

(work in progress)

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- Brief intro to Minority Games
- Lookup-table versus inner product MGs
- Generalization of GFA for MGs with real histories
- Analysis of macroscopic laws
 - fake history limit
 - true history: random matrix theory

- At each round ℓ :
 all get info $I(\ell) \in \{I_1, \dots, I_p\}$
 each places a bid $b_i(\ell) \in \mathbb{R}$
- Those in the **minority** group win
 $\sum_j b_j(\ell) > 0 : \quad b_i(\ell) < 0$ wins
 $\sum_j b_j(\ell) < 0 : \quad b_i(\ell) > 0$ wins
- Each agent i has S strategies
 converting $I(\ell)$ into bid $b_i(\ell)$
- dynamics:
 choice of strategy $a \in \{1, \dots, S\}$,
 as a function of time, for all agents,
 based on strategies' performance



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Lookup table MGS (Challet, Marsili, Zhang, ...):

info : m step history, $\lambda(\ell) = (\text{sgn}[A(\ell-1)], \dots, s$
strat : lookup table, $\mathbf{R}_{ia} = (R_{ia}^1, \dots, R_{ia}^p)$
bid : strategy a , info λ , $b_i(\ell) = R_{ia}^\lambda$

history window: $\Delta t = \Delta \ell / N = \mathcal{O}(N^{-1} \log N)$

Inner product MGS (Cavagna, Sherrington, Garrahan

info : p step history, $\lambda(\ell) = (f[A(\ell-1)], \dots, j$
strat : linear functional, $\mathbf{R}_{ia} = (R_{ia}^1, \dots, R_{ia}^p)$
bid : strategy a , info λ , $b_i(\ell) = p^{-\frac{1}{2}} \sum_{\mu \leq p} R_{ia}^\mu \lambda_\mu(\ell)$

history window: $\Delta t = \Delta \ell / N = \mathcal{O}(1)$

- Inner product MGs closer to how real agents predict time series (generalized linear models, ARMA, ARIMA, etc)
- Lookup-table MGs analyzed extensively using replica & GFA techniques (including real history version)
- Not true for inner product MGs
- Lookup table: history window $\mathcal{O}(N^{-1} \log N) \rightarrow 0$
Inner product: history window $\mathcal{O}(N^0)$

history strings:

$$\lambda(\ell) = (\text{sgn}[A(\ell - 1)], \dots, \text{sgn}[A(\ell - m)]) \quad 2^m$$

$$\pi_\lambda = \lim_{L \rightarrow \infty} \frac{1}{L} \sum_{\ell \leq L} \delta_{\lambda, \lambda(\ell)} \quad \varrho(f) = \lim_{N \rightarrow \infty} 2^{-m} \sum_{\lambda} \delta_{\lambda, \lambda(f)}$$

$$\alpha < \alpha_c$$

$$\alpha > \alpha_c$$

$$\varrho(f)$$

$$\varrho(f)$$

f

f

- Inner product MGs with real market history
no such thing as $\varrho(f)$, history string is of length $\mathcal{O}(L)$

(always $S = 2$)

- decision noise:

$$\text{sgn}[q_i(\ell)] \rightarrow \sigma[q_i(\ell), z_i(\ell)] \quad \begin{array}{l} \text{additive : } \sigma[q, z] = \text{sgn} \\ \text{multipl : } \sigma[q, z] = \text{sgn} \end{array}$$

- real vs fake history:

$$A(\ell - \mu) \rightarrow (1 - \zeta)A(\ell - \mu) + \zeta Z(\ell, \mu)$$

$$Z(\ell, \mu) : \text{ Gaussian, } \langle Z \rangle = 0$$

$$\text{consistent : } Z(\ell, \mu) = Z(\ell - \mu), \quad \langle Z(\ell)Z(\ell') \rangle = \xi$$

$$\text{inconsistent : } Z(\ell, \mu) \text{ all indep, } \langle Z(\ell, \mu)Z(\ell', \mu') \rangle = 0$$

$$\begin{aligned} \text{strategy :} & \quad \mathbf{R}_{ia} = (R_{ia}^1, \dots, R_{ia}^p) \quad 2^m = p \\ \text{bid :} & \quad b_i(\ell) = R_{ia}^\lambda \end{aligned}$$

Inner product MGs:

$$\begin{aligned} \text{info :} & \quad \lambda(\ell) = (f[A(\ell-1)], \dots, f[A(\ell-1)]) \\ \text{strategy :} & \quad \mathbf{R}_{ia} = (R_{ia}^1, \dots, R_{ia}^p) \\ \text{bid :} & \quad b_i(\ell) = p^{-\frac{1}{2}} \sum_{\mu \leq p} R_{ia}^\mu \lambda_\mu(\ell) \end{aligned}$$

- add bid perturbation term:

$$A(t) = \frac{1}{\sqrt{N}} \sum_i b_i(t) + A_e(t)$$

[]: over $\{R_{ia}^{\nu}\}$

$$\overline{Z[\psi]} = \overline{\langle e^{-i \sum_{il} \psi_i(l) \sigma[q_i(l), z_i(l)]} \rangle}$$

manipulations:

$$\overline{Z[\psi]} = \int \mathcal{D}C \mathcal{D}\hat{C} \mathcal{D}G \mathcal{D}\hat{G} \dots e^{N\Psi[C, \hat{C}, G, \hat{G}, \dots]}$$

saddle-point equations

for dynamic order parameters

$$C(\ell, \ell') = \lim_{N \rightarrow \infty} \frac{1}{N} \sum_i \overline{\langle \sigma[q_i(\ell), z_i(\ell)] \sigma[q_i(\ell'), z_i(\ell')] \rangle}$$

$$G(\ell, \ell') = \lim_{N \rightarrow \infty} \frac{1}{N} \sum_i \frac{\partial}{\partial \theta_i(\ell')} \overline{\langle \sigma[q_i(\ell), z_i(\ell)] \rangle}$$

- Effective single agent process.

$$\frac{d}{dt}q(t) = \theta(t) - \alpha \int_0^t dt' R(t, t') \sigma[q(t')] + \sqrt{\alpha} \eta(t)$$

Gaussian $\eta(t)$, $\langle \eta(t) \rangle = 0$, $\langle \eta(t)\eta(t') \rangle$

$$\sigma[q] = \int Dz \sigma[q, z]$$

- closed eqns for order parameters:

$$C(t, t') = \langle \sigma[q(t)]\sigma[q(t')] \rangle_{\star} \quad G(t, t') = \frac{\delta}{\delta\theta(t')} \langle \sigma$$

- All complications: in relation between $\{R, \Sigma\}$ and σ involves effective overall bid process

$$\begin{aligned}
C(t, t') &= C(t - t'), & G(t, t') &= G(t - t') \\
R(t, t') &= R(t - t'), & \Sigma(t, t') &= \Sigma(t - t')
\end{aligned}$$

static order pars:

$$\chi = \int_0^\infty dt G(t), \quad \chi_R = \int_0^\infty dt R(t)$$

$$c = \lim_{t \rightarrow \infty} C(t), \quad \phi : \text{fraction of 'frozen' ag}$$

given by

$$\phi = 1 - \text{Erf}[u]$$

$$c = \sigma^2[\infty] \left\{ 1 - \text{Erf}[u] + \frac{1}{2u^2} \text{Erf}[u] - \frac{1}{u\sqrt{\pi}} e^{-u^2} \right\}$$

$$\chi = \text{Erf}[u] / \alpha \chi_R$$

$$u = \sqrt{\alpha \chi_R} \sigma[\infty] / S_0 \sqrt{2}, \quad S_0^2 = \Sigma(\infty)$$

- $S_0^2 = \Sigma(\infty)$ and $\chi_R = \int_0^\infty dt R(t)$

$$\chi_R = \lim_{\delta \rightarrow 0} \left\{ \overline{W}[0, 0; \{A, Z\}] + \sum_{\ell=1}^{\infty} \frac{\partial}{\partial A(0)} \langle \langle \overline{W}[\ell, 0; \{A, Z\}] \rangle \rangle \right\}$$

$$S_0^2 = \lim_{\delta \rightarrow 0} \lim_{L \rightarrow \infty} \frac{\tilde{\eta}}{L^2 \delta} \sum_{\ell, \ell'=1}^L \langle \langle \overline{W}[\ell, \ell'; \{A, Z\}] A(\ell) A(\ell') \rangle \rangle$$

- Differences between models: $\overline{W}[\ell, \ell'; \{A, Z\}]$
similarity between histories observed at times ℓ and

lookup table : $\overline{W}[\ell, \ell'; \{A, Z\}] = \delta_{\lambda(\ell, A, Z), \lambda(\ell', A, Z)}$

inner product : $\overline{W}[\ell, \ell'; \{A, Z\}] =$

$$\frac{1}{\alpha N} \sum_{\lambda \leq p} f[(1-\zeta)A(\ell-\lambda) + \zeta Z(\ell, \lambda)] f[(1-\zeta)A(\ell'-\lambda) + \zeta Z(\ell', \lambda)]$$

Gaussian random fields:

$$\langle \phi_\ell \rangle_{\{\phi|A,Z\}} = 0, \quad \langle \phi_\ell \phi_{\ell'} \rangle_{\{\phi|A,Z\}} = \frac{1}{2} [1 + C(\ell, \ell')] \bar{W}$$

Inner product MG with inconsistent take market I

- Put $\zeta \rightarrow 1$,
calculate kernels (grammatically),
define $\kappa = \int Dz f^2[Sz]$
- Inner product and lookup table MGs
differ only in characteristic amplitude,
that can be transformed away:

$$G_{\text{IP}}(t, t') = \kappa^{-1} G_{\text{LU}}(t, t'), \quad \theta_{\text{IP}}(t) = \kappa \theta_{\text{LU}}(t), \quad \sigma_{\text{IP}}^2$$

- define random matrix $\mathbf{B}(\ell)$:

$$B_{\lambda\lambda'}(\ell) = \mathcal{F}_\lambda[\ell, A, Z] \mathcal{F}_{\lambda'}[\ell, A, Z]$$

$$\text{LU : } \quad \mathcal{F}_\lambda[\ell, A, Z] = \sqrt{\alpha N} \delta_{\lambda, \lambda(\ell, A, Z)} \quad \lambda$$

$$\text{IP : } \quad \mathcal{F}_\lambda[\ell, A, Z] = f[(1 - \zeta)A(\ell - \lambda) + \zeta Z(\ell, \lambda)] \quad \lambda$$

- $\{R, \Sigma\}$ can be written in terms of

$$\Delta_{r+1}(\ell_0, \dots, \ell_r) = \frac{1}{p} \text{Tr} \left\langle \left\langle \mathbf{B}(\ell_0) \mathbf{B}(\ell_1) \dots \mathbf{B}(\ell_r) \right\rangle \right\rangle_{\{A\}}$$

$$\tilde{\Delta}_{r+r'+2}(\ell_0, \dots, \ell_r; \ell'_0, \dots, \ell'_{r'}) =$$

$$\frac{1}{p} \text{Tr} \left\langle \left\langle \mathbf{B}(\ell_0) \mathbf{B}(\ell_1) \dots \mathbf{B}(\ell_r) \mathbf{B}(\ell'_{r'}) \dots \mathbf{B}(\ell'_1) \mathbf{B}(\ell'_0) \right\rangle \right\rangle_{\{A\}}$$

$$\tilde{\tilde{\Delta}}_{r+r'+1}(\ell_0, \ell_1, \dots, \ell_r; \ell'_0, \ell'_1, \dots, \ell'_{r'}) =$$

$$\frac{1}{p} \text{Tr} \left\langle \left\langle \mathbf{B}(\ell_0) \mathbf{B}(\ell_1) \dots \mathbf{B}(\ell_r) \mathbf{B}(\ell'_{r'}) \dots \mathbf{B}(\ell'_1) \right\rangle \right\rangle_{\{A\}}$$

$$\prod_{i=1}^r \mathbf{B}(l_i) \rightarrow \mathbf{B}^r(A, Z) \quad \mathbf{B}(A, Z) = \lim_{L \rightarrow \infty} \frac{1}{L} \sum_{\ell \leq L}$$

$$\mu_i(A, Z) \text{ ev of } \mathbf{B}(A, Z) : \quad \varrho(\mu) = \left\langle \left\langle \frac{1}{p} \sum_{i=1}^p \delta[\mu - \mu_i(A, Z) \right. \right.$$

$$\Delta_{r+1}(l_0, \dots, l_r) \rightarrow \int_0^\infty d\mu \varrho(\mu) \mu^{r+}$$

$$\tilde{\Delta}_{r+r'+2}(l_0, \dots, l_r; l'_0, \dots, l'_{r'}) \rightarrow \int_0^\infty d\mu \varrho(\mu) \mu^{r+}$$

$$\tilde{\tilde{\Delta}}_{r+r'+1}(l_0, l_1, \dots, l_r; l'_0, l'_1, \dots, l'_{r'}) \rightarrow \int_0^\infty d\mu \varrho(\mu) \mu^{r+}$$

$$\chi_R = \int_0^\infty d\mu \varrho(\mu) \frac{\mu}{1 + \mu\chi} \quad S_0^2 = (1 + c) \int_0^\infty d\mu \varrho(\mu)$$

eigenvalue distribution $\varrho(\mu)$ of random matrix $\mathbf{B}(A)$ takes over role of history frequency distribution $\varrho(f)$ in lookup table MGs

- link with lookup table MG theory:

$$B_{\lambda\lambda'}(A, Z) = p\delta_{\lambda\lambda'}\pi_{\lambda}(A, Z)$$

$$\varrho(\mu) = \lim_{p \rightarrow \infty} \frac{1}{p} \sum_{\lambda} \langle\langle \delta [\mu - p\pi_{\lambda}(A, Z)] \rangle\rangle_{\{A, Z\}}$$

(recover the history frequency distr)

- Generation functional analysis of MGs can be generalised to include lookup table & inner product as special case
- Exact macroscopic theory in terms of effective singular value decomposition and (in case of real histories) an effective overall bid process
- Fake histories:
 - inner product and lookup table MGs differ only in simple rescaling of observables, phase diagrams identical
- Real histories:
 - Short history correlation time ansatz can be generalised
 - Role of history frequency distribution in lookup table taken over by eigenvalue spectrum of a random matrix (randomness: generated by the effective bid process)
- To be finished:
 - calculation of random matrix eigenvalue spectrum