

Spike timing dependent plasticity (STDP) to make robot navigation intelligent

Akira Imada

Brest State Technical University (Belarus)

This is not a success report
but
a guideline of our ongoing project
or
a report of what we are currently planning.

To start with
a benchmark to know
if what we are planning will work or not.

Path Planning

or

Robot Navigation

⇓ where

aim is usually to **minimize** the route from start to goal, but...

**Task of minimizing navigation
on its own
is not so difficult**



A-star, Genetic Algorithm, Reinforcement Learning, Neural Network,
Ant Colony Optimization, etc...

Genetic Algorithm (GA)



Agent decides actions following its **chromosome**

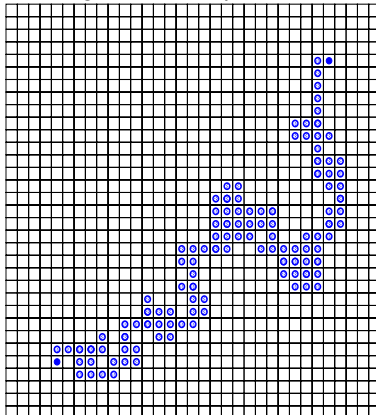
e.g.

with chromosomes like

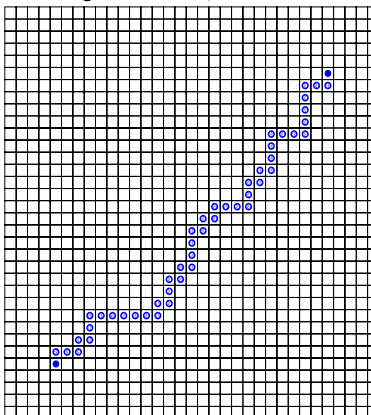
(4 1 1 2 1 3 2 3 3 4 2 1 3 4 3 2 3 3 2 1 1 3 2 1 3 ...)

Random walk evolved to be minimized

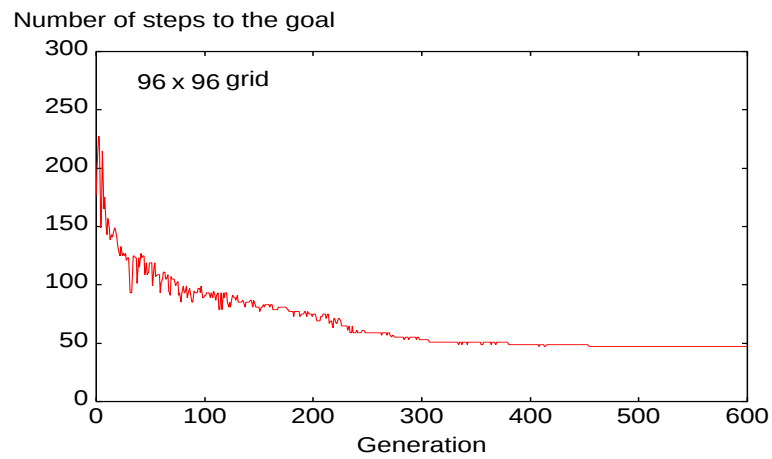
96x96 grid 178 steps



96x96 grid 48 steps



Task was easy



Let's extend it

From Minimization to Maximization



We don't care goal points but to maximize an exploration

A Camel in a Desert

The 52nd problem

in the

“Propositiones ad acuendos inventes” (in Latin)

by

Alcuin of York (732–804)



Can a camel maximize its exploration in a desert
surviving with grains on his back?

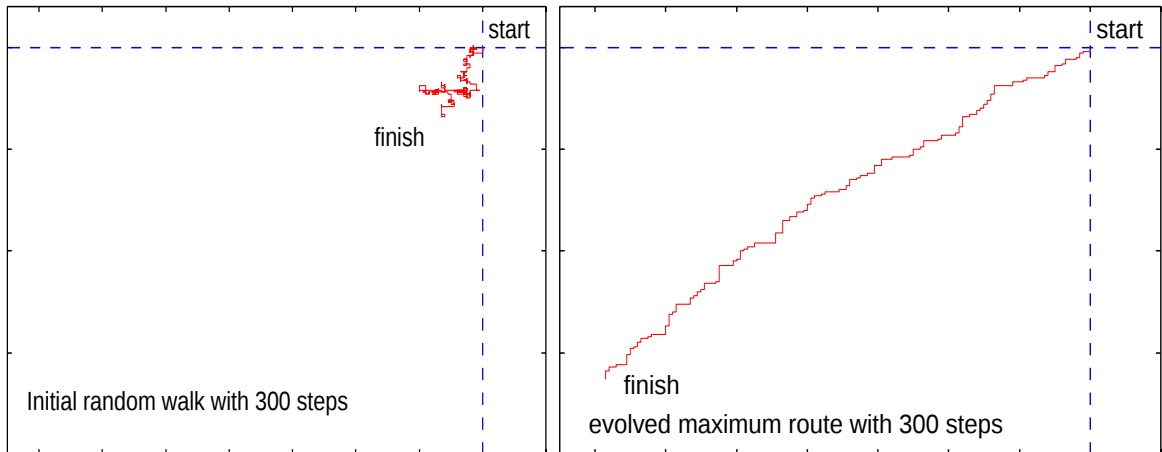


Modern version: From Camel to Jeep in a Desert

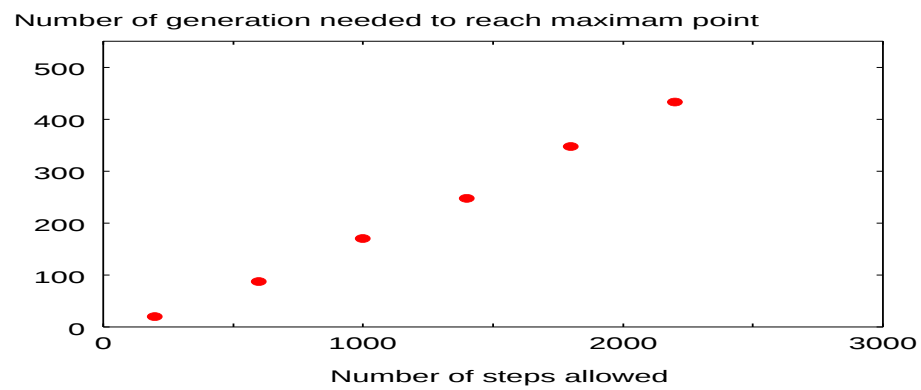
Can a jeep maximize its exploration
in a desert with a tank of fuel?



Still it's not so difficult by GA for example



Task was rather easier



The complexity is $O(N)$

Why not extend it further

**Let's make it return to the point
it started**



while also maximizing an exploration

Planet Land-rover Problem

From A (start) to B (goal) minimizing its route



From A to A maximizing its route

(This is our proposal here as a benchmark)

Can the rover return to the base
with its exploration being maximum
spending a limited energy given when it starts?



Much more demanding



Taking GA as an example, what might be a fitness?

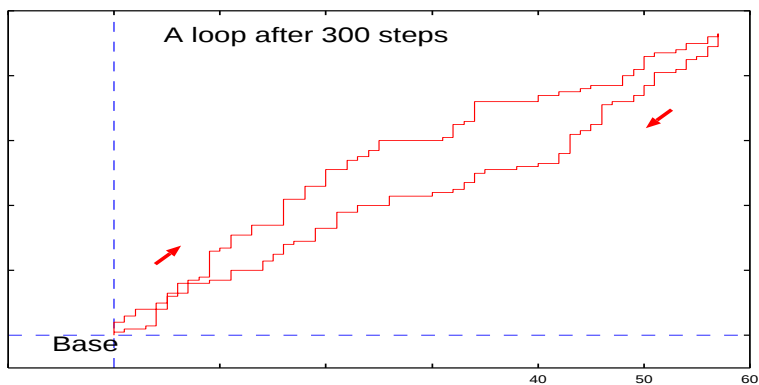
e.g.

{total length} & {how often it has crossed previous route?}

with

Multi Objective GA (MOGA)

A Heuristic create such a route



What heuristic do you guess?

Topic of today's talk:

What is intelligence?



Intelligence should be spontaneous, flexible, or unpredictable more or less

Stolle et al. (2002)

“... every day we might be cooking a different breakfast, but the kitchen layout is the same from day to day.”

“I beg your pardon?”

Do they change the expression for an easier understanding?

or

Do they repeat the expression, maybe with a bigger voice?

What if your canary stop singing?

Legendary three strategies in Japan.

- (1) Just wait until she sings again.
- (2) Try something so that she sings again.
- (3) Kill her if she doesn't sing any more?



**To be intelligent,
action should be different more or less
even in an exactly identical situation**

Performance is not intelligent

if we use

deterministic A-star, GA, NN with fixed weights?

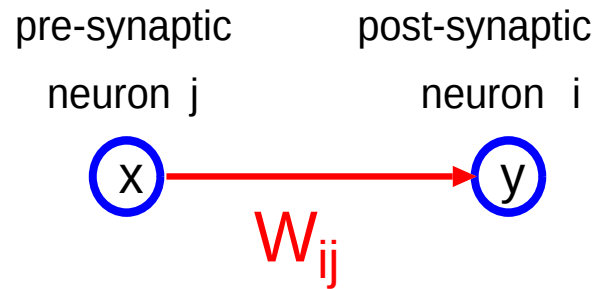
**Let's make a learning occur
during behavior**



Evolution of Learning

A notation

Hebbian Learning of NN



$$w_{ij}(t + 1) = w_{ij}(t) + \eta x_j(t) y_i(t) \quad (x_j, y_i \in [0, 1])$$

Floreano's approach (2000)



Modification of w_{ij} during exploration
with either one of four Hebbian and Hebbian-like rules

Hebbian learning

$$(1) \quad \Delta w = (1 - w)xy$$

Hebbian type learning

Weaken if the post-synaptic is active, while the pre-synaptic is not

$$(2) \quad \Delta w = w(-1 + x)y + (1 - w)xy$$

Weaken if the pre-synaptic is active, while the post-synaptic is not

$$(3) \quad \Delta w = wx(-1 + y) + (1 - w)xy$$

And

Strengthen if the two have similar activity and weaken otherwise.

$$(4) \quad \Delta w = \begin{cases} (1 - w)F(x, y) & \text{if } F(x, y) > 0 \\ wF(x, y) & \text{otherwise} \end{cases}$$

where

$$F(x, y) = \tanh(4(1 - |x - y|) - 2)$$

Evolution of leaning

Which rule and what parameter's value should be assigned
to each of the synapses?



Evolution leads the combination to the optimum.

Stanley's implementation (2003)

$$\Delta w = \begin{cases} \eta_1(1-w)xy + \eta_2wx(y-1) & \text{for excitatory neuron} \\ -\eta_1(1-w)xy + \eta_2(1-w)x(1-y) & \text{for inhibitory neuron} \end{cases}$$

$\Downarrow$

Evolution is on (η_1, η_2) of each synaptic weight

Their result

Starting with random weights every time anew
the weights are modified step by step by the rule it learned

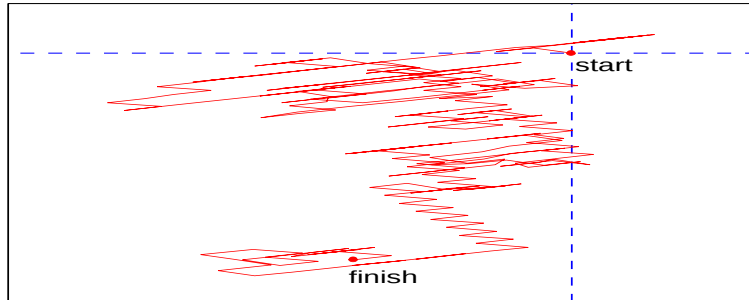


Every run is different depending on the initial random weights

More general learning (Durr et al. 2008)

$$\Delta w = \eta(Axy + Bx + Cy + D)$$

An example of exploration by a recurrent NN with random weights



Can it learn so that path will finish at the start point?

Our Aim



Learning should occur during a random exploration

with a **spiking neural network** seeking its biological plausibility

Learning by STDP

(Spiking neuron's version of Hebbian Learning)



The topic is still open!

Meunier et al. (2005)



*“Up to now, nobody has been able to show how it is possible to learn
with STDP...”*

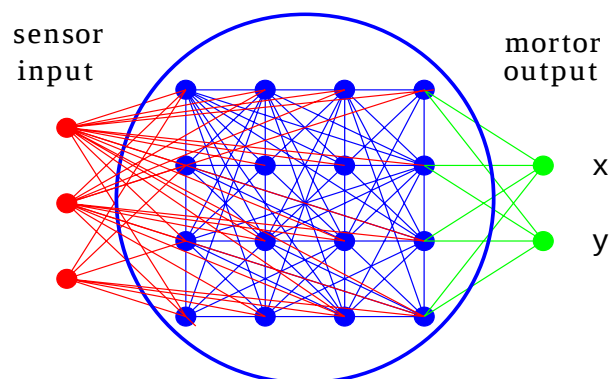
Farries et al. (2007)



“Although synaptic plasticity is widely believed to be a major component of learning, it is unclear how STDP itself could serve as a mechanism for general purpose learning.”

Why not challenge?

Architecture we are planning

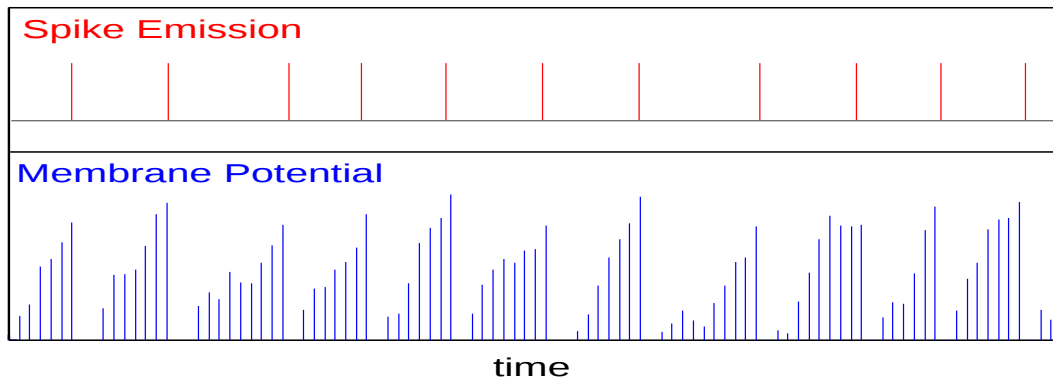


$\Downarrow$ a.k.a.

Echo State Network, Liquid State Network or Reservoir Computing

Integrate-and-Fire Model

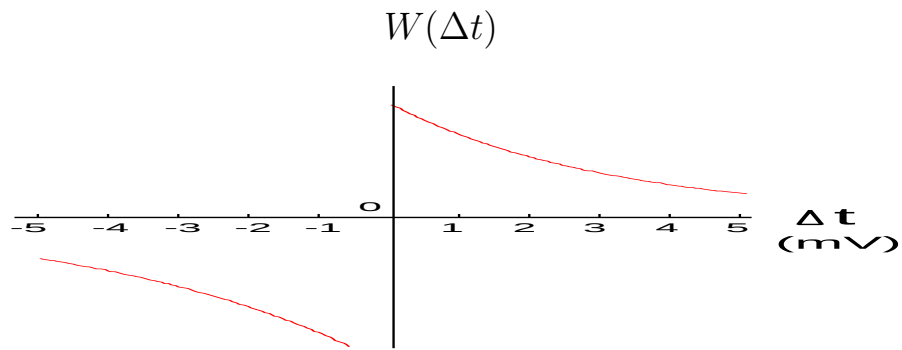
$$u_i(t) = u_r + (u_i(t - \delta t) - u_r) \exp(-\delta t / \tau) + \sum_j w_{ij} f_j(t - \delta t)$$



What is STDP?

$$W(\Delta t) = \begin{cases} A_+ \exp(-\Delta t/\tau_+) & \text{if } \Delta t \geq 0 \\ -A_- \exp(-\Delta t/\tau_-) & \text{if } \Delta t < 0 \end{cases}$$

where $\Delta t = t_{post} - t_{pre}$



In short

potentiation when a pre-synaptic neuron fires
shortly before a post-synaptic neuron
and

depression when the pre-synaptic neuron fires shortly after.

It's too basic to be used

Reward-modulated STDP Learning

(Florian 2007)

$$w_{ij}(t + \delta t) = w_{ij}(t) + \gamma r(t + \delta t) \zeta_{ij}(t)$$

where

$$\zeta_{ij}(t) = P_{ij}^+(t) f_i(t) + P_{ij}^-(t) f_i(t)$$

$$P_{ij}^+(t) = P_{ij}^+(t - \delta t) \exp(-\delta t / \tau_+) + A_+ f_j(t)$$

$$P_{ij}^-(t) = P_{ij}^-(t - \delta t) \exp(-\delta t / \tau_-) + A_- f_j(t)$$

Reinforcement Learning (RL)

What an agent learns is a *Policy*

and

Policy is how to select an *Action* in a given *Situation (State)*

maximizing

total *Rewards* it occasionally will receive from the environment

**Policy indicates agent
which action should be chosen
in any possible situation**



While in GA

Agent decides actions following its **chromosome**

in RL

Agent decides actions following its **policy**

However

our world is with **no obstacle, no wall, no corridor**, only one goal



Everywhere no reward except for one point (goal)



A needle in a Haystack

(Rewards are not likely to be encountered by a random exploration.)

We have not succeeded yet



**Nevertheless we think of
a further extention of the problem
as a bigger challange**



What if the rover carries container(s) for fuel?

Jeep Problem

Where jeep explore is 1-D desert.

Jeep can unload its fuels anywhere in the desert.

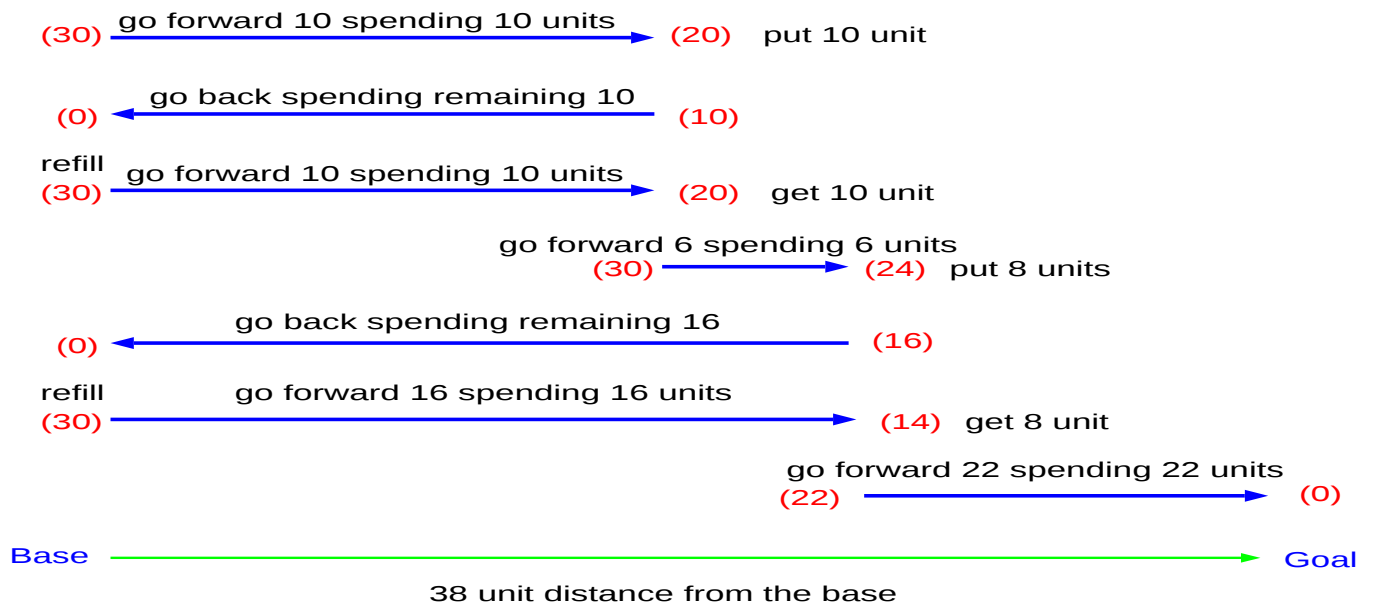
Fuels can be filled only at the base.

Jeep can go back to the base n times to re-fill its tank.

↓ thus

The jeep should maximize its penetration

An example of a success



**Let's give the rover the same condition
(though in 2-D)**



A very tough benchmark and infinitely large number of solutions



Good as a benchmark to check *so-far-mentioned* intelligence

Let me conclude

To claim artificial NN to be intelligent
a different action more or less should be made under identical situation.



We want to, or we are going to realize it
by
Spiking NN with learning by STDP
also
to be more biologically plausible.

SONY's AIBO



can excellently learn but still repeats same action in the same situation

**Hoping collaborations
to design
an agent like real human intelligence**

Thank You!