

Spin Glass models of Syntactic Parameters

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this lecture based on:

- Karthik Siva, Jim Tao, Matilde Marcolli, *Syntactic Parameters and Spin Glass Models of Language Change*, Linguistic Analysis, Vol. 41 (2017) N. 3-4, 559–608.

Spin Glass viewpoint

- historical examples syntactic parameters can flip by effect of interaction between languages: Sanskrit is believed to have flipped some syntactic parameters by influence of Dravidian languages...
- physicist viewpoint: binary variables (up/down spins) that flip by effect of interactions: **Spin Glass Model**
 - focus on linguistic change caused by language interactions
 - think of syntactic parameters as spin variables
 - spin interaction tends to align (ferromagnet)
 - strength of interaction proportional to bilingualism (MediaLab)
 - role of temperature parameter: probabilistic interpretation of parameters & amount of code-switching in bilingual populations
 - not all parameters are independent: entailment relations
 - Metropolis–Hastings algorithm: simulate evolution

The Ising Model of spin systems on a graph G

- **graph**: vertices = languages, edges = language interaction (strength proportional to bilingual population); over each vertex a set of spin variables (syntactic parameters)
- configurations of spins $s : V(G) \rightarrow \{\pm 1\}$
- magnetic field B and correlation strength J : Hamiltonian

$$H(s) = -J \sum_{e \in E(G): \partial(e) = \{v, v'\}} s_v s_{v'} - B \sum_{v \in V(G)} s_v$$

- first term measures degree of alignment of nearby spins
- second term measures alignment of spins with direction of magnetic field

Equilibrium Probability Distribution

- Partition Function $Z_G(\beta)$

$$Z_G(\beta) = \sum_{s: V(G) \rightarrow \{\pm 1\}} \exp(-\beta H(s))$$

- Probability distribution on the configuration space: **Gibbs measure**

$$\mathbb{P}_{G,\beta}(s) = \frac{e^{-\beta H(s)}}{Z_G(\beta)}$$

- low energy states weight most
- at low temperature (large β): ground state dominates; at higher temperature (β small) higher energy states also contribute

Average Spin Magnetization

$$M_G(\beta) = \frac{1}{\#V(G)} \sum_{s: V(G) \rightarrow \{\pm 1\}} \sum_{v \in V(G)} s_v \mathbb{P}(s)$$

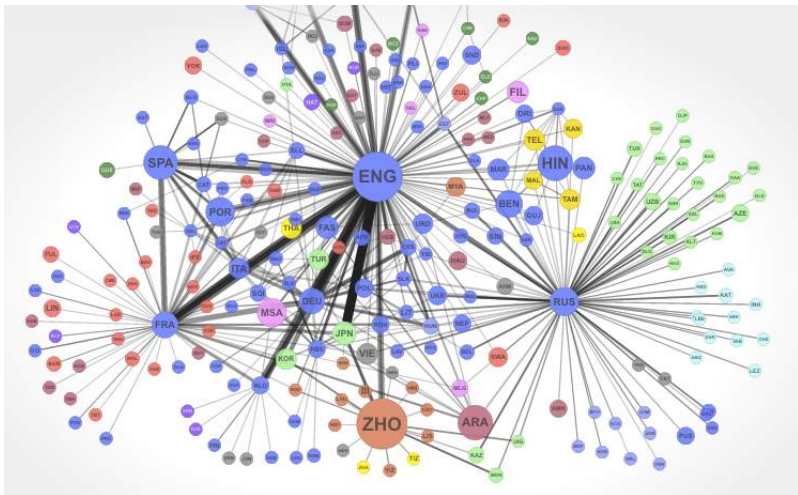
- Free energy $F_G(\beta, B) = \log Z_G(\beta, B)$

$$M_G(\beta) = \frac{1}{\#V(G)} \frac{1}{\beta} \left(\frac{\partial F_G(\beta, B)}{\partial B} \right) \Big|_{B=0}$$

- **if all syntactic parameters were independent:** just have several uncoupled Ising models (low temperature: converge to more prevalent up/down state in initial configuration; high temperature fluctuations around zero magnetization state)

Syntactic Parameters and Ising/Potts Models

- characterize set of $n = 2^N$ languages \mathcal{L}_i by binary strings of N syntactic parameters (Ising model)
- or by ternary strings (Potts model) if take values ± 1 for parameters that are set and 0 for parameters that are not defined in a certain language
- a system of n interacting languages = graph G with $n = \#V(G)$
- languages \mathcal{L}_i = vertices of the graph (e.g. language that occupies a certain geographic area)
- languages that have interaction with each other = edges $E(G)$ (geographical proximity, or high volume of exchange for other reasons)



graph of language interaction (detail) from Global Language Network of MIT MediaLab, with interaction strengths J_e on edges based on number of book translations (or Wikipedia edits)

- if only one syntactic parameter, would have an Ising model on the graph G : configurations $s : V(G) \rightarrow \{\pm 1\}$ set the parameter at all the locations on the graph
- variable interaction energies along edges (some pairs of languages interact more than others)
- magnetic field B and correlation strength J : Hamiltonian

$$H(s) = - \sum_{e \in E(G): \partial(e) = \{v, v'\}} \sum_{i=1}^N J_e s_{v,i} s_{v',i}$$

- if N parameters, configurations

$$\underline{s} = (s_1, \dots, s_N) : V(G) \rightarrow \{\pm 1\}^N$$

- if all N parameters are independent, then it would be like having N non-interacting copies of a Ising model on the same graph G (or N independent choices of an initial state in an Ising model on G)

Metropolis–Hastings

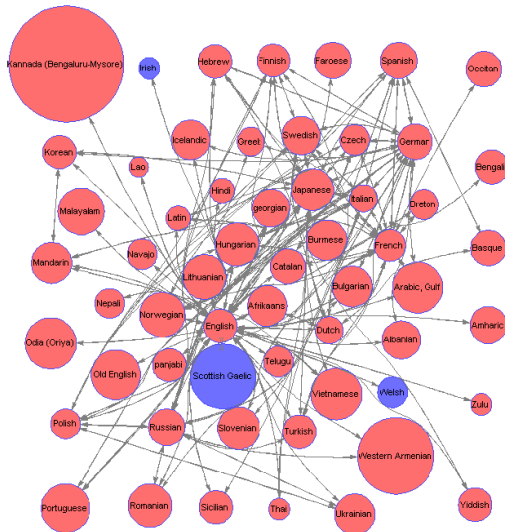
- detailed balance condition $\mathbb{P}(s)\mathbb{P}(s \rightarrow s') = \mathbb{P}(s')\mathbb{P}(s' \rightarrow s)$ for probabilities of transitioning between states (Markov process)
- transition probabilities $\mathbb{P}(s \rightarrow s') = \pi_A(s \rightarrow s') \cdot \pi(s \rightarrow s')$ with $\pi(s \rightarrow s')$ conditional probability of proposing state s' given state s and $\pi_A(s \rightarrow s')$ conditional probability of accepting it
- Metropolis–Hastings choice of acceptance distribution (Gibbs)

$$\pi_A(s \rightarrow s') = \begin{cases} 1 & \text{if } H(s') - H(s) \leq 0 \\ \exp(-\beta(H(s') - H(s))) & \text{if } H(s') - H(s) > 0. \end{cases}$$

satisfying detailed balance

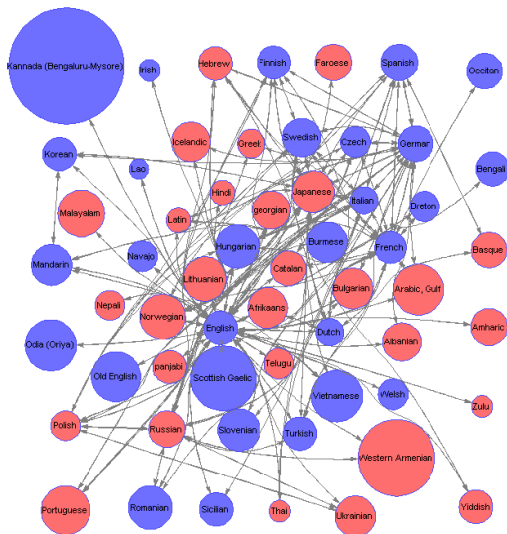
- selection probabilities $\pi(s \rightarrow s')$ single-spin-flip dynamics
- ergodicity of Markov process \Rightarrow unique stationary distribution

Example: Single parameter dynamics *Subject-Verb* parameter



Initial configuration: most languages in SSWL have +1 for *Subject-Verb*; use interaction energies from MediaLab data

Equilibrium: low temperature all aligned to +1; high temperature:



Temperature: fluctuations in bilingual users between different structures (“code-switching” in Linguistics)

Entailment relations among parameters

- Example: $\{p_1, p_2\} = \{\text{Strong Deixis, Strong Anaphoricity}\}$

	p_1	p_2
l_1	+1	+1
l_2	-1	0
l_3	+1	+1
l_4	+1	-1

$\{l_1, l_2, l_3, l_4\} = \{\text{English, Welsh, Russian, Bulgarian}\}$

Strong Deixis +1: governs possible positions of demonstratives in the nominal domain

Strong Anaphoricity +1: obligatory dependence on an antecedent in a local and asymmetric relation to anaphor

Modeling Entailment

- variables: $S_{\ell,p_1} = \exp(\pi i X_{\ell,p_1}) \in \{\pm 1\}$, $S_{\ell,p_2} \in \{\pm 1, 0\}$ and $Y_{\ell,p_2} = |S_{\ell,p_2}| \in \{0, 1\}$
- Hamiltonian $H = H_E + H_V$

$$H_E = H_{p_1} + H_{p_2} = - \sum_{\ell, \ell' \in \text{languages}} J_{\ell \ell'} \left(\delta_{S_{\ell,p_1}, S_{\ell',p_1}} + \delta_{S_{\ell,p_2}, S_{\ell',p_2}} \right)$$

$$H_V = \sum_{\ell} H_{V,\ell} = \sum_{\ell} J_{\ell} \delta_{X_{\ell,p_1}, Y_{\ell,p_2}}$$

$J_{\ell} > 0$ anti-ferromagnetic

- two parameters: *temperature* as before and coupling *energy of entailment*
- if freeze p_1 and evolution for p_2 : Potts model with external magnetic field

Acceptance probabilities

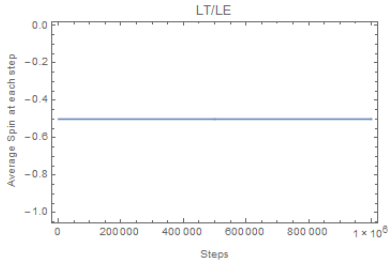
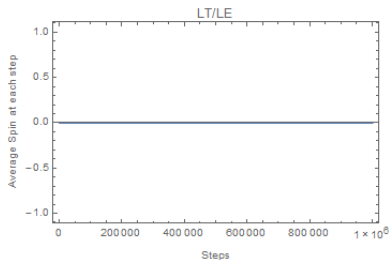
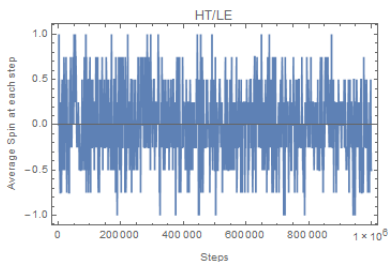
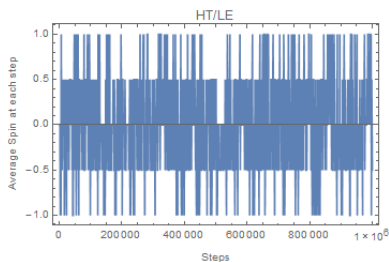
$$\pi_A(s \rightarrow s \pm 1 \pmod{3}) = \begin{cases} 1 & \text{if } \Delta_H \leq 0 \\ \exp(-\beta \Delta_H) & \text{if } \Delta_H > 0. \end{cases}$$

$$\Delta_H := \min\{H(s + 1 \pmod{3}), H(s - 1 \pmod{3})\} - H(s)$$

Equilibrium configuration

(p_1, p_2)	HT/HE	HT/LE	LT/HE	LT/LE
l_1	(+1, 0)	(+1, -1)	(+1, +1)	(+1, -1)
l_2	(+1, -1)	(-1, -1)	(+1, +1)	(+1, -1)
l_3	(-1, 0)	(-1, +1)	(+1, +1)	(-1, 0)
l_4	(+1, +1)	(-1, -1)	(+1, +1)	(-1, 0)

Average value of spin



p_1 left and p_2 right in low entailment energy case

- when consider more realistic models (at least the 28 languages and 63 parameters of Longobardi–Guardiano with all their entailment relations) **very slow convergence** of the Metropolis–Hastings dynamics even for low temperature
- how to get better information on the dynamics? consider set of languages as codes and an induced dynamics in the space of code parameters
- to be discussed later: a coding theory perspective on code parameters; induced dynamics on the space of codes shows more easily long term behavior of the system

How to improve this dynamical model?

- language change is related to mechanisms of language acquisition
- dynamical systems models of language acquisition were proposed by Berwick and Niyogi based on a Markov model on a space of possible grammars (in the formal languages sense)
- would like to couple the spin glass dynamics capturing language interaction through code-switching and bilingualism to a dynamical model of language acquisition

Next to be discussed: how to detect relations between syntactic parameters? what is the manifold of syntax?