

Observation, Communication and Intelligence in Agent-Based Systems

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Presentation Outline

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2 Methodology

3 Experiments

4 Results

5 Conclusions and future work

Motivation

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References

Problem description and goals

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How do two factors affect *intelligence* in cooperative agent-based systems?

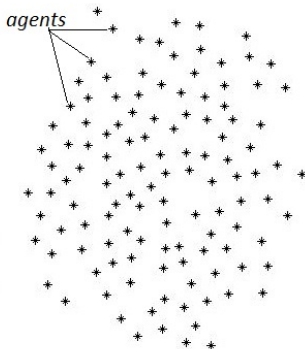
- Communication
- Observation

How do they depend on each other?

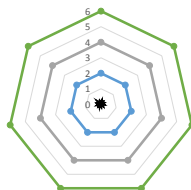
- and how does this reflect on the collective performance of the agents?

Problem description

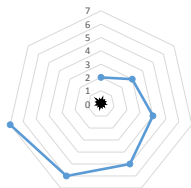
Which factor has the greater influence?



Communication range



Observation range



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- to the best of our knowledge, no studies have applied formal intelligence tests for this purpose.
- in the real-world agents have limited sensitivity of the environment (observations)
- relying on communication to improve their performance can be inevitable.
- benefits of quantifying the influence
 - understanding the rules of information aggregation in a multiagent setting
 - predicting the usefulness and expected performance of these systems over different configurations.

Approach

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- *Quantify* the performance of artificial agents.
 - Not every evaluation metric can be used as a formal (universal) intelligence test
- Use (machine) intelligence tests
 - evaluate a group of artificial agents collaborating in different settings
- Adjust communication and observation abilities
 - controlled experiments
 - how these changes are reflected by their measured intelligence

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What: Anytime Universal Intelligence Test (anYnt)

Why :

- test derived from formal and mathematical considerations
- builds upon Legg and Hutter's definition of universal intelligence
- used in practice to evaluate artificial agents in a dynamic setting

How :

- agent-environment framework [4].
- controlled experiments
 - 1 Observation
 - 2 Communication

Motivation

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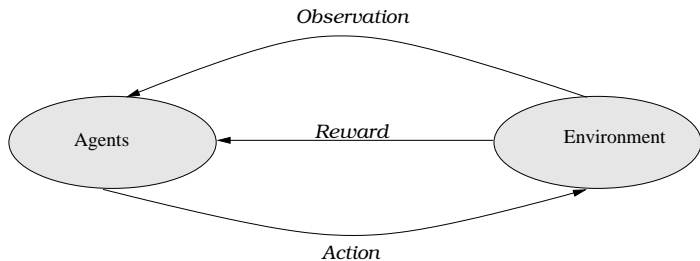
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Agent-Environment Framework



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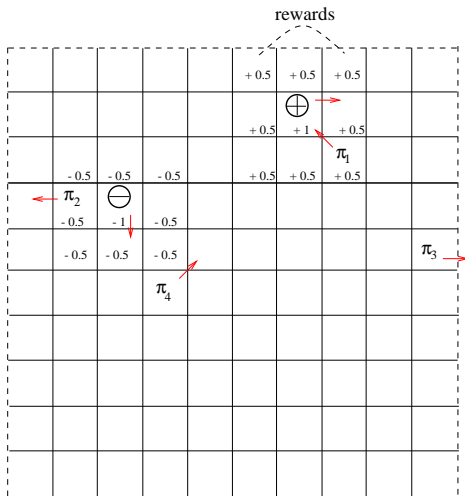
Λ (lambda) environment: Anytime Intelligence Test

An interactive setting:

- Set of agents $\Pi = \{\pi_1, \pi_2, \pi_3, \pi_4\}$
- \oplus, \ominus : positive/negative rewards
- Toroidal environment space
- Measured complexity of movement patterns
- Balanced environment
- Dynamic setting
- Universal Intelligence [4]

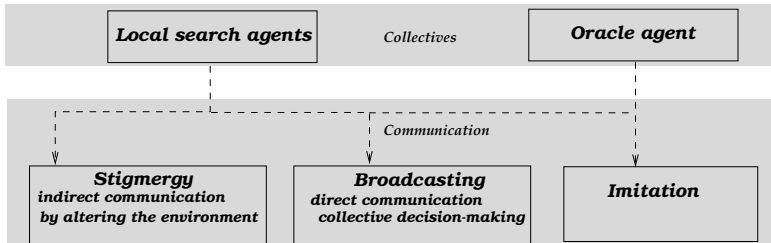
$$\Upsilon(\pi, U) := \sum_{\mu=i}^{\infty} p_U(\mu) \cdot V_{\mu}^{\pi}$$
$$= \sum_{\mu=i}^{\infty} p_U(\mu) \cdot E \left(\sum_{i=1}^{\infty} r_i^{\mu, \pi} \right)$$

s.t. μ : environment coded on a universal machine U with probability $p_U(\mu)$



Communication modes

Set of agents $\Pi = \{\pi_1, \pi_2, \pi_3, \dots, \pi_n\}$



Measuring Uncertainty and Information

Shannon's entropy [6]

- Environment μ with N possible states
- $H(\mu)$: uncertainty in a given environment μ
- $H(o)$: amount of information in an observation o
 - \log_2 the number of states in o , or the minimum number of bits needed to describe o
- $H(c)$: amount of information in a communication range c
 - \log_2 the number of states in c , or the minimum number of bits needed to describe c

At the beginning of a test the entropy is maximal.

Probability $p(s_\mu)$ of a given state s_μ occurring follows a uniform distribution and is equal to $1/|N|$.

$$H(\mu) = - \sum_{s_\mu \in N} p(s_\mu) \log_2 p(s_\mu) = \log_2 |N| \text{ bits.}$$

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- 1 initialize Λ environment
 - set environment space dimensions
 - generate set of agents to be evaluated Π
 - initialize $H(o)$ and $H(c)$ parameters
- 2 evaluate Π over 1000 episodes
- 3 record **average score** $\Upsilon(H_o, H_c) \in [-1.0, 1.0]$ (used as intelligence/performance measure)
- 4 change entropy of $H(o)$ or $H(c)$
- 5 repeat experiment (from step 2) with new $H(o)$ or $H(c)$

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Results: stigmergy or indirect communication

Observation entropies $H(o)$ \longrightarrow

Communication entropies $H(c)$ \longleftarrow

	0.04	0.11	0.22	0.36	0.54	0.76	1.01	1.30	1.62	1.99	2.38	2.82	3.29	3.79	4.33	4.91	5.53	6.18	6.86	7.58	8.34	9.14	9.97	10.84
10.84	0.0506	0.0693	0.0927	0.1120	0.1329	0.1517	0.1735	0.1998	0.2256	0.2545	0.2753	0.2903	0.3059	0.3277	0.3431	0.3615	0.3833	0.4076	0.4234	0.4328	0.4439	0.4470	0.4472	0.4470
9.97	0.0469	0.0646	0.0837	0.1078	0.1290	0.1502	0.1712	0.1978	0.2256	0.2667	0.2990	0.3179	0.3297	0.3422	0.3609	0.3778	0.3933	0.4140	0.4291	0.4385	0.4448	0.4475	0.4468	0.4473
9.14	0.0450	0.0657	0.0883	0.1008	0.1203	0.1434	0.1738	0.1950	0.2486	0.2907	0.3245	0.3419	0.3491	0.3652	0.3823	0.3930	0.4073	0.4213	0.4335	0.4398	0.4470	0.4472	0.4469	0.4465
8.34	0.0441	0.0618	0.0824	0.1022	0.1200	0.1409	0.1766	0.2188	0.2637	0.3064	0.3394	0.3588	0.3695	0.3779	0.3955	0.4059	0.4180	0.4273	0.4339	0.4458	0.4470	0.4459	0.4464	0.4468
7.58	0.0461	0.0702	0.0849	0.1046	0.1222	0.1497	0.1841	0.2333	0.2793	0.3249	0.3541	0.3707	0.3845	0.3945	0.4046	0.4131	0.4231	0.4311	0.4419	0.4450	0.4449	0.4466	0.4461	0.4471
6.86	0.0489	0.0677	0.0863	0.1056	0.1274	0.1584	0.2024	0.2542	0.3105	0.3445	0.3696	0.3828	0.3943	0.4005	0.4111	0.4202	0.4268	0.4389	0.4410	0.4448	0.4458	0.4485	0.4463	0.4467
6.18	0.0486	0.0671	0.0875	0.1076	0.1313	0.1676	0.2249	0.2817	0.3274	0.3575	0.3752	0.3897	0.4002	0.4079	0.4153	0.4229	0.4325	0.4383	0.4388	0.4409	0.4432	0.4438	0.4451	0.4468
5.53	0.0478	0.0720	0.0927	0.1139	0.1470	0.1853	0.2503	0.3062	0.3457	0.3669	0.3827	0.3969	0.4053	0.4119	0.4172	0.4284	0.4327	0.4318	0.4360	0.4362	0.4386	0.4404	0.4438	0.4467
4.91	0.0506	0.0721	0.0959	0.1199	0.1556	0.2088	0.2733	0.3255	0.3614	0.3770	0.3872	0.4016	0.4080	0.4141	0.4236	0.4295	0.4293	0.4306	0.4336	0.4353	0.4370	0.4408	0.4442	0.4468
4.33	0.0506	0.0701	0.0965	0.1268	0.1726	0.2388	0.3002	0.3446	0.3698	0.3875	0.3941	0.4053	0.4136	0.4202	0.4246	0.4255	0.4271	0.4294	0.4299	0.4300	0.4330	0.4395	0.4436	0.4471
3.79	0.0498	0.0745	0.1018	0.1413	0.1971	0.2685	0.3224	0.3579	0.3803	0.3893	0.3989	0.4073	0.4157	0.4187	0.4167	0.4197	0.4231	0.4227	0.4220	0.4249	0.4294	0.4373	0.4432	0.4469
3.29	0.0528	0.0741	0.1050	0.1534	0.2286	0.2955	0.3363	0.3670	0.3859	0.3960	0.4037	0.4103	0.4165	0.4138	0.4168	0.4179	0.4162	0.4171	0.4158	0.4182	0.4256	0.4344	0.4425	0.4470
2.82	0.0486	0.0790	0.1126	0.1686	0.2504	0.3151	0.3456	0.3732	0.3870	0.3982	0.4038	0.4132	0.4161	0.4109	0.4119	0.4135	0.4108	0.4071	0.4122	0.4128	0.4186	0.4308	0.4414	0.4464
2.38	0.0533	0.0809	0.1203	0.1881	0.2643	0.3216	0.3506	0.3770	0.3906	0.4011	0.4066	0.4046	0.4056	0.4054	0.4018	0.3999	0.4014	0.3987	0.4034	0.4102	0.4165	0.4279	0.4403	0.4463
1.99	0.0530	0.0836	0.1338	0.2069	0.2717	0.3202	0.3493	0.3731	0.3897	0.3976	0.3947	0.4017	0.4001	0.3958	0.3948	0.3917	0.3913	0.3956	0.3948	0.4023	0.4113	0.4285	0.4404	0.4465
1.62	0.0533	0.0898	0.1457	0.2127	0.2681	0.3125	0.3444	0.3705	0.3893	0.3902	0.3907	0.3919	0.3939	0.3867	0.3846	0.3856	0.3862	0.3843	0.3869	0.3990	0.4083	0.4269	0.4404	0.4461
1.30	0.0570	0.0929	0.1562	0.2140	0.2566	0.2991	0.3378	0.3690	0.3729	0.3830	0.3816	0.3879	0.3829	0.3784	0.3774	0.3800	0.3791	0.3811	0.3873	0.3899	0.4062	0.4238	0.4391	0.4469
1.01	0.0565	0.1033	0.1582	0.2100	0.2538	0.2894	0.3195	0.3447	0.3522	0.3741	0.3772	0.3811	0.3755	0.3741	0.3739	0.3752	0.3779	0.3805	0.3830	0.3926	0.4056	0.4247	0.4397	0.4468
0.76	0.0604	0.1051	0.1568	0.1969	0.2339	0.2635	0.2985	0.3269	0.3488	0.3615	0.3665	0.3693	0.3705	0.3718	0.3727	0.3745	0.3757	0.3811	0.3799	0.3886	0.4017	0.4243	0.4403	0.4467
0.54	0.0645	0.1002	0.1448	0.1849	0.2159	0.2476	0.2758	0.3061	0.3352	0.3500	0.3577	0.3635	0.3705	0.3736	0.3757	0.3771	0.3747	0.3794	0.3813	0.3890	0.4008	0.4220	0.4392	0.4468
0.36	0.0660	0.0956	0.1355	0.1637	0.1899	0.2197	0.2580	0.2882	0.3188	0.3381	0.3560	0.3650	0.3705	0.3741	0.3781	0.3798	0.3828	0.3873	0.3853	0.3899	0.4022	0.4216	0.4405	0.4467
0.22	0.0605	0.0917	0.1150	0.1514	0.1777	0.2061	0.2378	0.2714	0.3023	0.3317	0.3498	0.3643	0.3768	0.3797	0.3817	0.3877	0.3910	0.3916	0.3941	0.3958	0.4058	0.4231	0.4420	0.4472
0.11	0.0548	0.0779	0.0997	0.1313	0.1592	0.1893	0.2186	0.2520	0.2864	0.3165	0.3372	0.3630	0.3728	0.3845	0.3911	0.3952	0.4016	0.4032	0.4024	0.4090	0.4137	0.4274	0.4422	0.4469
0.04	0.0486	0.0676	0.0919	0.1146	0.1400	0.1715	0.2025	0.2317	0.2671	0.2987	0.3307	0.3542	0.3734	0.3890	0.3970	0.4031	0.4091	0.4116	0.4141	0.4194	0.4265	0.4305	0.4422	0.4464

Indirect communication

Results: direct communication

Observation entropies $H(o) \rightarrow$

Communication entropies $H(c) \uparrow$

	0.04	0.11	0.22	0.36	0.54	0.76	1.01	1.30	1.62	1.99	2.38	2.82	3.29	3.79	4.33	4.91	5.53	6.18	6.86	7.58	8.34	9.14	9.97	10.84	
10.84	0.3487	0.3843	0.4119	0.4271	0.4352	0.4424	0.4443	0.4454	0.4469	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470
9.97	0.3487	0.3843	0.4119	0.4271	0.4352	0.4424	0.4443	0.4454	0.4469	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470
9.14	0.3487	0.3843	0.4119	0.4271	0.4352	0.4424	0.4443	0.4454	0.4469	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470
8.34	0.3487	0.3843	0.4119	0.4271	0.4352	0.4424	0.4443	0.4454	0.4469	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470
7.58	0.3487	0.3843	0.4119	0.4271	0.4352	0.4424	0.4443	0.4454	0.4469	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470
6.86	0.3487	0.3843	0.4119	0.4271	0.4352	0.4424	0.4443	0.4454	0.4469	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470	0.4470
6.18	0.3449	0.3799	0.4057	0.4208	0.4282	0.4364	0.4385	0.4400	0.4418	0.4416	0.4420	0.4429	0.4432	0.4440	0.4444	0.4447	0.4455	0.4462	0.4469	0.4469	0.4470	0.4469	0.4469	0.4469	0.4470
5.53	0.3377	0.3698	0.3947	0.4120	0.4205	0.4290	0.4316	0.4326	0.4353	0.4356	0.4359	0.4378	0.4391	0.4403	0.4415	0.4422	0.4435	0.4452	0.4463	0.4465	0.4469	0.4470	0.4469	0.4469	0.4469
4.91	0.3294	0.3588	0.3888	0.4028	0.4111	0.4191	0.4217	0.4241	0.4253	0.4277	0.4308	0.4321	0.4333	0.4367	0.4377	0.4398	0.4417	0.4437	0.4455	0.4464	0.4472	0.4470	0.4469	0.4468	0.4468
4.33	0.3153	0.3557	0.3767	0.3901	0.4016	0.4079	0.4115	0.4156	0.4167	0.4193	0.4222	0.4269	0.4280	0.4311	0.4352	0.4365	0.4401	0.4418	0.4449	0.4461	0.4470	0.4469	0.4468	0.4468	0.4468
3.79	0.3069	0.3415	0.3633	0.3801	0.3890	0.3976	0.4016	0.4046	0.4080	0.4125	0.4153	0.4183	0.4219	0.4264	0.4295	0.4350	0.4380	0.4407	0.4436	0.4460	0.4471	0.4471	0.4469	0.4471	0.4471
3.29	0.3002	0.3349	0.3544	0.3709	0.3760	0.3844	0.3912	0.3942	0.3985	0.4023	0.4068	0.4100	0.4179	0.4216	0.4276	0.4326	0.4360	0.4407	0.4431	0.4458	0.4467	0.4470	0.4469	0.4471	0.4471
2.82	0.2926	0.3124	0.3395	0.3561	0.3614	0.3732	0.3780	0.3822	0.3873	0.3944	0.3979	0.4047	0.4103	0.4163	0.4229	0.4306	0.4342	0.4390	0.4430	0.4458	0.4470	0.4470	0.4467	0.4469	0.4469
2.38	0.2713	0.3035	0.3237	0.3433	0.3503	0.3590	0.3651	0.3716	0.3741	0.3829	0.3906	0.3965	0.4030	0.4113	0.4174	0.4259	0.4315	0.4383	0.4422	0.4454	0.4466	0.4475	0.4468	0.4468	0.4468
1.99	0.2624	0.2916	0.3074	0.3278	0.3395	0.3483	0.3535	0.3588	0.3656	0.3713	0.3827	0.3898	0.3989	0.4032	0.4136	0.4241	0.4309	0.4375	0.4420	0.4454	0.4469	0.4472	0.4470	0.4468	0.4468
1.62	0.2444	0.2764	0.2979	0.3130	0.3233	0.3341	0.3413	0.3507	0.3547	0.3640	0.3756	0.3830	0.3918	0.4025	0.4108	0.4212	0.4288	0.4358	0.4413	0.4444	0.4468	0.4471	0.4467	0.4468	0.4468
1.30	0.2303	0.2669	0.2875	0.3060	0.3114	0.3237	0.3316	0.3365	0.3466	0.3548	0.3652	0.3765	0.3830	0.3951	0.4072	0.4140	0.4263	0.4357	0.4410	0.4451	0.4474	0.4472	0.4468	0.4468	0.4468
1.01	0.2201	0.2494	0.2714	0.2913	0.2981	0.3104	0.3202	0.3285	0.3365	0.3434	0.3561	0.3671	0.3789	0.3905	0.4033	0.4158	0.4231	0.4338	0.4397	0.4442	0.4468	0.4475	0.4466	0.4469	0.4469
0.76	0.2100	0.2408	0.2598	0.2780	0.2876	0.2969	0.3071	0.3184	0.3225	0.3362	0.3468	0.3627	0.3735	0.3849	0.3994	0.4098	0.4211	0.4311	0.4397	0.4445	0.4468	0.4475	0.4468	0.4470	0.4470
0.54	0.1959	0.2220	0.2405	0.2616	0.2734	0.2800	0.2955	0.3037	0.3124	0.3251	0.3390	0.3536	0.3653	0.3825	0.3967	0.4108	0.4211	0.4304	0.4385	0.4447	0.4469	0.4476	0.4471	0.4467	0.4467
0.36	0.1750	0.2062	0.2285	0.2386	0.2591	0.2728	0.2820	0.2909	0.3023	0.3170	0.3298	0.3484	0.3627	0.3768	0.3910	0.4050	0.4189	0.4283	0.4382	0.4432	0.4469	0.4477	0.4468	0.4469	0.4469
0.22	0.1536	0.1798	0.2008	0.2240	0.2412	0.2551	0.2689	0.2824	0.2952	0.3089	0.3214	0.3379	0.3517	0.3732	0.3878	0.4045	0.4176	0.4267	0.4372	0.4428	0.4470	0.4473	0.4469	0.4466	0.4466
0.11	0.1245	0.1577	0.1780	0.1981	0.2143	0.2311	0.2505	0.2646	0.2784	0.2950	0.3129	0.3314	0.3507	0.3665	0.3844	0.4005	0.4141	0.4290	0.4373	0.4434	0.4471	0.4476	0.4464	0.4470	0.4470
0.04	0.0879	0.1194	0.1405	0.1642	0.1803	0.2043	0.2238	0.2455	0.2636	0.2847	0.3026	0.3240	0.3427	0.3635	0.3782	0.3971	0.4133	0.4273	0.4367	0.4435	0.4470	0.4478	0.4469	0.4463	0.4463

Direct communication

Results: imitation

Observation entropies $H(o)$ \longrightarrow

Communication entropies $H(c)$ \longleftarrow

	0.04	0.11	0.22	0.36	0.54	0.76	1.01	1.30	1.62	1.99	2.38	2.82	3.29	3.79	4.33	4.91	5.53	6.18	6.86	7.58	8.34	9.14	9.97	10.84	
10.84	0.7978	0.7978	0.7978	0.7978	0.7978	0.7978	0.7978	0.7978	0.7978	0.7978	0.7978	0.7978	0.7978	0.7978	0.7978	0.7978	0.7978	0.7978	0.7978	0.7978	0.7978	0.7978	0.7978	0.7978	0.7978
9.97	0.7977	0.7977	0.7977	0.7978	0.7977	0.7978	0.7978	0.7978	0.7978	0.7977	0.7978	0.7978	0.7977	0.7977	0.7977	0.7977	0.7978	0.7977	0.7978	0.7978	0.7977	0.7977	0.7978	0.7978	0.7978
9.14	0.7976	0.7976	0.7976	0.7977	0.7976	0.7977	0.7977	0.7976	0.7976	0.7977	0.7978	0.7977	0.7977	0.7977	0.7977	0.7977	0.7978	0.7978	0.7978	0.7978	0.7978	0.7978	0.7978	0.7978	0.7978
8.34	0.7975	0.7975	0.7975	0.7976	0.7976	0.7976	0.7975	0.7976	0.7976	0.7976	0.7976	0.7976	0.7977	0.7977	0.7977	0.7976	0.7977	0.7978	0.7978	0.7978	0.7978	0.7977	0.7979	0.7979	0.7978
7.58	0.7977	0.7975	0.7977	0.7975	0.7977	0.7974	0.7976	0.7976	0.7976	0.7975	0.7975	0.7977	0.7976	0.7977	0.7978	0.7978	0.7978	0.7978	0.7979	0.7977	0.7979	0.7980	0.7978	0.7978	0.7978
6.86	0.7973	0.7976	0.7975	0.7974	0.7973	0.7973	0.7976	0.7975	0.7977	0.7976	0.7975	0.7976	0.7973	0.7976	0.7974	0.7978	0.7979	0.7979	0.7979	0.7980	0.7979	0.7979	0.7978	0.7978	0.7979
6.18	0.7973	0.7971	0.7973	0.7969	0.7972	0.7971	0.7976	0.7973	0.7973	0.7974	0.7971	0.7973	0.7972	0.7976	0.7976	0.7976	0.7976	0.7976	0.7977	0.7980	0.7981	0.7978	0.7978	0.7979	0.7979
5.53	0.7973	0.7972	0.7971	0.7976	0.7973	0.7971	0.7974	0.7974	0.7974	0.7975	0.7974	0.7972	0.7975	0.7974	0.7972	0.7979	0.7975	0.7979	0.7979	0.7976	0.7978	0.7978	0.7978	0.7978	0.7979
4.91	0.7966	0.7964	0.7966	0.7966	0.7965	0.7968	0.7970	0.7972	0.7966	0.7970	0.7969	0.7967	0.7970	0.7975	0.7974	0.7974	0.7974	0.7978	0.7979	0.7976	0.7980	0.7978	0.7978	0.7978	0.7979
4.33	0.7968	0.7961	0.7960	0.7960	0.7960	0.7962	0.7964	0.7964	0.7966	0.7968	0.7968	0.7974	0.7975	0.7969	0.7978	0.7977	0.7977	0.7983	0.7981	0.7980	0.7982	0.7981	0.7980	0.7982	0.7981
3.79	0.7960	0.7956	0.7953	0.7956	0.7954	0.7955	0.7958	0.7955	0.7958	0.7963	0.7967	0.7960	0.7969	0.7972	0.7967	0.7970	0.7969	0.7970	0.7973	0.7976	0.7979	0.7982	0.7977	0.7979	0.7979
3.29	0.7937	0.7952	0.7945	0.7947	0.7961	0.7938	0.7947	0.7949	0.7949	0.7957	0.7952	0.7949	0.7955	0.7966	0.7962	0.7968	0.7962	0.7968	0.7972	0.7971	0.7979	0.7981	0.7979	0.7979	0.7979
2.82	0.7916	0.7913	0.7915	0.7913	0.7915	0.7922	0.7925	0.7934	0.7922	0.7944	0.7929	0.7943	0.7938	0.7955	0.7953	0.7954	0.7962	0.7962	0.7973	0.7970	0.7988	0.7976	0.7975	0.7976	0.7976
2.38	0.7851	0.7834	0.7821	0.7846	0.7833	0.7856	0.7859	0.7849	0.7837	0.7852	0.7871	0.7895	0.7900	0.7900	0.7914	0.7937	0.7932	0.7948	0.7968	0.7971	0.7983	0.7976	0.7972	0.7975	0.7975
1.99	0.7610	0.7593	0.7615	0.7633	0.7619	0.7593	0.7587	0.7654	0.7641	0.7618	0.7702	0.7655	0.7721	0.7737	0.7793	0.7815	0.7882	0.7918	0.7950	0.7958	0.7972	0.7977	0.7968	0.7970	0.7970
1.62	0.7127	0.7040	0.7115	0.7101	0.7045	0.7092	0.7086	0.7039	0.7062	0.7140	0.7183	0.7253	0.7337	0.7497	0.7555	0.7615	0.7754	0.7809	0.7895	0.7913	0.7958	0.7961	0.7953	0.7952	0.7952
1.30	0.6097	0.6166	0.6097	0.6138	0.6055	0.6237	0.6146	0.6345	0.6267	0.6436	0.6527	0.6614	0.6854	0.7006	0.7162	0.7322	0.7526	0.7692	0.7812	0.7874	0.7930	0.7925	0.7917	0.7921	0.7921
1.01	0.5052	0.5024	0.5071	0.5055	0.5126	0.5064	0.5218	0.5310	0.5573	0.5743	0.5995	0.6151	0.6418	0.6680	0.6911	0.7166	0.7427	0.7603	0.7730	0.7844	0.7891	0.7912	0.7902	0.7898	0.7898
0.76	0.4050	0.4179	0.4161	0.4078	0.4130	0.4397	0.4653	0.4780	0.5062	0.5369	0.5583	0.5941	0.6191	0.6506	0.6790	0.7054	0.7311	0.7566	0.7704	0.7792	0.7867	0.7874	0.7867	0.7867	0.7867
0.54	0.3244	0.3319	0.3414	0.3232	0.3450	0.3796	0.4192	0.4419	0.4697	0.5038	0.5380	0.5716	0.6015	0.6442	0.6649	0.6982	0.7248	0.7471	0.7634	0.7771	0.7829	0.7837	0.7818	0.7815	0.7815
0.36	0.2465	0.2639	0.2783	0.2940	0.3215	0.3555	0.3781	0.4181	0.4469	0.4841	0.5207	0.5547	0.5910	0.6277	0.6591	0.6891	0.7136	0.7397	0.7580	0.7699	0.7783	0.7774	0.7772	0.7771	0.7771
0.22	0.1997	0.2101	0.2366	0.2667	0.2938	0.3299	0.3571	0.3970	0.4341	0.4675	0.5009	0.5394	0.5784	0.6125	0.6492	0.6783	0.7058	0.7336	0.7499	0.7657	0.7713	0.7726	0.7722	0.7725	0.7725
0.11	0.1589	0.1932	0.2122	0.2377	0.2735	0.3148	0.3483	0.3819	0.4159	0.4624	0.4988	0.5302	0.5725	0.6052	0.6417	0.6744	0.7008	0.7270	0.7463	0.7599	0.7662	0.7676	0.7668	0.7671	0.7671
0.04	0.1319	0.1672	0.2036	0.2318	0.2640	0.3031	0.3337	0.3740	0.4166	0.4658	0.4883	0.5272	0.5632	0.5998	0.6343	0.6666	0.6975	0.7198	0.7419	0.7544	0.7615	0.7625	0.7615	0.7620	0.7620

Imitation

Step 1

←

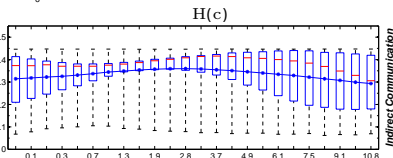
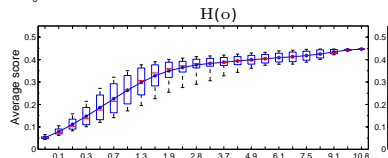
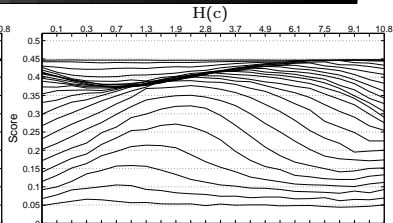
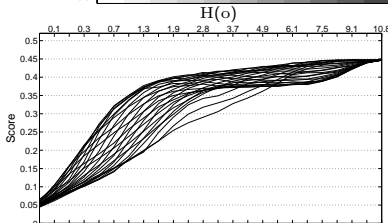
Step 2

←

Step 3

	Observation entropies : H(o)																								
	0.04	0.11	0.22	0.36	0.54	0.76	1.01	1.30	1.62	1.99	2.38	2.82	3.29	3.79	4.33	4.91	5.53	6.18	6.86	7.58	8.34	9.14	9.97	10.84	
10.84	0.0506	0.0683	0.0827	0.1120	0.1329	0.1517	0.1735	0.1968	0.2256	0.2545	0.2793	0.2993	0.3059	0.3277	0.3431	0.3615	0.3803	0.4076	0.4234	0.4328	0.4439	0.4470	0.4472	0.4470	0.4470
9.97	0.0469	0.0646	0.0837	0.1078	0.1290	0.1502	0.1712	0.1978	0.2256	0.2667	0.2990	0.3179	0.3297	0.3422	0.3609	0.3778	0.3933	0.4140	0.4291	0.4385	0.4448	0.4470	0.4472	0.4468	0.4473
9.14	0.0450	0.0657	0.0863	0.1008	0.1203	0.1434	0.1738	0.1950	0.2486	0.2907	0.3245	0.3419	0.3491	0.3652	0.3823	0.3930	0.4073	0.4213	0.4335	0.4398	0.4470	0.4472	0.4465	0.4465	0.4465
8.34	0.0441	0.0618	0.0824	0.1022	0.1200	0.1409	0.1766	0.2188	0.2637	0.3064	0.3394	0.3558	0.3695	0.3779	0.3905	0.4059	0.4180	0.4273	0.4339	0.4454	0.4470	0.4469	0.4464	0.4468	0.4468
7.58	0.0461	0.0702	0.0849	0.1046	0.1222	0.1497	0.1841	0.2233	0.2793	0.3249	0.3541	0.3707	0.3845	0.3945	0.4046	0.4131	0.4231	0.4311	0.4419	0.4450	0.4449	0.4450	0.4451	0.4451	0.4451
6.86	0.0486	0.0677	0.0863	0.1056	0.1274	0.1584	0.2024	0.2542	0.3105	0.3445	0.3696	0.3828	0.3943	0.4005	0.4111	0.4202	0.4268	0.4389	0.4440	0.4448	0.4458	0.4460	0.4465	0.4463	0.4467
6.18	0.0486	0.0671	0.0875	0.1076	0.1313	0.1676	0.2249	0.2817	0.3274	0.3575	0.3752	0.3897	0.4002	0.4079	0.4153	0.4229	0.4325	0.4383	0.4488	0.4469	0.4432	0.4438	0.4438	0.4451	0.4468
5.53	0.0478	0.0730	0.0927	0.1130	0.1470	0.1853	0.2593	0.3062	0.3457	0.3669	0.3827	0.3969	0.4053	0.4119	0.4172	0.4294	0.4327	0.4318	0.4360	0.4362	0.4386	0.4404	0.4438	0.4467	0.4467
4.91	0.0506	0.0721	0.0959	0.1190	0.1556	0.2088	0.2733	0.3255	0.3614	0.3770	0.3872	0.4016	0.4080	0.4141	0.4206	0.4296	0.4293	0.4306	0.4336	0.4353	0.4370	0.4408	0.4442	0.4468	0.4468
4.33	0.0506	0.0701	0.0965	0.1268	0.1726	0.2388	0.3002	0.3446	0.3898	0.3875	0.3941	0.4053	0.4126	0.4202	0.4246	0.4255	0.4271	0.4294	0.4299	0.4300	0.4330	0.4365	0.4436	0.4471	0.4471
3.79	0.0498	0.0745	0.1018	0.1413	0.1971	0.2695	0.3224	0.3579	0.3803	0.3892	0.3989	0.4073	0.4157	0.4187	0.4187	0.4231	0.4227	0.4220	0.4249	0.4294	0.4304	0.4273	0.4432	0.4469	0.4469
3.29	0.0528	0.0741	0.1050	0.1534	0.2298	0.2955	0.3380	0.3570	0.3659	0.3660	0.4037	0.4102	0.4165	0.4128	0.4168	0.4179	0.4182	0.4171	0.4158	0.4152	0.4225	0.4244	0.4425	0.4470	0.4470
2.82	0.0486	0.0790	0.1126	0.1686	0.2504	0.3151	0.3450	0.3732	0.3870	0.3982	0.4028	0.4132	0.4081	0.4109	0.4119	0.4135	0.4108	0.4071	0.4122	0.4128	0.4186	0.4208	0.4414	0.4464	0.4464
2.38	0.0533	0.0809	0.1203	0.1881	0.2643	0.3216	0.3506	0.3770	0.3906	0.4011	0.4066	0.4048	0.4056	0.4054	0.4018	0.3999	0.4014	0.3987	0.4034	0.4102	0.4165	0.4279	0.4403	0.4463	0.4463
1.99	0.0530	0.0830	0.1338	0.2060	0.2717	0.3202	0.3493	0.3731	0.3897	0.3976	0.3947	0.4017	0.4001	0.3958	0.3948	0.3917	0.3913	0.3956	0.3948	0.4023	0.4113	0.4285	0.4404	0.4465	0.4465
1.62	0.0530	0.0898	0.1457	0.2127	0.2681	0.3125	0.3444	0.3705	0.3893	0.3962	0.3907	0.3919	0.3909	0.3867	0.3846	0.3856	0.3862	0.3843	0.3869	0.3990	0.4083	0.4269	0.4404	0.4461	0.4461
1.30	0.0570	0.0929	0.1562	0.2140	0.2546	0.2891	0.3378	0.3690	0.3729	0.3830	0.3816	0.3879	0.3809	0.3784	0.3774	0.3800	0.3791	0.3811	0.3873	0.3899	0.4062	0.4238	0.4391	0.4469	0.4469
1.01	0.0565	0.1033	0.1582	0.2100	0.2535	0.2864	0.3195	0.3447	0.3622	0.3741	0.3772	0.3811	0.3755	0.3741	0.3739	0.3752	0.3779	0.3805	0.3830	0.3926	0.4056	0.4247	0.4397	0.4469	0.4469
0.76	0.0604	0.1051	0.1568	0.1969	0.2339	0.2635	0.2885	0.3269	0.3488	0.3615	0.3685	0.3693	0.3705	0.3718	0.3727	0.3745	0.3757	0.3811	0.3799	0.3886	0.4017	0.4243	0.4403	0.4467	0.4467
0.54	0.0645	0.1002	0.1448	0.1849	0.2159	0.2478	0.2758	0.3061	0.3352	0.3500	0.3577	0.3635	0.3705	0.3736	0.3757	0.3771	0.3747	0.3794	0.3813	0.3890	0.4008	0.4240	0.4392	0.4468	0.4468
0.36	0.0660	0.0956	0.1355	0.1637	0.1899	0.2197	0.2580	0.2882	0.3188	0.3381	0.3560	0.3650	0.3705	0.3741	0.3781	0.3798	0.3828	0.3873	0.3853	0.3899	0.4022	0.4216	0.4405	0.4467	0.4467
0.22	0.0605	0.0917	0.1150	0.1514	0.1777	0.2061	0.2378	0.2714	0.3023	0.3317	0.3498	0.3643	0.3768	0.3797	0.3817	0.3877	0.3910	0.3916	0.3941	0.3958	0.4058	0.4231	0.4420	0.4472	0.4472
0.11	0.0548	0.0779	0.0997	0.1313	0.1552	0.1893	0.2186	0.2520	0.2864	0.3165	0.3372	0.3630	0.3728	0.3845	0.3911	0.3952	0.4016	0.4032	0.4024	0.4090	0.4127	0.4274	0.4422	0.4469	0.4469
0.04	0.0486	0.0676	0.0919	0.1146	0.1400	0.1715	0.2025	0.2317	0.2671	0.2987	0.3307	0.3542	0.3734	0.3890	0.3970	0.4031	0.4091	0.4116	0.4141	0.4194	0.4265	0.4305	0.4422	0.4464	0.4464

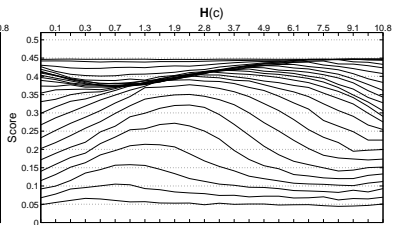
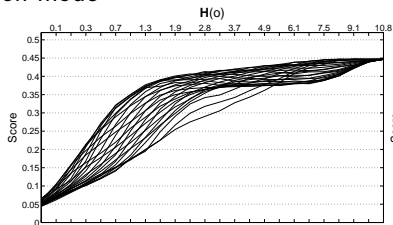
Indirect communication



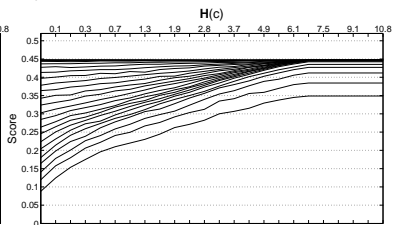
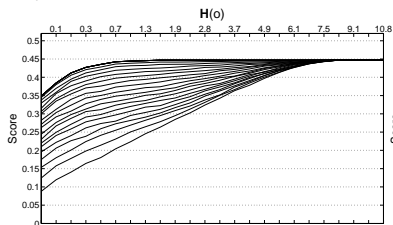
Indirect Communication

Communication mode

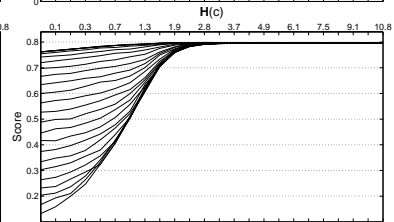
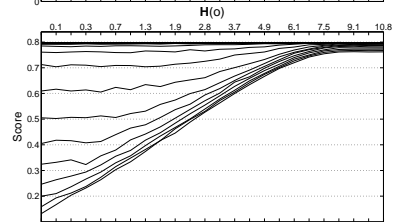
Stigmergy

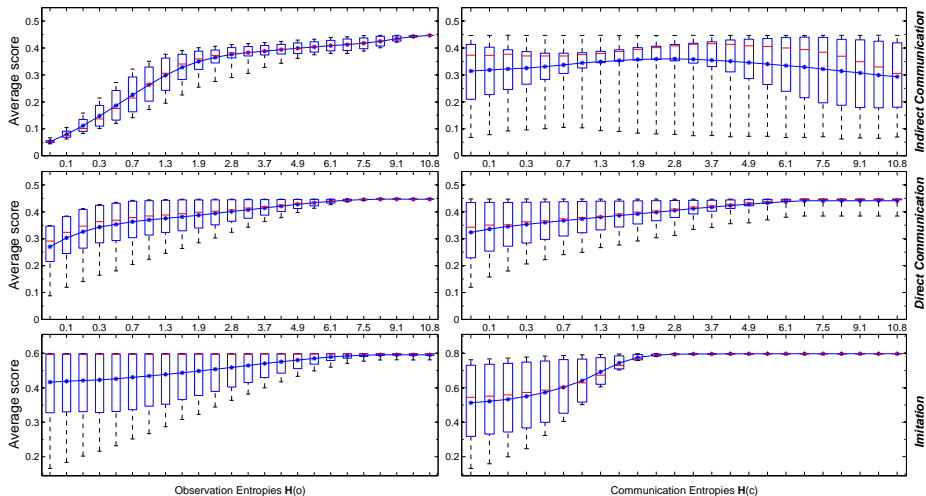


Direct



Imitation





Whisker plot showing the variation in test scores across different entropy values $H(c)$ for fixed entropies $H(o)$ (left-side), and vice versa (right-side). The central mark (in red) is the median while the edges of the box represent the 25th and 75th percentiles of the scores and the whiskers extend to the most extreme score values. The blue line-plot shows the average scores at each of the intermediate entropy values.

Coefficient of success ϕ

Definition

Communication-over-observation coefficient of success is:

$$\phi = \frac{\text{success}(\text{communication} > \text{observation})}{\text{number of tests}}$$

- ϕ close to 1 \implies communication has higher influence on intelligence
- ϕ close to 0 \implies observation has higher influence on intelligence

Coefficient of success ϕ

Motivation

Methodology

Experiments

Results

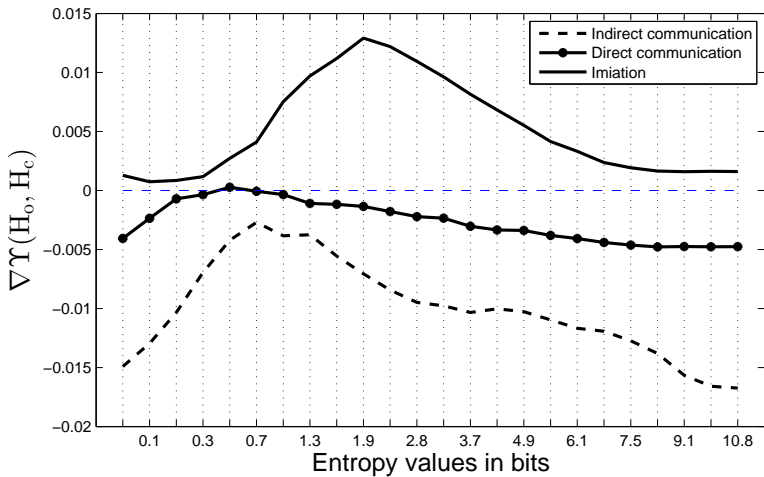
Conclusions
and future
work

Q/A

References

<i>Communication mode</i>	ϕ	<i>Dependency</i>
stigmergy	0.0399	high
direct communication	0.9855	low
imitation	0.9928	low

Entropy analysis



Presentation Outline

- 1 Motivation
- 2 Methodology
- 3 Experiments
- 4 Results
- 5 Conclusions and future work**

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Scope of experiments

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- Different Environment spaces
 - different uncertainties
- Number of evaluated agents
- Heterogeneous collectives
- Different communication methods
- Measure *learning* rather than *search*
 - how does the collective performance vary with the task/problem complexity ?

What is the bottom line re communication and observation?

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- Dull systems with low perception abilities can be re-compensated for and significantly improved by introducing a **measurable** amount of communication
- Increasing the communication entropies **does not** imply a **monotonic** increase in intelligence
- Communication can be highly **dependent** on observation and this dependency can be **measured**
- There exists a **restricted range of entropies** where communication achieves its **highest influence** on the group's intelligence

Observation, Communication and Intelligence in Agent-Based Systems

Possible applications:

- searching for a moving target while avoiding injury (collective pursuit problems)
- nest selection when there is one and only one best nest
- pattern recognition problems

**Thank you
Questions?**

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References



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Coefficient of success

Definition

Let $S = \{(x, y) \in E \times E \mid x > y\}$. The communication-over-observation coefficient of success is: $\phi = (\sum_S \inf(\mathcal{Y}(x, y), \mathcal{Y}(y, x)) \div |S|)$, where $\inf(a, b)$ is a function that returns 1 if $a < b$, or zero otherwise.

Knowing that the test scores are of the form $\mathcal{Y}(H_o, H_c)$:

- ϕ close to 1 \implies communication has higher influence on intelligence
- ϕ close to 0 \implies observation has higher influence on intelligence