

Computational Social Cognitive Neuroscience

Leroy Todd



THE UNIVERSITY OF
TENNESSEE
KNOXVILLE

What is Social Cognitive Neuroscience?

- Encompasses any cognitive process that deals with conspecifics, either in a group scenario or one-on-one
- Are the key cognitive processes that govern language, perception, memory/attention also responsible for social interaction and formation?
- What governs the social aspect of our lives?

Social Learning

- Learning is more than a person stumbling through actions themselves until they get better or passively taking in information to build self-organized structures
- Mimicry or mirroring is a heavily studied area of social cognitive neuroscience

Mirror Neurons

- Neurons primarily in the premotor cortex that activate when observing actions of other conspecifics
- Activate in the subject as if it was the subject itself doing the action
- Research indicates these neurons distinguish from biological vs non-biological actions

Learning From Others

- Vicarious Reinforcement Learning is a learning strategy where the subject observes the actions and outcomes of someone else.
- Predictions seem to take place in the dlPFC while outcomes predicted in vmPFC

$$V = V_{t-1} + \alpha \delta_t$$

Judging Others Intentions

- Action Imitation as a learning mechanism can work similarly to vicarious reinforcement learning
- Work or Shirk: can shirk for gain but need to think about the intentions of others.
- Self choice held in vmPFC while non-executed choices held in dmPFC

$$I_{t+1} = I_t * E_t$$

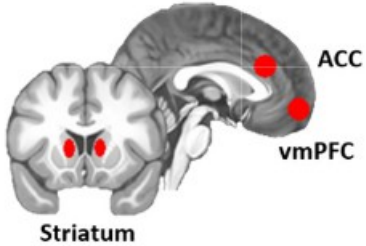
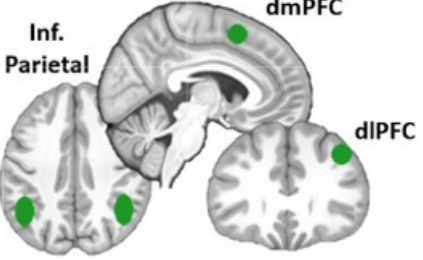
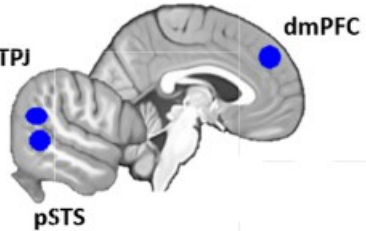
Learning About Others

- Learning about others is different than passively observing actions
- One game players paired with an advising confederate picked cards with hidden value

$$L_{t+1} = L_t + \beta \Delta_t$$

- Subjective influence counts too. Over several studies, things that violated 'social norms' saw increased activity in the ACCg

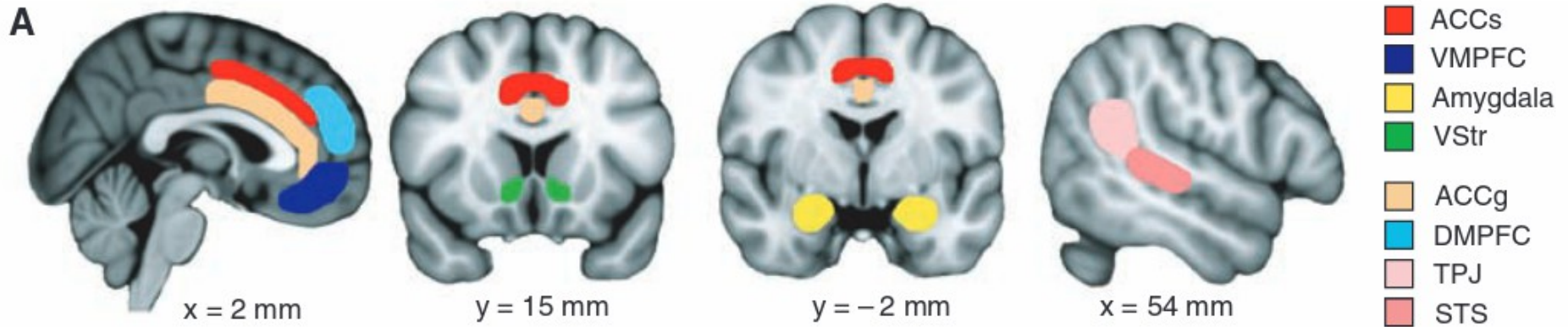
Building A Computational Model

Strategy	Vicarious reward learning	Action imitation	Bayesian inference
Generic learning rule	$Value_{t+1} = Value_t + \alpha_v * oRPE$	$Action_{t+1} = Action_t + \alpha_v * APE$	$Intention_{t+1} = Intention_t * Evidence_t$ (Posterior = Prior * Likelihood)
Possible computation	$oRPE = \text{other person's actual reward} - \text{expected reward}$	$APE = \text{other person's action} - \text{predicted action}$	Bayesian update = $Intention_{t+1} - Intention_t$
Main neural correlate	 <p>ACC vmPFC Striatum</p>	 <p>dmPFC dIPFC Inf. Parietal</p>	 <p>dmPFC pSTS TPJ</p>
Example behaviors	<ul style="list-style-type: none"> Reward & punishment learning Learning preferences, choices and attitude of others 	<ul style="list-style-type: none"> Motor learning Learning sequences of actions Reward & preference learning when outcome unavailable or inference strategy too demanding 	<ul style="list-style-type: none"> Learning other people's goals and intentions Strategic and competitive interactions Integrating multiple social signals (status, confidence, expertise, attitudes, group size, decisions, etc)
Pros & Cons	<div style="border: 1px solid black; border-radius: 15px; padding: 10px;"> <p> ↗ Computationally easy ↗ Maps onto RL framework </p> <p> ↘ Slow learning ↘ Inflexible </p> </div>		<div style="border: 1px solid black; border-radius: 15px; padding: 10px;"> <p> ↗ Flexible (high accuracy) ↗ Fast learning </p> <p> ↘ Computationally demanding ↘ Risk of overfitting </p> </div>

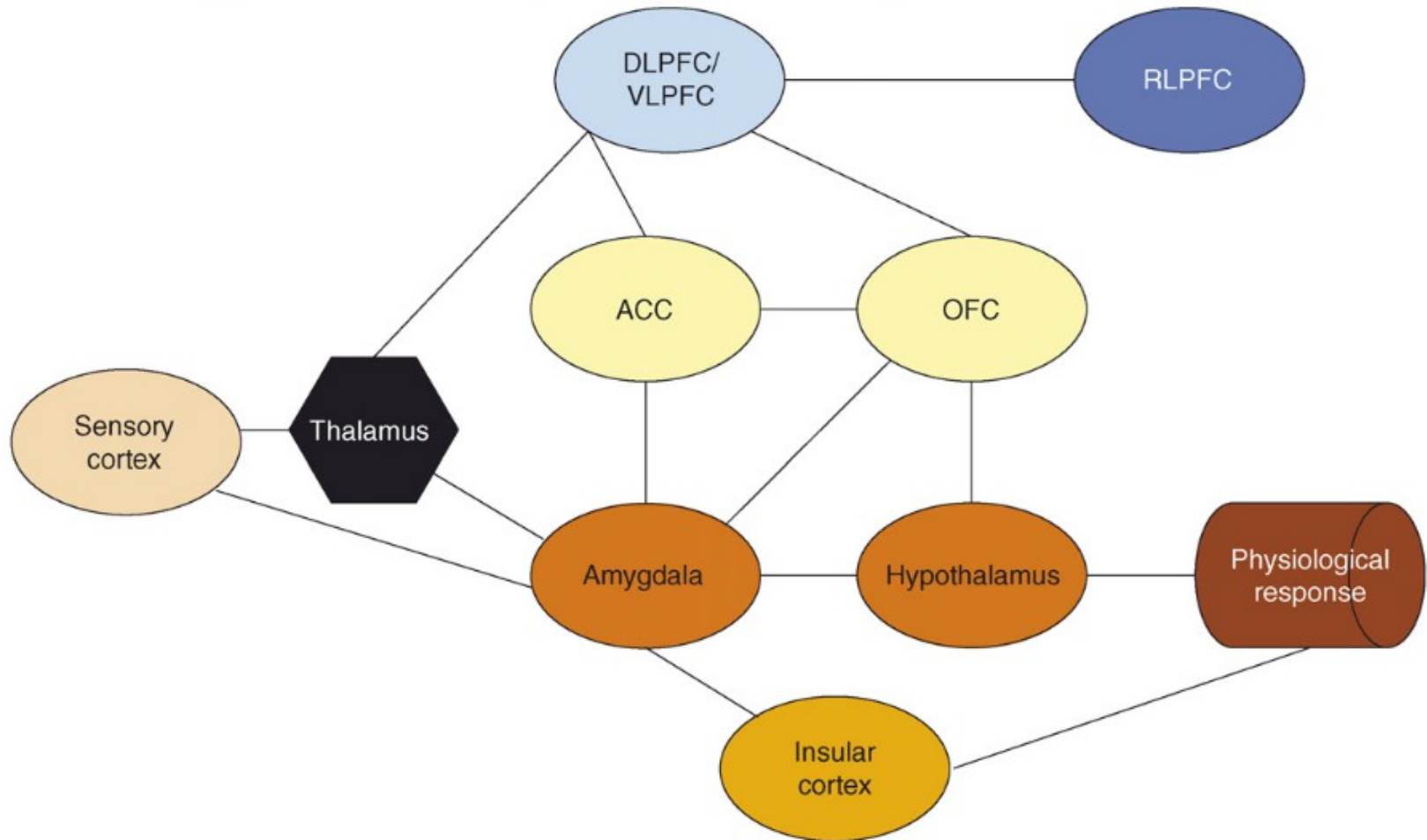
Reflexive and Reflective Systems

- Systems which govern how we think, react, judge, and otherwise function at a very high level
- Reflexive: fast operating, slow learning, bidirectional, parallel processing
- Reflective: slow operating, fast learning, symbolic

Circuitry of Multiple Neural Regions



Neural Model of Evaluation



TRENDS in Cognitive Sciences

Conclusion

- This “circuit” in the brain is a very, very rough model but has a lot of influence
- A lot more work needs to be done before we reach a computational model on the level of Emergent/Leabra
- Not even close to answering higher-order self-organization principles

References

Caroline Juliette Charpentier & John P. O'Doherty (2018): The Application of Computational Models to Social Neuroscience: Promises and Pitfalls, *Social Neuroscience*, DOI: 10.1080/17470919.2018.1518834

Cheong, J. H., Jolly, E. , Sul, S. and Chang, L. J. (2017). Computational Models in Social Neuroscience. In *Computational Models of Brain and Behavior*, A. A. Moustafa (Ed.). DOI: 10.1002/9781119159193.ch17

Erratum: Social cognitive neuroscience: where are we heading? *Trends in Cognitive Sciences*, Volume 8, Issue 7, July 2004, Pages 293

Behrens TE, Hunt LT, Rushworth MF (2009): Computation of Social Behavior, DOI: 10.1126/science.1169694

Ajay B. Satpute, Matthew D. Lieberman (2006): Integrating automatic and controlled processes into neurocognitive models of social cognition. DOI 10.1016/j.brainres.2006.01.005

Cunningham WA, Zelazo PD (2007): Attitudes and evaluations: a social cognitive neuroscience perspective. DOI: 10.1016/j.tics.2006.12.005