



Docitive Networks

A Step Beyond Cognition

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Baseline Cognition

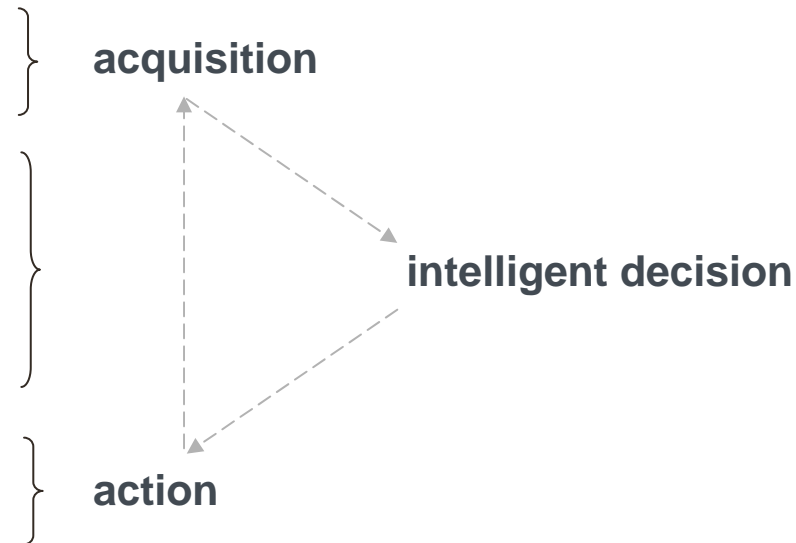
Formal Definition of Cognition

- A cognitive system “involves perception; knowing, remembering; thinking, judging, problem solving; and intelligent actuation” [1]:

Formal Definition

- perception
- knowing
- remembering
- judging
- problem solving
- actuation

Cognitive Radio Cycle



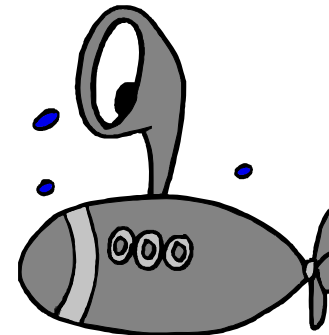
Alternative Definition of Cognition

- A cognitive system “a system which is working properly under conditions it was initially not designed for” [2]:

- humans fly:



- humans dive:



- humans think:

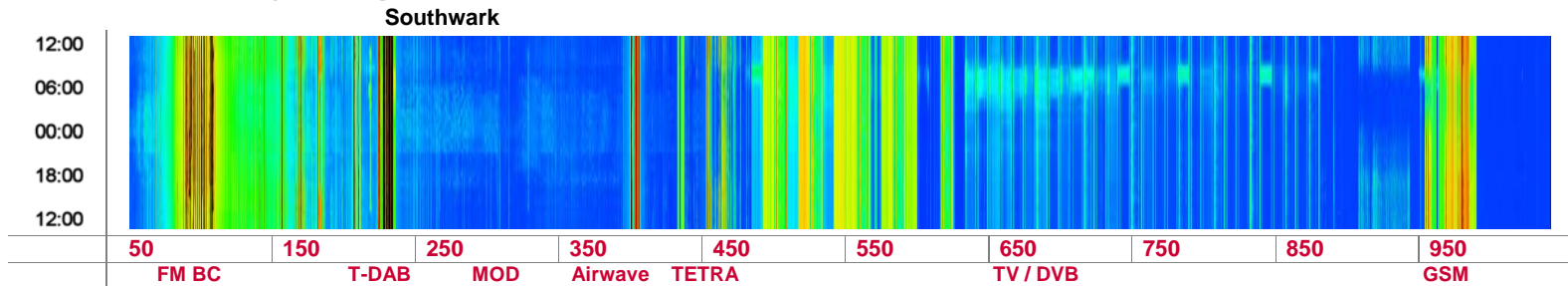


Misleading Connotation of Cognition

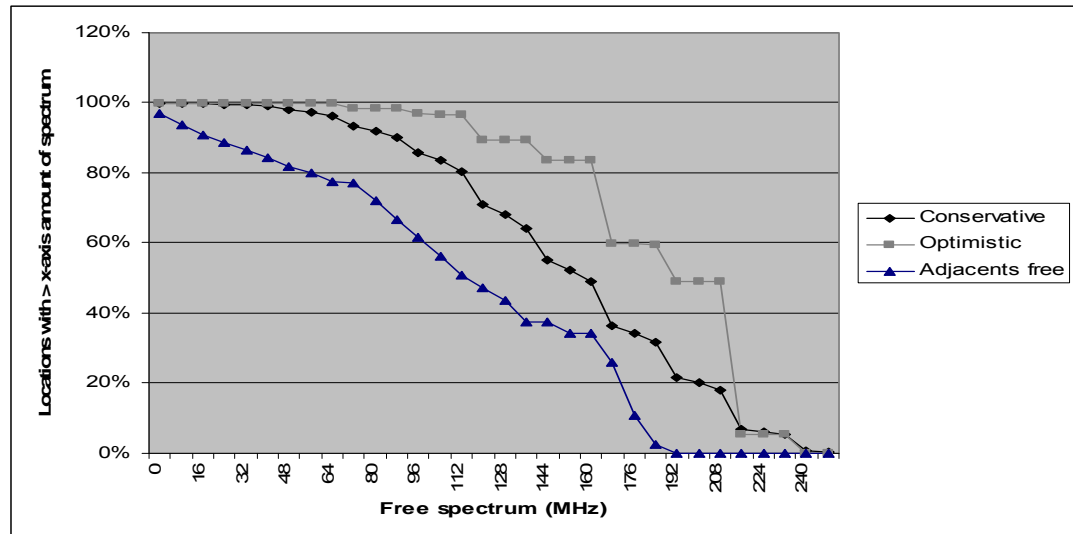
- An estimated 95-97% of publications containing the term “cognitive” are actually opportunistic at best.
- The main thrust of all these publications has been along ...
 - spectrum is scarce in general - is it? -
 - thus let's use spectrum holes - many? -
 - for this we need spectral sensing - useful? -
 - then act by comparing thresholds - new ? -
- Clearly not intentional, but “cognitive” is often misleading:
 - once a fashionable buzzword for getting a proposal funded
 - impressive transition to hype and then overripe well before actually used
 - waning reputation of cognitive makes it difficult to support subsequent research

Example: Interleaved Spectrum

- Apparent large amount of spectrum unused [3]:



- However [3]:



Example: Interleaved Spectrum

- Study conducted by Ofcom concludes [3]:
 - benefits estimated to have value of £200m-£300m
 - DTT e.g. has an equivalent value in the region of £50bn
 - thus maximum of 0.5% probability of interference
- Methods potentially guaranteeing this interference probability:
 - spectral sensing
 - global beacon channels
 - geolocation → preferred solution
- Problems with sensing [3]:
 - hidden node sensing margin is up to 35dB
 - this is beyond any non-cooperative sensing techniques
 - cooperative sensing not answer either (unreliability/costly planning)

Overview

- 1. Baseline Cognition**
- 2. From Cognition to Docition**
- 3. Wireless Multi-Agent Systems**
- 4. Vision & Challenges**
- 5. Conclusions**

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From Cognition to Docition

A Case For Docitive Systems

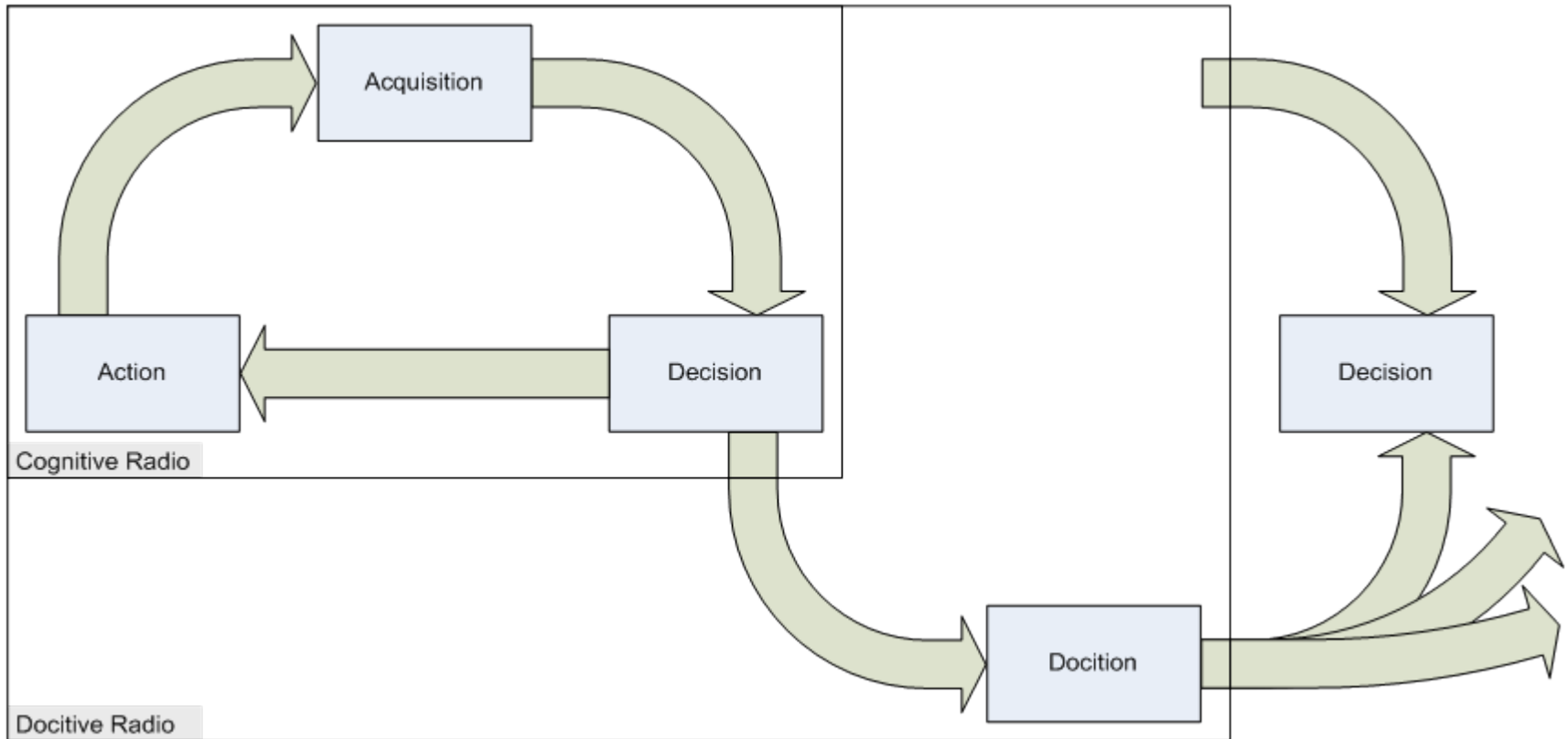
- State of human cognition heavily depends on teachers encountered during one's life, who generally impact:
 - learning space
 - learning speed
 - teaching abilities
- Concept of Docitive Systems is inspired by so-far-successful Problem Based Learning (PBL) concept:
 - mimics the well-functioning society-driven teacher-pupil paradigm
 - introduce rigorous framework for above observations, where

radios are encouraged to teach other radios

- with the aim to significantly improve performance of current (cognitive) systems
- origin is from “docere” = “to teach” (“cognoscere” = “to know/learn”)

Docitive System Cycle

- Extension of Cognitive Cycle by Docition:



Cognitive Part of the Cycle

■ Acquisition:

- individual and/or collaborative sensing;
- docitive information from neighboring nodes;
- environmental/docitive information from databases; etc.

■ (Intelligent) Decision:

- core of a cognitive radio which learns and draws decisions;
- majority today are simple opportunistic decision-making algorithms;
- more sophisticated unsupervised, supervised or reinforcement learning available.

■ Action:

- ensures that the intelligent decisions are actually carried out;
- typically handled by a suitably reconfigurable software defined radio (SDR);
- also through policy enforcement protocols, among others.

Docitive Part of the Cycle

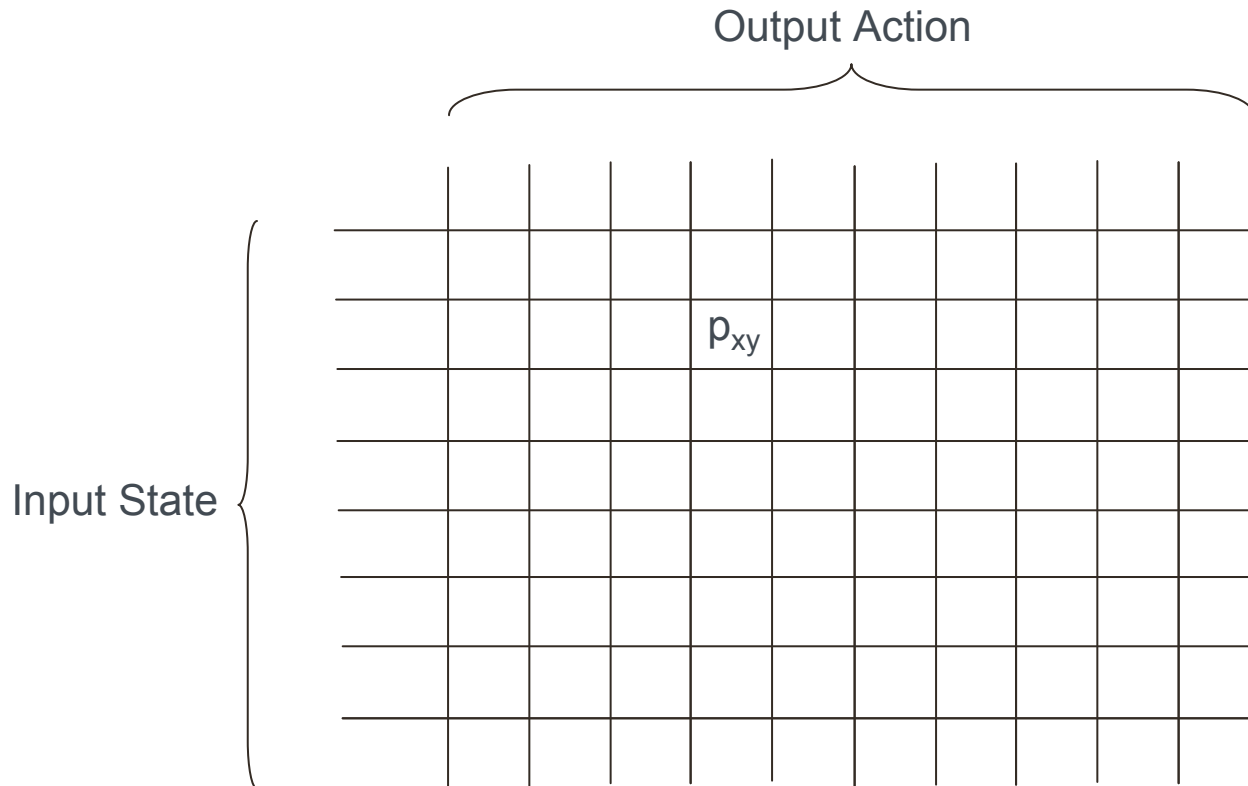
- Concepts borrowed from Problem Based Learning (PBL):
 - proponents: Lev Vygotsky, John Dewey, Jean Piaget, Michael Gardener, etc.
 - teachers are encouraged to be coaches – not information givers;
 - pupils work as a team using critical thinking to synthesize and apply knowledge; they apprehend through dialogue, questioning, reciprocal teaching, and mentoring
- Docition:
 - realized by means of entity facilitating knowledge dissemination and propagation;
 - paradigm comprising dissemination of information which facilitates learning.
- State-of-the-art:
 - sharing of end results (e.g. cooperative sensing or central database);
 - multi-agent systems (machine learning community);
 - distributed artificial intelligence (AI community).

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Wireless Multi-Agent Systems

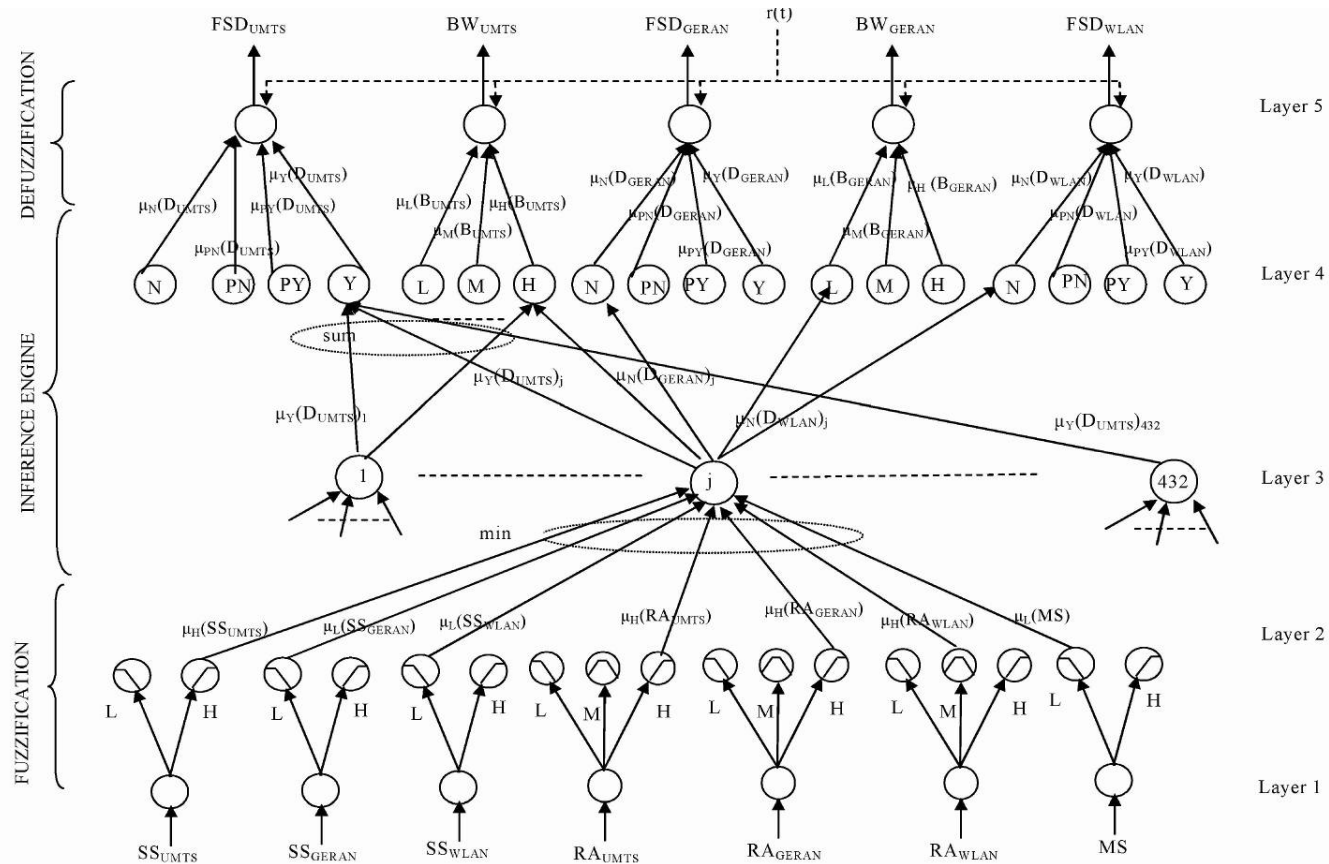
Single-Agent Baseline System

- For given environmental state-action space find transition probabilities such that reward is maximized:



Single-Agent Baseline System

- For given environmental state-action space find transition probabilities such that reward is maximized:



Multi-Agent Systems

- Q-Learning, being a typical learning mechanism for single agent systems, can be adapted to distributed settings:
 - implementation of decentralized Q-learning;
 - training process is extremely complex for increasing state-action space;
 - nodes thus could learn some disjoint or random parts of the state-action space;
 - this facilitates learning but does not yield the end-result *per sé*.
- We propose investigation into different degrees of cooperation, essentially trading cognition versus docation:
 - independent learners
 - cooperative learners
 - team learners

Degree of Cooperation

■ Independent Learners:

- nodes do not cooperate since they ignore actions and rewards of other nodes;
- they learn their strategies independently.

■ Cooperative Learners (different degrees):

- independent learning but sharing of instantaneous information about states;
- share sequences of state, action/reward/learned state-specific decision policies;
- perform joint tasks yielding longer learning but less oscillations.

■ Team Learners:

- multi-agent system is regarded as a single agent in which each joint action is represented as a single action;
- optimal Q-values for the joint actions are learned using single-agent Q-learning;
- no communication is needed between the nodes but they all have to observe the joint action and all individual rewards.

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Vision & Challenges

Quantifying Cognition & Docition

- Quest for viable measure for “intelligence”:
 - attempted across numerous domains over past centuries
 - no generic answer as most are application tailored (e.g. IQ test)
- There is a common trait, however:
 - intelligence is related to ability to bring order from seeming disorder
 - example of quasi-uniformly distributed random numbers:

16	3	2	13
5	10	11	8
9	6	7	12
4	15	14	1

- Universal quantifying measure for order & disorder: **entropy**

Entropy as Guiding Metric

- We draw the following qualitative observations:
 - “stupid” radio: increases disorder at output w.r.t. input
 - “clever” radio: decreases disorder at output w.r.t. input
- Straightforward rough quantitative formulation:

$$\text{intelligence} = \text{input entropy} - \text{output entropy}$$

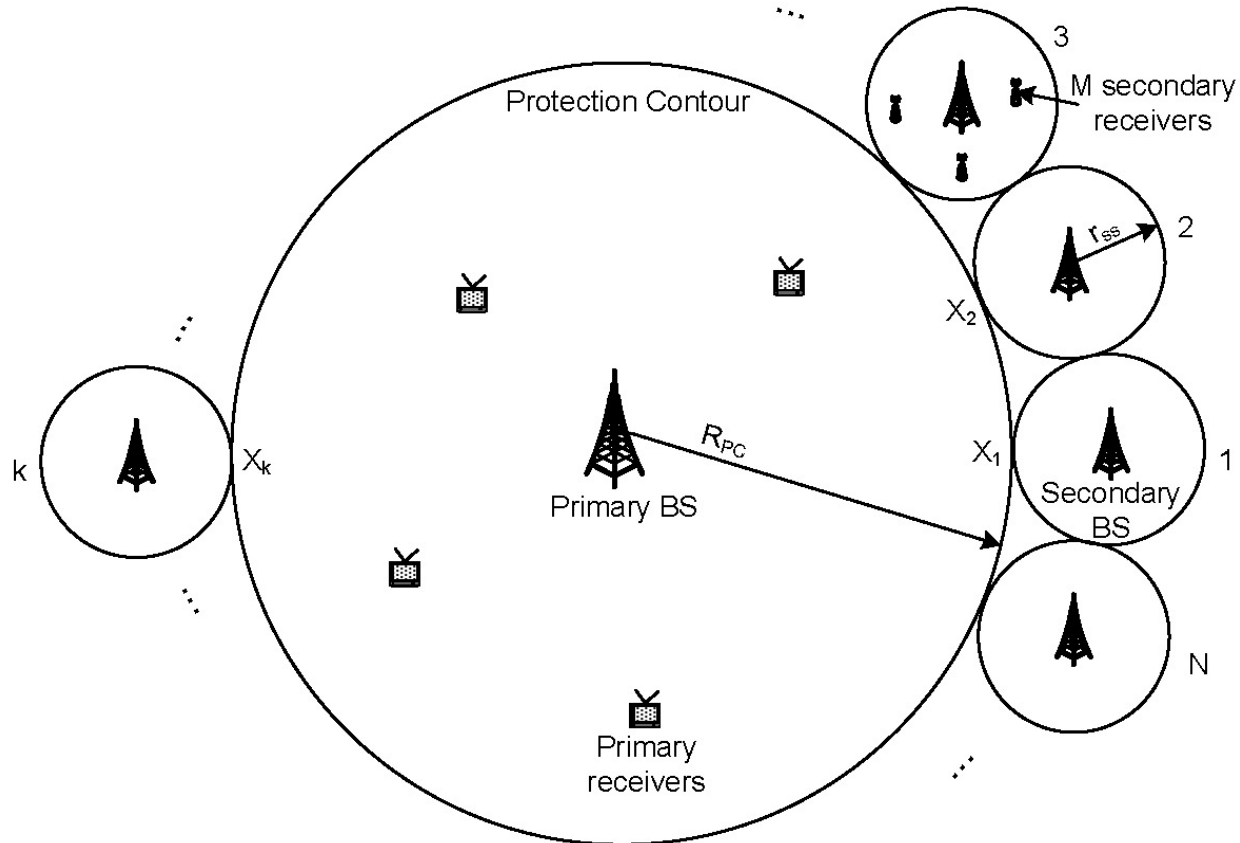
- Direct implications onto docitive systems:
 - input and output entropies are easy to measure and observe in a real system
 - facilitates the establishments of intelligence gradients in a system
 - allows establishing teaching costs along these gradients
 - docition should follow the steepest gradient

Entropy as Guiding Metric

- The following example metrics pertain to a single radio:
 - **current intelligence** = actual input entropy – achieved output entropy
 - **maximal intelligence** = max. input entropy – minimal output entropy
 - **learning ability** = (maximal – current) intelligence
- The following example metrics pertain to the network:
 - **intelligence gradient** = Δ of current intelligence between nodes
 - **degree of docition** = sum over intelligence gradient / number of nodes
- Example configurations:
 - selfish power control: Gaussian distribution $\rightarrow I_1$
 - cognitive radio with SINR region: uniform distribution $\rightarrow I_2 > I_1$
 - cognitive radio with SINR target: Dirac-delta distribution $\rightarrow I_3 > I_2 > I_1$

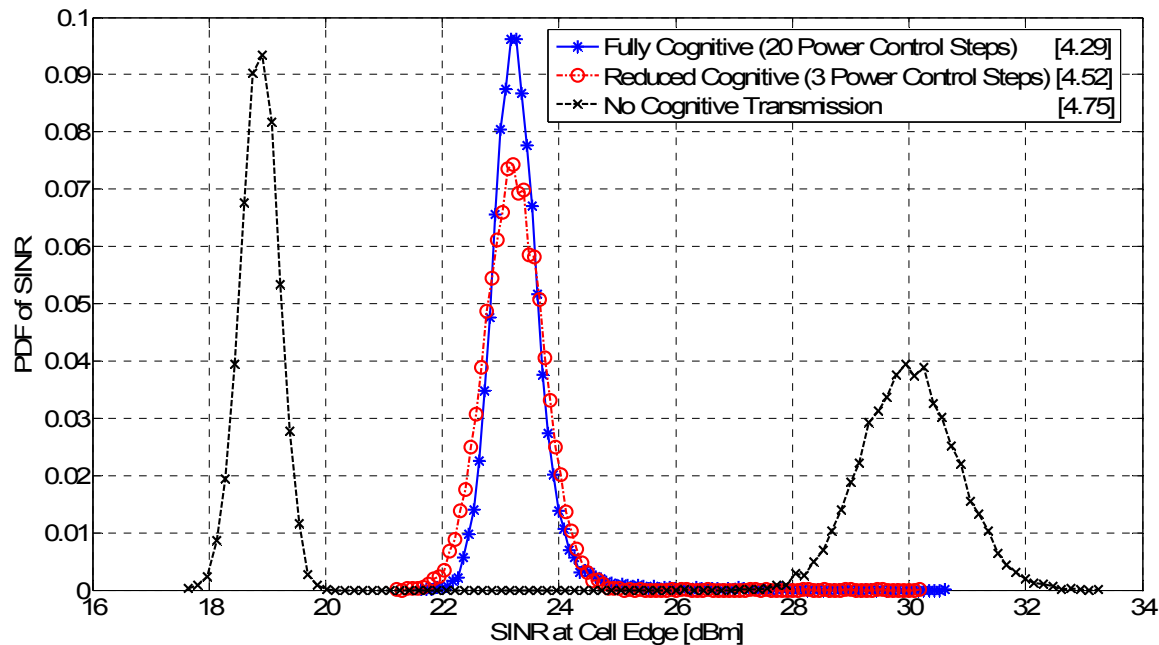
Example Cognitive Wireless System

- Primary DTV receivers in protection zone, surrounded by secondary cognitive users with constraints on SINR levels:



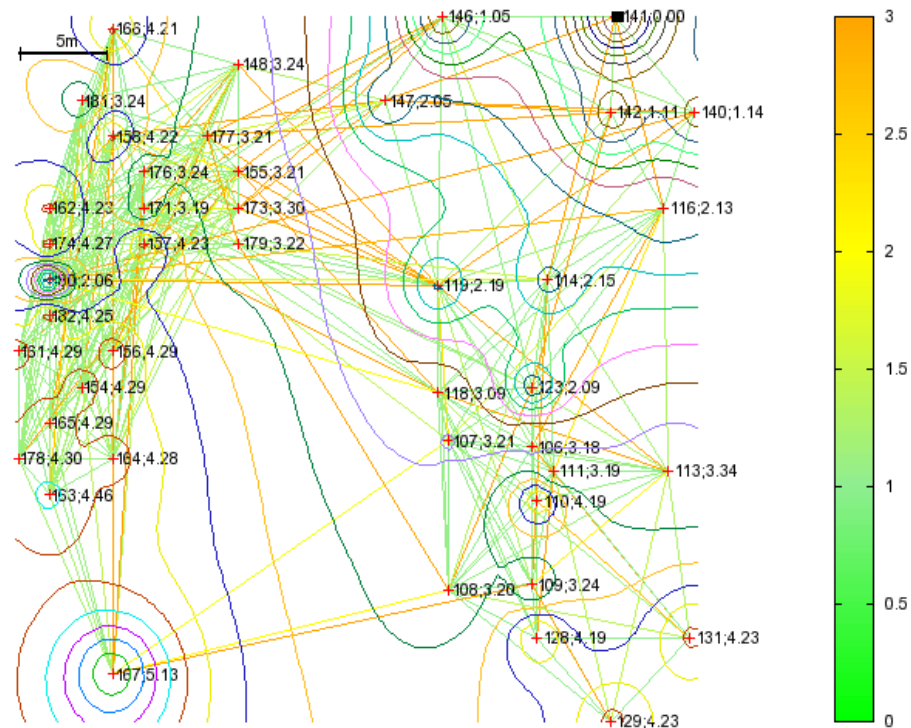
Example Cognitive Wireless System

- Different degrees of cognition yield different intelligence levels:
 - full cognitive scenario with 20 power control steps (entropy = 4.29)
 - reduced cognition with 3 power control steps only (entropy = 4.52)
 - no cognition but only selfish power control (entropy = 4.75)



Example Cognitive Wireless System

- Building cognitive gradients by incorporating:
 - gradient of intelligence
 - mapping it to potential gains
 - weigh it against cost of cooperation



Challenging Open Problems

- Numerous interesting and pertinent research areas open up, such as:
 - Information Theory: How much side information needs to be taught to pupils?
Impact of feedback, renewal rate, etc.?
 - Wireless Channel: What are the coherence times of the channel?
Do they allow sufficient time for learning/teaching?
 - PHY Layer: How much rate/energy should go into teaching?
Which PHY states should be taught?
 - MAC Layer: Can we re-use known broadcast approaches?
Which MAC states should be taught?
 - System: What is the optimal ratio teachers versus pupil?
What is the optimal teaching schedule?
Should every pupil also be teacher?
What exactly is best taught?
Facilitator of emergent behavior?

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Conclusions

Conclusions

■ Cognitive Systems:

- very well but fairly generically defined
- little truly cognitive systems available today

■ Docitive Systems:

- yet another generic concept with room for (mis)interpretation
- framework with no claim for total novelty and open for discussions
- “bad teacher teaches end-result – good teacher facilitates learning”
- CD + DR = more efficient radio

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