

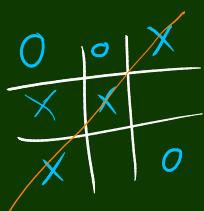
# Playing cationnally against irrational players with monads

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Example: Tic-Tac-Toe ( $\times$  plays first)



$\times$  wins

X	X	0
0	O	X
X	O	X

draw

- If both players play rationally  
then the optimal outcome of the game is a draw. } use monads J & K + explain this
- But what if one of them, or both, play irrationally? } use monads T, J<sub>T</sub> & K

Example:  $T$  = probability distribution matrix

1. Both players are rational:

The game is a draw with probability 100 %

2.  $X$  is rational and  $O$  plays randomly with uniform distribution:

The game is a draw with probability 0.5 %  
and an  $X$  win with probability 99.5 %

3.  $X$  plays uniformly randomly and  $O$  plays rationally:

draw 8.4 %

$O$  win 91.6 %

4. Both play randomly:

draw 12.7 %

$X$  win 58.5 %

$O$  win 28.8 %

Example:  $T =$  non-empty powerset monad

This accounts for one or more players playing erratically  
(with no assigned probabilities)

1. Both erratic:  
game can be any of X win, draw, O win.
2. One erratic:  
game can be any of the other wins or draw
3. Both rational:  
draw

Example:  $T =$  Identity monad

This amounts to our previous work,  
with all players rational.

What is presented here generalizes  
previous work.

# Plan of the talk

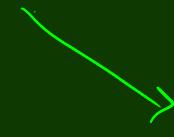
1. • How to specify combinatorial games of perfect information using a certain monad  $K$  (double dualization) and use it to calculate optimal outcomes.
  - How to use a certain monad  $J$  (selection) and a monad morphism  $J \rightarrow K$  to calculate optimal strategies.
2. How to add irrationality with an affine strong monad  $T$ ,  
an algebra  $\alpha: TR \rightarrow R$ ,  
and a generalization  $J_T$  of  $J$ .  
(when  $T = \text{Id}$ , we have  $J_T = J$ )
  - $\eta_1: 1 \rightarrow T1$   
is iso

## Assumption

We work in a cartesian closed category.

Examples we have used so far for applications:

- Category of sets
- Kleene-Kreisel spaces
- Scott domains
- ( $\infty$ -)toposes
- compactly generated spaces
- Alex Simpson's QCB spaces
- TYPES



game theory  
proof theory  
searchable sets

## The monads K & J: motivation

X set of goods.

R set of prices.

p:  $X \rightarrow R$  table of prices.

$\varepsilon: (X \rightarrow R) \rightarrow X$  selects a cheapest good in a given table.

$\phi: (X \rightarrow R) \rightarrow R$  selects the lowest price in a given table.

Fundamental equation:  $\underbrace{p(\varepsilon p)}_{\text{price of cheapest good}} = \underbrace{\phi p}_{\text{lowest price in table}}$

$$\phi = \min$$

$$\varepsilon = \arg\min$$

$$p(\arg\min p) = \min p$$

## The monads K & J: motivation

X set of possible plays in a game

R set of possible outcomes

p:  $X \rightarrow R$  the outcome of a play

$\varepsilon: (X \rightarrow R) \rightarrow X$  selects an optimal play

$\phi: (X \rightarrow R) \rightarrow R$  selects the optimal outcome

Fundamental equation:  $\underbrace{p(\varepsilon p)}_{\text{outcome of optimal play}} = \underbrace{\phi p}_{\text{optimal outcome of the game}}$

## The monads K & J: motivation

X set of things

0 < 1

R set of truth values

false true

p: X → R predicate

⊓ = max

ε: (X → R) → X Hilbert's ε

ϕ: (X → R) → R existential quantifier

Fundamental equation: p(ε p) =  $\underbrace{\phi p}_{\exists x. p x}$

ϕ = ∃

ε = Hilbert's ε

Hilbert's definition of ∃  
in his ε-calculus.

## The monads K & J: motivation

X set of things

0 < 1  
false true  
 $\forall = \min$

R set of truth values

p:  $X \rightarrow R$  predicate

$\varepsilon: (X \rightarrow R) \rightarrow X$

$\phi: (X \rightarrow R) \rightarrow R$  universal quantifier

Fundamental equation:  $p(\varepsilon p) = \underbrace{\phi p}_{\forall x. p x}$

$\phi = \forall$

$\varepsilon =$  in every pub  $p$  there is a person  $x$  such that  
if  $x$  drinks then everybody drinks.

## The monads K & J : definitions

double dualization (aka continuation monad)

monad:

- $KX = \underbrace{(X \rightarrow R)}_P \rightarrow R$
- $\eta_X : X \xrightarrow{\phi} KX$   
 $x \mapsto \lambda p. px$

- Given  $f : X \rightarrow Y$  we get

$$Kf : KX \rightarrow KY$$

$$\phi \mapsto \lambda p. \phi(\lambda x. p(fx))$$

- $\mu_X : K(KX) \rightarrow KX$   
 $\bar{\phi} \mapsto \lambda p. \bar{\phi}(\lambda \phi. \phi p)$

selection monad:

- $JX = \underbrace{(X \rightarrow R)}_P \rightarrow X$
- $\eta_X : X \xrightarrow{\varepsilon} JX$   
 $x \mapsto \lambda p. x$

selection monad

- $Jf : JX \rightarrow JY$

$$\varepsilon \mapsto \lambda p. f \varepsilon((\lambda x. p(fx)))$$

- $\mu_X : J(JX) \rightarrow JX$

$$\varepsilon \mapsto \lambda p. \varepsilon(\lambda \varepsilon. p(\varepsilon p)) p$$

Mond morphism  $J \rightarrow K$

$$\theta_X : JX \rightarrow KX$$

$$\varepsilon \mapsto \lambda p. p(\varepsilon p)$$

We often write  $\bar{\varepsilon}$  for  $\theta \varepsilon$ .

$$\bar{\varepsilon} p = p(\varepsilon p)$$

c.f. fundamental equation

E.g.  $\overline{\text{argmax}} = \text{max}$

" $\text{argmax}$  is a selection  
function for  $\text{max}$ "



We have monoidal monad structures

Because we have strong monads  $T = J$  and  $T = K$  on  $\text{C.C.}$

$$T(X \times T(Y) \xrightarrow{\otimes} T(X \times Y))$$

We want this  $\xrightarrow{\text{left-to-right}}$   $(\mu, \nu) \mapsto (T(\lambda x. t_{X,Y}(x, \nu))) \mu$

not this  $\xrightarrow{\text{right-to-left}}$   $(\mu, \nu) \mapsto (T(\lambda y. t_{Y,X}(\mu, y))) \nu$

The monads are not commutative.

We have monoidal monad structures

Because we have strong monads  $T = J$  and  $T = K$  on  $\mathbf{C}\mathbf{C}\mathbf{C}$ .

We use this  
to calculate  
optimal outcomes

We use this to  
calculate  
optimal strategies

$$\begin{array}{ccc} KX \times KY & \xrightarrow{\otimes} & K(X \times Y) \\ \uparrow \theta_X \times \theta_Y & & \uparrow \theta_X \\ JX \times JY & \xrightarrow{\otimes} & J(X \times Y) \end{array}$$





## Fundamental theorem (with easy proof)

For any  $\varepsilon : JX$ ,  $\delta : JY$  we have

$$\overline{\varepsilon \otimes \delta} = \overline{\varepsilon} \otimes \overline{\delta}.$$

In other words, if

$\varepsilon$  is a selection function for  $\phi : KX$

$\delta$  is a selection function for  $\psi : KY$

then  $\varepsilon \otimes \delta$  is a selection function for  $\phi \otimes \psi$ .

## Example

In every school there is a child  $x_0$  and a teacher  $y_0$  such that if  $x_0$  gives an apple to  $y_0$  then every child gives an apple to some teacher.

$X = \text{set of children}$

$Y = \text{set of teachers}$

$$(x_0, y_0) := (\exists \text{rg-}A \otimes \exists \text{rg-}E)(p)$$

$P(x, y) = x \text{ gives an apple to } y$

Definition of history free game  
of fixed length  $n$  for simplicity

1. Sets of moves  $X_0, X_1, \dots, X_{n-1}$
2. Quantifiers  $\phi_0: K X_0, \dots, \phi_{n-1}: K X_{n-1}$
3. Outcome function  $p: X_0 \times X_1 \times \dots \times X_{n-1} \rightarrow R$ 
  - Then the optimal outcome is  $(\phi_0 \otimes \phi_1 \otimes \dots \otimes \phi_{n-1})(\Phi)$
  - If the quantifiers have selection functions  $\varepsilon_0, \varepsilon_1, \dots, \varepsilon_{n-1}$ ,  
then an optimal play is given by  $(\varepsilon_0 \otimes \varepsilon_1 \otimes \dots \otimes \varepsilon_{n-1})(P)$

## Calculation of optimal strategy

If the moves  $x_0, x_1, \dots, x_n$  have been played, then the next optimal move to play is

$$x_{k+1} = \text{first move of } (\varepsilon_{k+1} \otimes \varepsilon_{k+2} \otimes \dots \otimes \varepsilon_{n-1}) \\ ((\lambda_{x_{k+1}, x_{k+2}, \dots, x_{n-1}} \cdot p(x_0, \dots, x_{n-1})))$$

History dependent (as e.g.  $T_1 \leftarrow T_2 \leftarrow T_0$ )

In brief summary

$$\otimes^k : kX \times (X \rightarrow kY) \rightarrow k(X \times Y)$$

$$\otimes^J : JX \times (X \rightarrow JY) \rightarrow J(X \times Y)$$

↑  
previous move

## Irrational players

Consider an additional monad  $T$   
(e.g. of probability distributions)

with a monad algebra  $\alpha: TR \rightarrow R$

$$J_T X = (X \rightarrow R) \rightarrow TX$$

- Examples
1. Select an argument of  $p:X \rightarrow R$  with uniform distribution among maximal values (rational player)
  2. Select all elements of  $X$  with uniform distribution (irrational player)

How do we compute optimal outcomes and  
optimal strategies?

As before:

- The mound  $J_T$  is also strong.
- We use  $\otimes_{J_T}$  to compute strategies in equilibrium  
(this generalizes Nash equilibrium)

→ We implemented them in Haskell for the purpose  
of computation.  
→ In Agda for the purposes of verification.

