# **Pseudo-Random Number Generators** Functional Programming and Intelligent Algorithms

Prof Hans Georg Schaathun

Høgskolen i Ålesund

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Review

### Outline





- 3 Random initial weights
- 4 Closur



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# Review of yesterday

- + What did you learn yesterday?
- △ What is your greatest challenge?



Review

# The implementation of the perceptron

- How does your neural network perform?
- What needs improvement



# Outline





- 3 Random initial weights
- 4 Closur



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Randomness

# Randomness

What is randomness?



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# Probabilistic programs

#### How do we create probabilistic computer programs? I.e. how do we make the computer act at random?



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# Two options

#### True randomness uses physical sources of entropy

- /dev/random on many systems
- 2 random-fu in Haskell

# Pseudo-random number generators (PRNG) are deterministic but random-*looking*

- random, standard package in Haskell
- random-tf, more recent Haskell package

# Linear Congruential Generators A classic PRNG

- Recurrence  $x_i = a + cx_{i-1} \mod m$
- Seed (initial state) x<sub>0</sub>
- Pseudo-random sequence [*x*<sub>0</sub>, *x*<sub>1</sub>, *x*<sub>2</sub>, ...]
- Known as Lehmer's algorithm

Will this pseudo-random sequence look random?



#### Ciphers in counter mode PRNG through cryptography

- Cipher  $e_k(x) = y$
- Pseudo-random sequence [x<sub>0</sub>, x<sub>1</sub>, x<sub>2</sub>,...] where
  x<sub>i</sub> = e<sub>k</sub>(i)

Why does this give good randomness?

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# The PRNG is a state machine

#### The state decides what the PRNG will output

- Lehmer's algorithm  $x_{i-1}$  is the state
- Counter mode i is the state
- State transition
  - Lehmer's algorithm  $x \mapsto a + cx \mod m$
  - Counter mode  $i \mapsto i + 1$
- Output function
  - Lehmer's algorithm  $x \mapsto a + cx \mod m$
  - Counter mode  $i \mapsto e_k(i+1)$

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# State machines in functional programming

How do we handle state machines in functional programming?

- What is special about functional programming?
- What is difficult?
- How can we do it?



# Outline



### 2 Randomness



#### 4 Closur



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#### Exercise

Given a TFGen object, how do you generate an random, infinite list of Word32 objects?



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# Splitting a PRNG

# Problem. After you have generated the infinite list, how do you get the updated state?

- 1 split :: TFGen -> (TFGen,TFGen)
- 2 (g',newstate) = split g
- Use g' to generate the list
- 🕘 newstate **is your new state**

#### Exercise

Given a TFGen object, how do you generate an random, infinite list of Word32 objects?

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# Splitting a PRNG

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- Use g' to generate the list
- Inewstate is your new state

#### Exercise

Given a TFGen object, how do you generate an random, infinite list of Word32 objects?

### Question

#### Where do you get the initial state?



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# **Different options**

- Hardcode an arbitrary seed
- Use initialisation functions in the library
  - initTFGen
- Use a library which provides true random values
  - random-fu

12 N 4 12

Closure

### Outline



- 2 Randomness
- 3 Random initial weights





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# **Tuning parameters**

#### Distribution of random initial weights?

- 2  $\beta$  in the sigmoid function?
- Number of iterations?

Artificial intelligence is not an exact science.

- Trial and error.
- Test different choices.

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# **Tuning parameters**

- Distribution of random initial weights?
- **2**  $\beta$  in the sigmoid function?
- Number of iterations?

Artificial intelligence is not an exact science.

- Trial and error.
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# Some guidelines

- Weights:  $-1/\sqrt{n} \le w \le 1/\sqrt{n}$ 
  - where n is the number of inputs to the layer
- The weights should have similar magnitude
- Small  $\beta \beta \leq 3$ 
  - **(1)**  $\beta = 1$  is a good starting point

# Number of epochs



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### Exercise

- Update the initNeuron and initNetwork functions with randomised weights
- You need to import a library, either
  - System.Random; or
  - System.Random.TF
- Initialise the PRNG
- Generate and use the weights for the network