

A parameterless biologically inspired control algorithm robust to nonlinearities, dead-times and low-pass filtering effects

Fabio DallaLibera^{*‡} Shuhei Ikemoto[†] Takashi Minato[‡] Hiroshi Ishiguro^{‡§} Emanuele Menegatti^{*} Enrico Pagello^{*}

^{*}Department of Information Engineering, University of Padua, Padova, Italy

[†]Department of Multimedia Engineering, Osaka University, Osaka, Japan

[‡]ERATO, Japan Science and Technology Agency, Osaka, Japan

[§]Department of Systems Innovation, Osaka University, Osaka, Japan

Abstract

An algorithm for robot control inspired by Escherichia Coli chemotaxis is presented. A model that explains the performance increase due to introduction of random perturbations is provided and used to derive a completely adaptive and parameterless algorithm. Practical applicability is shown by a mobile robot navigation task in which the robot structure is unknown and the robot undergoes hardware damages.

Related works

Escherichia Coli proceeds by alternating two movements

CCW rotation

straight movement

CW rotation

random direction change

If the conditions improve: longer straight movements = less random changes

Biased random walk toward food sources

Robots with similar strategy [2]

Proceed by alternating two behaviors:

- Straight movement
- Random rotation

Advantages

- Robust to noise
- Do not stuck in local minima
- Robots distribute among multiple sources

Problem

The robot may not reach the target due to hardware faults

Example: An encoder breaks

“go forward” becomes “spinning”

Advanced fault detection and recovery techniques [3,4]

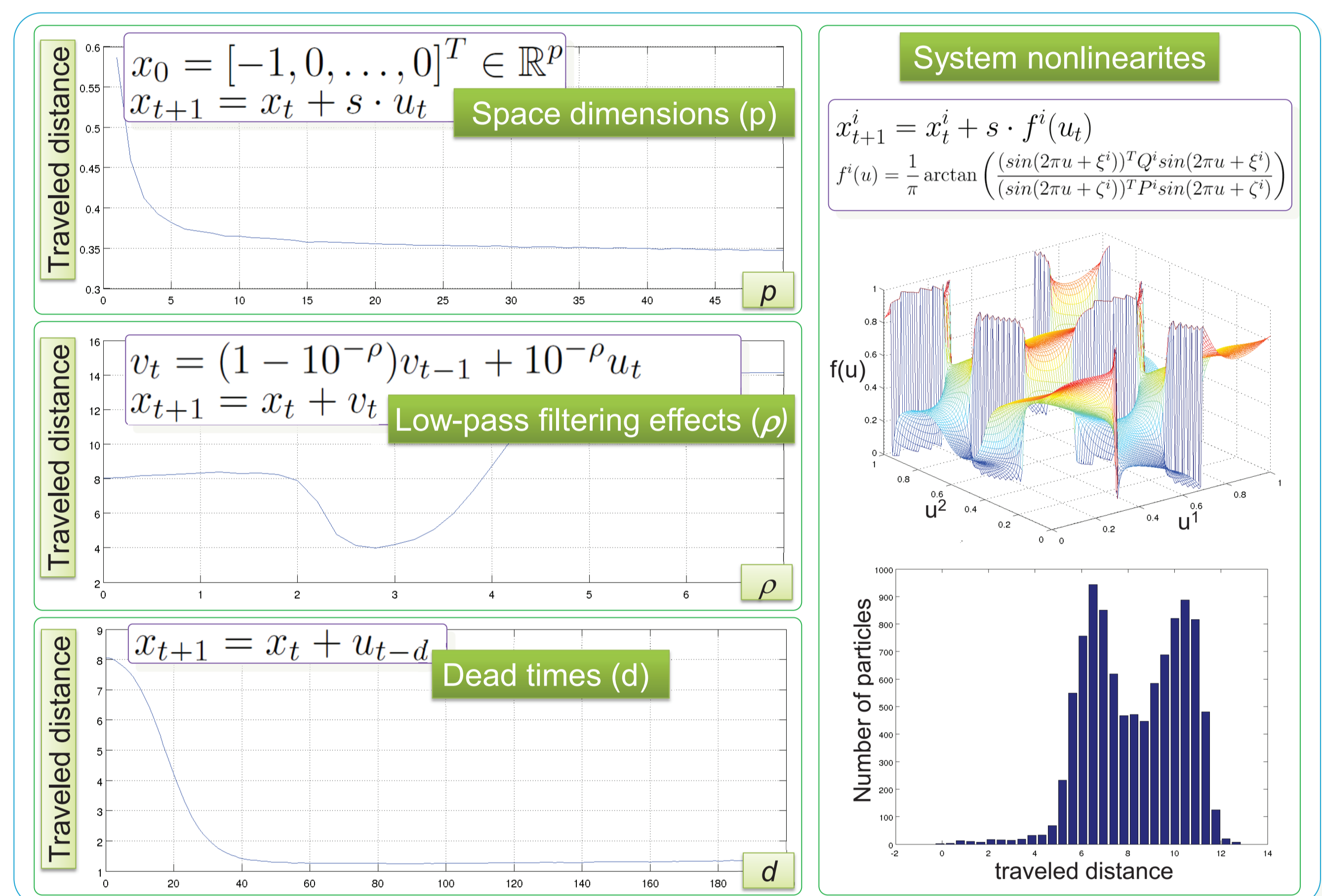
Require precise identification of possible faults or very high computational power

Proposed approach

Biased random walk directly in the command space

- Extreme simplicity
- Appropriate behaviors that exploit the working hardware are found

Performance measurements



Experiment

Setup

Robot: simulated mobile robot with two independent wheels

Sensory information: number of red pixels in a simulated omnidirectional camera

Task: reach a red hemisphere

Simulated four damages

Change in the size of a wheel

Uncontrollability of a wheel

Change of the rotation axis

Obscuration of 20% of the camera

Automatic noise adaptation based on the estimation of the motor command variance

$$\delta_0^i = 1.1$$

$$\sigma_t^i = \frac{(u_t^i - u_{t-1}^i)^2}{2}$$

$$\delta_{t+1}^i = \begin{cases} 1/\delta_t^i & \text{if } t \text{ odd} \wedge \sigma_t^i \geq \sigma_{t-2}^i \\ \delta_t^i & \text{otherwise} \end{cases}$$

$$\eta_{t+1}^i = \begin{cases} \eta_t^i \delta_{t+1}^i & \text{if } t \text{ odd} \\ \eta_t^i & \text{otherwise} \end{cases}$$

Control algorithm

If the conditions improved (e.g. got closer to the goal) then keep the previous motor command but add a random perturbation to it else choose randomly a new motor command

Adding random perturbations of opportune magnitude to the control input improves the performance

$$u_{t+1}^i = \begin{cases} u_t^i + \eta_t^i R & \text{if } \Delta A_t \geq 0 \\ \text{random selection} & \text{otherwise} \end{cases}$$

Example: 2D case

Distance traveled toward the goal vs Noise standard deviation η

heading Goal
Robot optimal heading

Probability density function

Deviation from optimal heading [deg]

$\eta=0.0056$
 $\eta=0.056$
 $\eta=0.3$

References

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