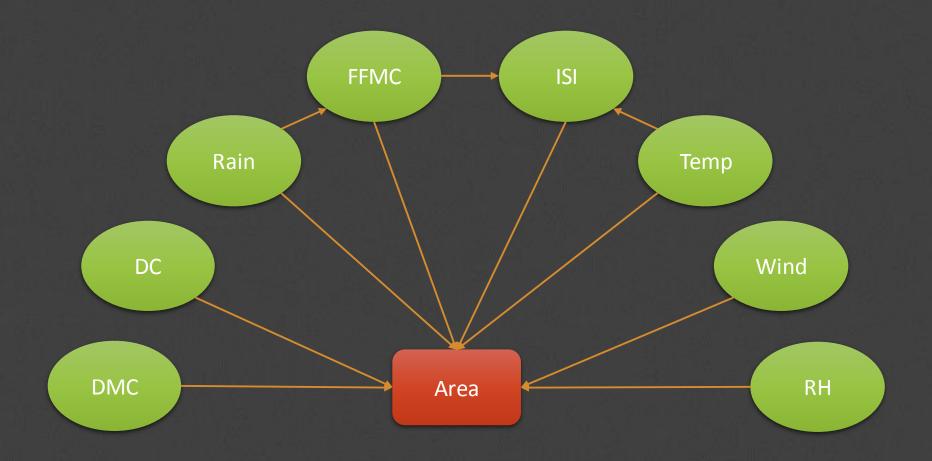
Experiments: Crashes



Difficulties

- Computation resources
- Sparse DAG

BBN reduction



Bayesian Probability

•
$$P(x|y) = P(x,y) \times P(y)$$

•
$$P(y|x) = \frac{P(x|y)P(y)}{P(x)} = \frac{P(x|y)P(y)}{\sum_{y} P(x|Y = y)P(Y = y)}$$

Bayesian Network

• A network which describe probability of cause-effect relation. For given Bayesian network(BN), one of the sample like (x1,...xn), its probability could be given by

Metropolis sampler

- Propose a transition with probability $TQ(y \rightarrow y')$, where y is current state, and y' is next state
- Accept with probability A=min(1, P(y')/P(y))
- If for all y, y' $TQ(y \rightarrow y') = TQ(y' \rightarrow y)$ then the resulting Markov chain satisfies detailed balance

Markov Chain Monte Carlo

```
Algorithm MCMC-Ask(X,e,bn, N) returns an estimate of P(X|e)

local variables: N[X], a vector of counts over X, initially zero

Z, the nonevidence variables in bn

x, the current state of the network, initially copied from e.

initialize x with random values for the variable for the variables in Z

for j=1 to N do

N[x] = N[x] + 1 where x is the value of X in x

for each Zi in Z do

sample the value of Zi in x from P(Zi | mb(Zi)) given the value of MB(Zi) in X

return Normalize(N[X])
```