## What is Linear Logic?

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## Linear Logic seems to be everywhere.

How I met LL: as a natural structure of a model of sequential computations (strong stability $\sim$ hypercoherences in 1992).

The same thing happened

- earlier for Berry's stable semantics (stability $\leadsto$ coherence spaces): this is how Girard discovered LL
- later for Scott semantics (Scott continuity $\leadsto$ prime algebraic complete lattices, Krivine, Huth, Winskel).
We know now dozens of models of programming languages which can nicely be described as models of LL.


## In what does it differ from usual logic?

- Logic is usually thought of as a formalism to express and prove properties of things.
- In the 20th century one understood that proofs are programs (Gentzen cut-elimination, Gödel Dialectica, Curry Howard correspondence). Logical formulas become types.
- Linear logic is a Curry Howard logic: LL formulas are types.
- LL shows up when one builds universes (categories) of spaces and morphisms representing computations between them.
This talk: illustration on the example of probabilistic coherence spaces.


## The category of substochastic matrices

## Example

For some reason, we want to build a simple theory of subprobabilistic distributions and substochastic matrices acting on them.

It can be described as a category:

- objects are sets $I, J, \ldots$
- a morphism $I \rightarrow J$ is a matrix $s \in\left(\mathbb{R}_{\geq 0}\right)^{I \times J}$ such that $\forall i \in I \quad \sum_{j \in J} s_{i, j} \leq 1$.
So a matrix $s: I \rightarrow I$ is a submarkovian chain (we accept loss of mass: possibly diverging computations).


## As a category

This simply means that we have objects (sets), morphisms (matrices), a way of composing them:
If $s: I \rightarrow J$ and $t: J \rightarrow K$ then $t s: I \rightarrow K$ is the product of matrices

$$
(t s)_{i, k}=\sum_{j \in J} s_{i, j} t_{j, k}
$$

And identity matrices $\mathrm{Id}_{I}: I \rightarrow I,\left(\mathrm{Id}_{I}\right)_{i, i^{\prime}}=\delta_{i, i^{\prime}}$.

## Matrices and vectors (distributions)

This seems very stupid, but there are interesting structures behind. . .

The singleton set $\mathbb{1}=\{*\}$.

- A matrix $x: \mathbb{1} \rightarrow I$ is just a subprobability distribution on $I$, $x \in \mathrm{D}(I)$.
- up to trivial iso $x \in \mathrm{D}(I)$ simply means $x \in\left(\mathbb{R}_{\geq 0}\right)^{\prime}$ with $\sum_{i \in I} x_{i} \leq 1$.
- If $s: I \rightarrow J$ then $s x \in \mathrm{D}(J)$ is the image distribution of $x$ :

$$
(s x)_{j}=\sum_{i \in I} s_{i, j} x_{i}
$$

## Codistributions and transpose

What is a matrix $x^{\prime}: I \rightarrow \mathbb{1}$, say $x^{\prime} \in \mathrm{D}^{\prime}(I)$ ?

- It means $x^{\prime} \in\left(\mathbb{R}_{\geq 0}\right)^{\prime}$ with $\forall i \in I x_{i}^{\prime} \leq 1$
- If $s: J \rightarrow I$ then the transpose $s^{\perp} \in\left(\mathbb{R}_{\geq 0}\right)^{J \times I}$ of $s$, defined by $s_{j, i}^{\perp}=s_{i, j}$ satisfies $s^{\perp} x^{\prime} \in \mathrm{D}^{\prime}(J)$
- If $x \in \mathrm{D}(I)$, that is $x: \mathbb{1} \rightarrow I$, then $x^{\prime} x: \mathbb{1} \rightarrow \mathbb{1}$ is just an element of $[0,1]$, notation

$$
\left\langle x, x^{\prime}\right\rangle=\left(x^{\prime} x\right)=\sum_{i \in I} x_{i} x_{i}^{\prime} \in[0,1] \quad(\mathrm{NB}: I \text { can be } \infty)
$$

## Adjunction

## Fact

$$
\left\langle s x, x^{\prime}\right\rangle=\left\langle x, s^{\perp} x^{\prime}\right\rangle=\sum_{i \in I, j \in J} x_{i} s_{i, j} x_{j}^{\prime}
$$

## Duality and linear negation

There is a duality between $\mathrm{D}(I)$ and $\mathrm{D}^{\prime}(I)$ similar to the duality between $\ell^{1}$ and $\ell^{\infty}$ in Banach spaces.

## Fact

$$
\begin{aligned}
\mathrm{D}^{\prime}(I) & =\left\{x^{\prime} \in\left(\mathbb{R}_{\geq 0}\right)^{\prime} \mid \forall x \in \mathrm{D}(I)\left\langle x, x^{\prime}\right\rangle \leq 1\right\} \\
\mathrm{D}(I) & =\left\{x \in\left(\mathbb{R}_{\geq 0}\right)^{\prime} \mid \forall x^{\prime} \in \mathrm{D}^{\prime}(I)\left\langle x, x^{\prime}\right\rangle \leq 1\right\}
\end{aligned}
$$

## A space of substochastic matrices

Let $I$ and $J$ be two sets.
Let $\operatorname{Stoc}(I, J)$ be the set of all $s: I \rightarrow J$, so
$\operatorname{Stoc}(I, J) \subseteq\left(\mathbb{R}_{\geq 0}\right)^{I \times J}$.

## Definition

If $u \in\left(\mathbb{R}_{\geq 0}\right)^{\prime}$ and $v \in\left(\mathbb{R}_{\geq 0}\right)^{J}$ let $u \otimes v \in\left(\mathbb{R}_{\geq 0}\right)^{1 \times J}$ be defined by $(u \otimes v)_{i, j}=u_{i} v_{j}$.

Fact
Stoc $(I, J)=$

$$
\left\{s \in\left(\mathbb{R}_{\geq 0}\right)^{\prime \times J} \mid \forall x \in \mathrm{D}(I) y^{\prime} \in \mathrm{D}^{\prime}(J) \quad\left\langle s, x \otimes y^{\prime}\right\rangle \leq 1\right\}
$$

Indeed

$$
\begin{aligned}
s \in \operatorname{Stoc}(I, J) & \Leftrightarrow \forall x \in \mathrm{D}(I) s x \in \mathrm{D}(J) \\
& \Leftrightarrow \forall x \in \mathrm{D}(I) \forall y^{\prime} \in \mathrm{D}^{\prime}(J)\left\langle s x, y^{\prime}\right\rangle \leq 1 \\
& \Leftrightarrow \forall x \in \mathrm{D}(I) \forall y^{\prime} \in \mathrm{D}^{\prime}(J)\left\langle s, x \otimes y^{\prime}\right\rangle \leq 1
\end{aligned}
$$

since

$$
\begin{aligned}
\left\langle s x, y^{\prime}\right\rangle & =\sum_{j \in J}\left(\sum_{i \in I} s_{i, j} x_{i}\right) y_{j}^{\prime} \\
& =\sum_{i \in I, j \in J} s_{i, j} x_{i} y_{j}^{\prime} \\
& =\left\langle s, x \otimes y^{\prime}\right\rangle
\end{aligned}
$$

## A common pattern!

In all these cases we have defined a $\mathcal{P} \subseteq\left(\mathbb{R}_{\geq 0}\right)^{\prime}$ for some set $l$.
This $\mathcal{P}$ is characterized by

$$
\mathcal{P}=\left\{x \in\left(\mathbb{R}_{\geq 0}\right)^{\prime} \mid \forall x^{\prime} \in \mathcal{P}^{\prime}\left\langle x, x^{\prime}\right\rangle \leq 1\right\}=\mathcal{P}^{\prime \perp}
$$

for some $\mathcal{P}^{\prime} \subseteq\left(\mathbb{R}_{\geq 0}\right)^{\prime}$. A predual of $\mathcal{P}$.

## Fact

The existence of such a $\mathcal{P}^{\prime}$ is equivalent to $\mathcal{P}=\mathcal{P}^{\perp \perp}$.
For all $\mathcal{P}, \mathcal{Q} \subseteq\left(\mathbb{R}_{\geq 0}\right)^{\prime}$

- $\mathcal{P} \subseteq \mathcal{Q} \Rightarrow \mathcal{Q}^{\perp} \subseteq \mathcal{P}^{\perp}$
- $\mathcal{P} \subseteq \mathcal{P}^{\perp \perp}$

So $\mathcal{P}^{\perp}=\mathcal{P}^{\perp \perp \perp}$ always holds.

## Probabilistic coherence spaces (PCS)

A PCS is a pair $X=(|X|, \mathrm{PX})$ where

- $|X|$ is a set (the web)
- $\mathrm{PX} \subseteq\left(\mathbb{R}_{\geq 0}\right)^{|X|}$ such that $\mathrm{PX}=\mathrm{P} X^{\perp \perp}$
- we also assume

$$
\forall a \in|X| \quad 0<\sup \left\{x_{a} \mid x \in \mathrm{P} X\right\}<\infty
$$

so that all coeffs remain finite.
$X^{\perp}=\left(|X|, \mathrm{P} X^{\perp}\right)$ is also a PCS.

Of course $(I, D(I))$ and $\left(I, \mathrm{D}^{\prime}(I)\right)$ are PCS simply denoted $\mathrm{D}(I)$ and $\mathrm{D}^{\prime}(I)$. We have $\mathrm{D}^{\prime}(I)=\mathrm{D}(I)^{\perp}$.
$\operatorname{Stoc}(I, J)$ is an instance of a more general construction:

## Definition

If $X$ and $Y$ are PCS, we define a PCS $X \multimap Y$ by
$|X \multimap Y|=|X| \times|Y|$ and

$$
\begin{aligned}
\mathrm{P}(X \multimap Y) & =\left\{s \in\left(\mathbb{R}_{\geq 0}\right)^{|X \multimap Y|} \mid \forall x \in \mathrm{PX} s x \in \mathrm{P} Y\right\} \\
& =\left\{x \otimes y^{\prime} \mid x \in \mathrm{PX} \text { and } y^{\prime} \in \mathrm{P} Y^{\perp}\right\}^{\perp} .
\end{aligned}
$$

Just as in the special case of $\operatorname{Stoc}(I, J)$. By construction, it is a PCS.

So we have $\operatorname{Stoc}(I, J)=(D(I) \multimap D(J))$.

## LL multiplicative constructs

- $\mathbb{1}$ unit object, $\mathrm{P} \mathbb{1}=[0,1]$, and $\mathbb{1}^{\perp}=\mathbb{1}$.
- $X \multimap Y$ is linear implication.
- $X^{\perp}=\left(|X|, \mathrm{P} X^{\perp}\right)$, linear negation, and we have $X^{\perp \perp}=X$ as in classical logic.
- $X \otimes Y=\left(X \multimap Y^{\perp}\right)^{\perp}$, multiplicative conjuction, tensor product, times. Then $|X \otimes Y|=|X| \times|Y|$ and

$$
\mathrm{P}(X \otimes Y)=\{x \otimes y \mid x \in \mathrm{P} X \text { and } y \in \mathrm{P} Y\}^{\perp \perp}
$$

Think of $A \wedge B=\neg(A \rightarrow \neg B)$ in classical logic.

- $X \not \subset Y=X^{\perp} \multimap Y=\left(X^{\perp} \otimes Y^{\perp}\right)^{\perp}$, multiplicative disjunction, cotensor product, par. Think of $A \vee B=\neg A \rightarrow B$.


## A category

We have now also a generalization of substochastic matrices: the elements $s$ of $\mathrm{P}(X \multimap Y)$.

Remember: they are characterized by a simple property. Given $s \in\left(\mathbb{R}_{\geq 0}\right)^{|X| \times|Y|}$, one has

$$
s \in \mathrm{P}(X \multimap Y) \Leftrightarrow \forall x \in \mathrm{P} X s x \in \mathrm{P} Y
$$

so $\operatorname{ld}_{|X|} \in \mathrm{P}(X \multimap X)$ and if $s \in \mathrm{P}(X \multimap Y)$ and $t \in \mathrm{P}(Y \multimap Z)$ then $t s \in \mathrm{P}(X \multimap Z)$. Because $(t s) x=t(s x)$.

## Matrices are linear maps

An element $s$ of $\mathrm{P}(X \multimap Y)$ is a linear morphism from $X$ to $Y$. And it is really linear (and continuous) in the sense that if $(x(i))_{i \in \mathbb{N}}$ are elements of $\mathrm{P} X$ such that $\sum_{i \in \mathbb{N}} x(i) \in \mathrm{P} X$, one has

$$
s\left(\sum_{i \in \mathbb{N}} x(i)\right)=\sum_{i \in \mathbb{N}} s x(i)
$$

and also $s(\lambda x)=\lambda(s x)$ for $\lambda \in[0,1]$.
$X^{\perp} \simeq(X \multimap \mathbb{1})$ so $X^{\perp}$ is the space of linear continuous forms on $X$, exactly like $E^{*}$ (linear dual) in linear algebra. And here we have $X^{\perp \perp} \simeq X$ exactly like $E^{* *} \simeq E$ in finite dimensional vector spaces.

Here this reflexivity holds also in infinite dimension (when $|X|$ is infinite). Very difficulte to achieve with vector spaces.

## Tensor product and multilinear maps

We have them for free: let $s \in \mathrm{P}\left(X_{1} \otimes \cdots \otimes X_{k} \multimap Y\right)$.
Then the map

$$
\begin{aligned}
\widehat{s}: \prod_{i=1}^{k} \mathrm{P} X_{i} & \rightarrow \mathrm{PY} \\
(x(1), \ldots, x(k)) & \mapsto s(x(1) \otimes \cdots \otimes x(k))
\end{aligned}
$$

is $k$-linear, that is, separately linear in each argument.

## A bilinear map

For instance, we can internalize matrix composition as a bilinear map:

$$
\gamma \in \mathrm{P}(((Y \multimap Z) \otimes(X \multimap Y)) \multimap(X \multimap Y))
$$

such that

$$
\forall t \in \mathrm{P}(Y \multimap Z), \forall s \in \mathrm{P}(X \multimap Y) \quad \widehat{\gamma}(t, s)=\gamma(t \otimes s)=t s
$$

namely $\gamma_{(b, c),\left(a, b^{\prime}\right)}=\delta_{b, b^{\prime}}$.

## Categories

The right categorical setting for describing the situation is that of a symmetric monoidal category (SMC), here the category Pcoh:

- objects are the PCS $X$
- morphisms from $X$ to $Y(\mathbf{P} \boldsymbol{\operatorname { c o h }}(X, Y))$ are the elements of $\mathrm{P}(X \multimap Y)$, identities and composition as described
- together with $\otimes$ which is a functor $\mathbf{P c o h}^{2} \rightarrow \mathbf{P c o h}$
- and additional structures expressing that $\otimes$ has $\mathbb{1}$ as neutral element, is associative, commutative
- and moreover it is closed, meaning that we have $X \multimap Y$ such that $\mathbf{P c o h}(Z \otimes X, Y) \simeq \mathbf{P} \boldsymbol{\operatorname { c o h }}(Z, X \multimap Y)$
- and $X^{\perp}=(X \multimap \mathbb{1})$ with $X^{\perp \perp} \simeq X$ (*-autonomy).


## Cartesian product

## Warning

$X \otimes Y$ is not the "cartesian product" (or categorical product) of $X_{1}$ and $X_{2}$

- there are no projections $p_{i} \in \mathbf{P} \mathbf{c o h}\left(X_{1} \otimes X_{2}, X_{i}\right)$ such that $p_{i}(x(1) \otimes x(2))=x(i)$ in general.
- and there is no duplication $d \in \mathbf{P} \boldsymbol{\operatorname { c o h }}(X, X \otimes X)$ such that $d x=x \otimes x$.

Take $X_{1}=\mathbb{1}$. Then for each $x \in \mathrm{P} X_{2}$ and $\lambda \in \mathrm{P} X_{1}=[0,1]$ we should have $p_{1}(\lambda \otimes x)=p_{1}(\lambda x)=\lambda$. This contradicts linearity in $x$ (take $x=0)$.

## Projection as marginalization

In some cases, there are projections, for instance, we have a linear morphisme $\theta_{l} \in \mathbf{P} \boldsymbol{\operatorname { c o h }}(\mathrm{D}(I), \mathbb{1})$ given by $\left(\theta_{I}\right)_{i, *}=1$

$$
\begin{aligned}
\theta_{l}: \mathrm{D}(I) & \rightarrow \mathbb{1} \\
x & \mapsto \sum_{i \in I} x_{i}
\end{aligned}
$$

Does not work for $\mathrm{D}^{\prime}(I)$ !

So by functoriality of $\otimes$ we have
$\pi_{2}=\theta_{I} \otimes \mathrm{Id} \in \mathbf{P c o h}(\mathrm{D}(I) \otimes \mathrm{D}(J), \mathrm{D}(J))$.
We have $\mathrm{D}(I) \otimes \mathrm{D}(J)=\mathrm{D}(I \times J)$. Given $z \in \mathrm{D}(I \times J)$, we have

$$
\pi_{2} z=\left(\sum_{i \in 1} z_{i, j}\right)_{j \in J}
$$

the marginal distribution.
The existence of $\theta_{\text {, }}$ is related to a crucial logical structure of $\mathrm{D}(I)$ : positivity.

## Similarity with vector spaces

Again, strong similarity with vector spaces: there is a cartesian product of vector space, the so-called direct product of vector spaces $E \times F$ (which coincides with direct sum $E \oplus F$ ).

## direct product vs. tensor product

But $E \times F$ does not coincide with the tensor product $E \otimes F!A$ linear map $E \times F \rightarrow G$ is not the same thing as a bilinear map $E \times F \rightarrow G$. Also $\operatorname{dim} E \otimes F=\operatorname{dim} E \operatorname{dim} F$ whereas $\operatorname{dim} E \times F=\operatorname{dim} E+\operatorname{dim} F$.

We also have a direct product $X \& Y$ and a direct sum $X \oplus Y$ in PCS, but they do not coincide.

If $\left(X_{i}\right)_{i \in I}$ is a family of PCS we can define $X=\&_{i \in I} X_{i}$ by

- $|X|=\bigcup_{i \in I}\{i\} \times\left|X_{i}\right|$
- and, for $z \in\left(\mathbb{R}_{\geq 0}\right)^{|X|}, z \in \mathrm{P} X$ if for all $i \in I$ one has $\pi_{i} z \in \mathrm{P} X_{i}$ where $\pi_{i} \in\left(\mathbb{R}_{\geq 0}\right)^{\left|X \multimap X_{i}\right|}$, the $i$ th projection is

$$
\left(\pi_{i}\right)_{(j, a), a^{\prime}}=\delta_{i, j} \delta_{a, a^{\prime}} .
$$

- so that $\mathrm{P} X \simeq \prod_{i \in I} \mathrm{P} X_{i}$ by $z \mapsto\left(\pi_{i} z\right)_{i \in I}$ and $(x(i))_{i \in I} \mapsto\langle x(i)\rangle_{i \in I}$ given by $z_{(i, a)}=x(i)_{a}$.
By construction we do have now linear projections $\pi_{i} \in \operatorname{Pcoh}\left(\&_{j \in J} X_{j}, X_{i}\right)$.

We can use duality to define the coproduct:

$$
\underset{i \in I}{\oplus} X_{i}=\left(\underset{i \in I}{\&} X_{i}^{\perp}\right)^{\perp}
$$

then we have

$$
\mathrm{P}\left(\underset{i \in I}{\oplus} X_{i}\right)=\left\{\left\langle\lambda_{i} x(i)\right\rangle_{i \in I} \mid \vec{\lambda} \in \mathrm{D}(I) \text { and } \forall i x(i) \in \mathrm{P} X_{i}\right\} \subseteq \mathrm{P}\left(\underset{i \in I}{ } X_{i}\right)
$$

## Beyond linearity: the exponential

## A polynomial function on matrices

Given $k \in \mathbb{N}$ imagine we want to consider the function

$$
\begin{aligned}
f: \mathrm{P}(X \multimap X) & \rightarrow \mathrm{P}(X \multimap X) \\
t & \mapsto t^{k}=\overbrace{t \cdots t}^{k \times}
\end{aligned}
$$

so that

$$
f(t)_{a, c}=\sum_{\substack{b_{0}, \ldots, b_{k} \in|X| \\ b_{0}=a, b_{k}=c}} t_{b_{0}, b_{1}} \cdots t_{b_{k-2}, b_{k-1}} t_{b_{k-1}, b_{k}} .
$$

This is not a linear function when $k>1: f(\lambda s)=\lambda^{k} f(s)$.

## An analytic function on matrices

Or even the function

$$
\begin{aligned}
g: \mathrm{P}(X \multimap X) & \rightarrow \mathrm{P}(X \multimap X) \\
t & \mapsto e^{-1} \sum_{k=0}^{\infty} \frac{1}{k!} t^{k}
\end{aligned}
$$

If $m \in \mathcal{M}_{\text {fin }}(I)$ (finite multiset) and $u \in\left(\mathbb{R}_{\geq 0}\right)^{\prime}$ we set

$$
u^{m}=\prod_{i \in I} u_{i}^{m(i)}
$$

and $u^{(!)} \in\left(\mathbb{R}_{\geq 0}\right)^{\mathcal{M}_{\mathrm{fin}}(I)}$ is defined by $u_{m}^{(!)}=u^{m}$.
Then we define a PCS ! $X$ by $|!X|=\mathcal{M}_{\mathrm{fin}}(|X|)$ and

$$
\mathrm{P}(!X)=\left\{x^{(!)} \mid x \in \mathrm{P} X\right\}^{\perp \perp} .
$$

## Fact

If $t \in\left(\mathbb{R}_{\geq 0}\right)^{|!X \multimap Y|}$, one has

$$
t \in \mathbf{P} \mathbf{c o h}(!X, Y) \Leftrightarrow \forall x \in \mathrm{P} X t x^{(!)} \in \mathrm{P} Y
$$

The function

$$
\begin{aligned}
\hat{t}: \mathrm{PX} & \rightarrow \mathrm{PY} \\
x & \mapsto t x^{(!)}
\end{aligned}
$$

is an "analytic function", $t$ (the powerseries) is completely determined by this function.

## Examples of analytic functions

Let $k \in \mathbb{N}$. Take $f \in\left(\mathbb{R}_{\geq 0}\right)^{!(X \multimap X) \multimap(X \multimap X)}$ given by

$$
f_{m,(a, c)}=\left\{\begin{array}{lc}
1 & \text { if } \exists b_{0}, \ldots, b_{k} \in|X| b_{0}=a, b_{k}=c \text { and } \\
\quad m=\left[\left(b_{0}, b_{1}\right),\left(b_{1}, b_{2}\right), \ldots,\left(b_{k-1}, b_{k}\right)\right] \\
0 \quad \text { otherwise } .
\end{array}\right.
$$

Then given $s \in \mathrm{P}(X \multimap X)$ we have $\widehat{f}(s)=s^{k}$.

Let $g \in\left(\mathbb{R}_{\geq 0}\right)^{!(X \multimap X) \rightarrow(X \multimap X)}$ given by

$$
g_{m,(a, c)}=\left\{\begin{array}{lc}
\frac{e^{-1}}{k!} & \text { if } \exists b_{0}, \ldots, b_{k} \in|X|, b_{0}=a, b_{k}=c \text { and } \\
\quad m=\left[\left(b_{0}, b_{1}\right),\left(b_{1}, b_{2}\right), \ldots,\left(b_{k-1}, b_{k}\right)\right] \\
0 & \text { otherwise }
\end{array}\right.
$$

Let $s \in \mathrm{P}(X \multimap X)$ and $t=\widehat{g}(s) \in \overline{\mathbb{R}} \geq 0|X \multimap X|$, we have

$$
\forall x \in \mathrm{P} X \quad \widehat{t}(x)=\sum_{k=0}^{\infty} \frac{e^{-1}}{k!} \widehat{s}^{k}(x) \in \mathrm{P} X
$$

because $\forall k \in \mathbb{N} \hat{s}^{k}(x) \in P X$ and $\sum_{k=0}^{\infty} \frac{e^{-1}}{k!}=1$.
Hence $\widehat{g}(s) \in \mathrm{P}(X \multimap X)$.
Since this holds for all $s \in \mathrm{P}(X \multimap X)$, we have $g \in \operatorname{Pcoh}(!(X \multimap X), X \multimap X)$.

## Example: stochastic automata

Let $A$ (alphabet) and $Q$ (states) be sets.
$D(A) \otimes D(Q) \multimap D(Q)$ is the space of stochastic automata.
The space of words is the "least" solution $W$ of
$W=\mathbb{1} \oplus(D(A) \otimes W)$. Then it is easy to see that $W=D\left(A^{<\omega}\right)$.
There is an analytic "iteration" function
$r \in \mathbf{P c o h}(W \otimes!(\mathrm{D}(A) \otimes \mathrm{D}(Q) \multimap \mathrm{D}(Q)), \mathrm{D}(Q) \multimap \mathrm{D}(Q))$.

$$
r_{w, m,(a, c)}= \begin{cases}1 & \text { if } w=\left\langle\alpha_{1}, \ldots, \alpha_{k}\right\rangle \\ & m=\left[\left(\alpha_{1}, b_{0}, b_{1}\right), \ldots,\left(\alpha_{k}, b_{k-1}, b_{k}\right)\right] \\ & b_{0}=a, b_{k}=c \\ 0 & \text { otherwise }\end{cases}
$$

So $r$ defines a function

$$
\begin{aligned}
\widehat{r}: \mathrm{PD}(W) \times \mathrm{P}(\mathrm{D}(A) \otimes \mathrm{D}(Q) \multimap \mathrm{D}(Q)) & \rightarrow \mathrm{P}(\mathrm{D}(Q) \multimap \mathrm{D}(Q)) \\
(z, s) & \mapsto r\left(z \otimes s^{(!)}\right)
\end{aligned}
$$

linear in its first argument but not in the second argument.

## Given

- $z \in \operatorname{PD}(W)$, that is $z$ is a subprobability distribution on words
- $s \in \mathrm{P}(\mathrm{D}(A) \otimes \mathrm{D}(Q) \multimap \mathrm{D}(Q))$ is a stochastic automaton

$$
\widehat{r}(z, s)=\sum_{k \in \mathbb{N}} \sum_{\alpha_{1}, \ldots, \alpha_{k} \in A} z_{\left\langle\alpha_{1}, \ldots, \alpha_{k}\right\rangle} s\left(\alpha_{k}\right) \cdots s\left(\alpha_{1}\right)
$$

where $s(\alpha) \in \mathrm{P}(\mathrm{D}(Q) \multimap \mathrm{D}(Q))$ is given by $s(\alpha)_{q, q^{\prime}}=s_{\alpha, q, q^{\prime}}$ : the transition matrix associated with letter $\alpha$.
If $i, f \in Q$ (initial and finite state), $\widehat{r}(z, s)_{i, f} \in[0,1]$ is the probability that we can reach $f$ starting from $i$.

## Pcoh is a very expressive setting

## Fact

If $s \in \mathrm{P}(!X \multimap X)$ then $\widehat{s}: \mathrm{PX} \rightarrow \mathrm{PX}$ is Scott continuous, that is

- $x \leq y \Rightarrow \widehat{s}(x) \leq \widehat{s}(y)$ (where $x \leq y$ simply means $\left.\forall a \in|X| x_{a} \leq y_{a}\right)$
- and if $(x(n))_{n \in \mathbb{N}}$ is a monotone sequence in PX , we have

$$
\widehat{s}\left(\sup _{n \in \mathbb{N}} x(n)\right)=\sup _{n \in \mathbb{N}} \widehat{s}(x(n)) .
$$

As a consequence $\widehat{s}$ has a least fixed point $\sup _{n \in \mathbb{N}} \widehat{s}^{n}(0) \in P X$.

And better, we have $\mathcal{Y} \in \mathbf{P} \boldsymbol{\operatorname { c o h }}(!(!X \multimap X), X)$ such that

$$
\forall s \in \mathrm{P}(!X \multimap X) \quad \widehat{\mathcal{Y}}(s)=\sup _{n \in \mathbb{N}} \widehat{s}^{n}(0)
$$

So we have general recursion in Pcoh.

## A simple example of fixed point

For instance consider

$$
t \in \operatorname{Pcoh}((\mathbb{1} \oplus \mathbb{1}) \otimes(!\mathbb{1} \multimap \mathbb{1}),!\mathbb{1} \multimap \mathbb{1})
$$

such that, for $x \in \mathrm{P}(\mathbb{1} \oplus \mathbb{1})$, $s \in \mathrm{P}(!\mathbb{1} \multimap \mathbb{1})$, $s^{\prime}=\widehat{t}(x, s) \in \mathrm{P}(!\mathbb{1} \multimap \mathbb{1})$ is characterized by

$$
\widehat{s}^{\prime}(y)=x_{\mathbf{t}} y+x_{\mathbf{f}} \widehat{s}(y)^{2}
$$

For each $x \in \mathrm{P}(\mathbb{1} \oplus \mathbb{1})$, the function $s \mapsto s^{\prime}$ has a least fixed point $s$ which satisfies

$$
\forall y \in[0,1] \quad \widehat{s}(y)=x_{\mathbf{t}} y+x_{\mathbf{f}} \widehat{s}(y)^{2}
$$

We can solve this equation:

$$
\widehat{s}(y)= \begin{cases}x_{\mathbf{t}} y & \text { if } x_{f}=0 \\ \frac{1-\sqrt{1-4 x_{\mathbf{t}} x_{\mathbf{f}} y}}{2 x_{\mathbf{f}}} & \text { otherwise }\end{cases}
$$

This can be written as a power series with $\geq 0$ coefficients in $y$, $x_{t}$ and $x_{f}$.

Using $\mathcal{Y}$ we have defined an element $f \in \operatorname{Pcoh}((\mathbb{1} \oplus \mathbb{1}) \otimes!\mathbb{1}, \mathbb{1})$ such that

$$
\widehat{f}(x, y)=x_{\mathbf{t}} y+x_{\mathbf{f}} \widehat{f}(x, y)^{2}
$$

We can also solve general "recursive systems of type equations", for instance find a unique "minimal" PCS $D$ such that

$$
D=\mathbb{1} \&(!D \multimap D)=\mathbb{1} \&\left(? D^{\perp} \ngtr D\right)
$$

that is, a model of the pure $\lambda$-calculus.

## A simpler example of recursive type

There is a "minimal solution" to the equation

$$
S=\mathbb{1} \&(S \oplus S)
$$

$|S|$ is obtained by iteration from $\emptyset$ of the following operation on sets:

$$
E \mapsto\{(1, *)\} \cup\{(2,(1, a)) \mid a \in E\} \cup\{(2,(2, a)) \mid a \in E\}
$$

so up to renaming

$$
|S|=\{0,1\}^{<\omega}
$$

An antichain is a subset $u^{\prime}$ of $|S|$ such that $\forall a, b \in u^{\prime} a \leq b \Rightarrow a=b$ where $\leq$ is the prefix order.

Then $x \in\left(\mathbb{R}_{\geq 0}\right)^{|S|}$ is in PS iff for any antichain $u^{\prime}$ one has $\sum_{a \in u^{\prime}} x_{a} \leq 1$.
For instance, if $s \in\{0,1\}^{\omega}$ then the $x \in\left(\mathbb{R}_{\geq 0}\right)^{|S|}$ such that

$$
x_{a}= \begin{cases}1 & \text { if } a \text { is a prefix of } s \\ 0 & \text { otherwise }\end{cases}
$$

is in PS.

More generally if $\mu$ is a sub-probability measure wrt. the Borelian $\sigma$-algebra of the Cantor space $\{0,1\}^{\omega}$, we can define $x \in\left(\mathbb{R}_{\geq 0}\right)^{|S|}$ by

$$
x_{a}=\mu\left\{s \in\{0,1\}^{\omega} \mid \text { a prefix of } s\right\}
$$

and then $x \in \mathrm{PS}$. Let us set $x=r(\mu)$.
Idea: antichains $\simeq$ open subsets of the Cantor space.

Let $t \in\left(\mathbb{R}_{\geq 0}\right)^{|S \rightarrow S|}$ be defined by

$$
t_{a, b}= \begin{cases}1 & \text { if } a=b 0 \text { or } a=b 1 \\ 0 & \text { otherwise }\end{cases}
$$

We have $t \in \operatorname{Pcoh}(S, S)$.
Simply because if $u^{\prime}$ is an antichain then $\left\{b 0, b 1 \mid a \in u^{\prime}\right\}$ is again an antichain.

Then for $x \in \mathrm{PS}$, we have $t \cdot x=x$ iff

$$
\forall b \in|S| \quad x_{b}=x_{b 0}+x_{b 1}
$$

which is equivalent to the existence of a subprobability distribution $\mu$ on the Cantor space such that $x=r(\mu)$.

## What is so special about!_, logically?

If we have $s \in \operatorname{Pcoh}(X \otimes X, Y)$, which induces the bilinear function

$$
\begin{aligned}
\widehat{s}: \mathrm{P} X \times \mathrm{P} X & \rightarrow \mathrm{PY} \\
(x(1), x(2)) & \mapsto s(x(1) \otimes x(2))
\end{aligned}
$$

we cannot "diagonalize": the map $f: \mathrm{PX} \rightarrow \mathrm{PY}$ defined by $f(x)=\widehat{s}(x, x)$ is not linear (it is quadratic).
We obtain the "cone" of measures on the Cantor space as the equalizer of $t$ and the identity.

In contrast if $s \in \operatorname{Pcoh}(!X \otimes!X, Y)$, which represents the two-parameter analytic function

$$
\begin{aligned}
\widehat{s}: \mathrm{P} X \times \mathrm{P} X & \rightarrow \mathrm{PY} \\
(x(1), x(2)) & \mapsto s\left(x(1)^{(!)} \otimes x(2)^{(!)}\right)
\end{aligned}
$$

then we can diagonalize: there is a $t \in \mathbf{P c o h}(!X, Y)$ such that

$$
\widehat{t}(x)=\widehat{s}(x, x)
$$

The deep reason is that we have $\mathrm{c}_{X}: \mathbf{P c o h}(!X,!X \otimes!X)$ such that

$$
c_{X} x^{(!)}=x^{(!)} \otimes x^{(!)}
$$

namely $\left(c_{X}\right)_{m,(l, r)}=\delta_{m, l+r}$. Then $t=s c_{X}$. This is Contraction, allows to duplicate data.

Similarly if $y \in P Y$, the constant function

$$
\begin{aligned}
\mathrm{P} X & \rightarrow \mathrm{PY} \\
x & \mapsto y
\end{aligned}
$$

is not linear (unless $y=0$ ). But there is $s \in \mathbf{P c o h}(!X, Y)$ such that $\widehat{s}(x)=s \cdot x^{(!)}=y$.

The deep reason is that we have $\mathrm{w}_{X} \in \mathbf{P} \operatorname{coh}(!X, \mathbb{1})$ such that $w_{X} x^{(!)}=1$. This is Weakening, allows to erase data.
$\left(w_{X}\right)_{m, *}=\delta_{m,[]}$

And now, what is LL?

## A possible answer

A logical formalization of this kind of situation, that is, of an idealized multi-linear algebra with the following features:

- It is non degenerate in the sense that $\otimes$ and its dual 8 are different operations, and similarly for direct product \& and direct sum $\oplus$.
- All objects are reflexive, in the sense that $A^{\perp \perp}=A$.
- There is an exponential !_ allowing to write non-linear proofs/programs.

LL can be split in 3 fragments:

- multiplicative: constants 1 (true), $\perp$ (false), binary connectives $\otimes$ (conjunction) and 8 (disjunction)
- additive: constants $T$ (true), 0 (false), binary connectives \& (conjuction) and $\oplus$ (disjunction)
- exponentials: unary connectives! and ?.

Linear negation is defined by induction

$$
\begin{aligned}
1^{\perp} & =\perp & \perp^{\perp} & =1 \\
(A \otimes B)^{\perp} & =A^{\perp} \wp B^{\perp} & (A 叉 B)^{\perp} & =A^{\perp} \otimes B^{\perp} \\
0^{\perp} & =\top & \top^{\perp} & =0 \\
(A \oplus B)^{\perp} & =A^{\perp} \& B^{\perp} & (A \& B)^{\perp} & =A^{\perp} \oplus B^{\perp} \\
(!A)^{\perp} & =?\left(A^{\perp}\right) & (? A)^{\perp} & =!\left(A^{\perp}\right)
\end{aligned}
$$

so that

$$
A^{\perp \perp}=A
$$

We define $A \multimap B=A^{\perp} \ngtr B$.

## Interpretation of formulas in Pcoh

Then we define in an obvious way $\llbracket A \rrbracket$ as a PCS for each formula A:

- $\llbracket 1 \rrbracket=\llbracket \perp \rrbracket=\mathbb{1}$ as indeed $\mathbb{1}^{\perp}=\mathbb{1}$ in Pcoh
- $\llbracket \top \rrbracket=\llbracket 0 \rrbracket=\pi$ the PCS such that $|\pi|=\emptyset$.
- $\llbracket A \otimes B \rrbracket=\llbracket A \rrbracket \otimes \llbracket B \rrbracket$ etc


## Example

$$
\begin{aligned}
& \llbracket 1 \oplus 1 \rrbracket=\mathbb{1} \oplus \mathbb{1}=\left(\{0,1\},\left\{\left(x_{0}, x_{1}\right) \in\left(\mathbb{R}_{\geq 0}\right)^{2} \mid x_{0}+x_{1} \leq 1\right\}\right) \\
& \llbracket 1 \& 1 \rrbracket=\mathbb{1} \& \mathbb{1}=\left(\{0,1\},\left\{\left(x_{0}, x_{1}\right) \in\left(\mathbb{R}_{\geq 0}\right)^{2} \mid x_{0}, x_{1} \leq 1\right\}\right) \\
& \llbracket(1 \& 1) \oplus(1 \& 1) \rrbracket=\left\{\left(x_{0}, x_{1}, x_{2}, x_{3}\right) \in\left(\mathbb{R}_{\geq 0}\right)^{4}\right. \\
& \\
& \left.\qquad \mid x_{0}+x_{2}, x_{0}+x_{3}, x_{1}+x_{2}, x_{1}+x_{3} \leq 1\right\}
\end{aligned}
$$

The $L L$ sequent calculus is a logical system which allows to prove sequents $\vdash \Gamma$ where $\Gamma$ is a list $\left(A_{1}, \ldots, A_{n}\right)$ of formulas.
Intuitively, the "," is a "meta" 8 connective. As in Gentzen LK, where the "," in the sequent $\vdash F_{1}, \ldots, F_{k}$ stands for a $\vee$.

A proof is a tree whose nodes are labeled by logical rules, written in the format

$$
\frac{\vdash \Gamma_{1} \quad \cdots \quad \vdash \Gamma_{k}}{\vdash \Delta}
$$

If $\pi$ is a proof of $\vdash A_{1}, \ldots, A_{k}$, one defines (by induction on the tree $\pi$ )

$$
\llbracket \pi \rrbracket \in \operatorname{Pcoh}\left(\mathbb{1}, \llbracket A_{1} \rrbracket \mathcal{P} \cdots \mathcal{P} \llbracket A_{k} \rrbracket\right)
$$

or equivalently

$$
\llbracket \pi \rrbracket \in \mathbf{P c o h}\left(\llbracket A_{1}^{\perp} \rrbracket \otimes \cdots \otimes \llbracket A_{i-1}^{\perp} \rrbracket \otimes \llbracket A_{i+1}^{\perp} \rrbracket \otimes \cdots \otimes \llbracket A_{k}^{\perp} \rrbracket, \llbracket A_{i} \rrbracket\right)
$$

## Multiplicative rules

Multiplicative constants:

$$
\overline{\vdash 1} \quad \frac{\vdash \Gamma}{\vdash \Gamma, \perp}
$$

Multiplicative connectives:

$$
\frac{\vdash \Gamma_{1}, A_{1} \quad \vdash \Gamma_{2}, A_{2}}{\vdash \Gamma_{1}, \Gamma_{2}, A_{1} \otimes A_{2}} \quad \frac{\vdash \Gamma, A_{1}, A_{2}}{\vdash \Gamma, A_{1} \not 8 A_{2}}
$$

Juxtaposition of contexts

## Additive rules

Additive constants:

$$
\text { no rule for } 0 \quad \overline{\vdash \Gamma, \top}
$$

Additive connectives:

$$
\frac{\vdash \Gamma, A_{i}}{\vdash \Gamma, A_{1} \oplus A_{2}}
$$

$$
\frac{\vdash \Gamma, A_{1} \quad \vdash \Gamma, A_{2}}{\vdash \Gamma, A_{1} \& A_{2}}
$$

Superposition of contexts

## Example

The "and" function of type

$$
\begin{aligned}
& (1 \oplus 1) \otimes(1 \oplus 1) \multimap 1 \oplus 1=((\perp \& \perp) \mathcal{P}(\perp \& \perp)) \mathcal{P}(1 \oplus 1)
\end{aligned}
$$

Interpreted by $t \in \mathbf{P c o h}((1 \oplus 1) \otimes(1 \oplus 1), 1 \oplus 1)$ such that

$$
\widehat{t}(x, y)=x_{\mathbf{t}} y_{\mathbf{t}} e_{\mathbf{t}}+\left(x_{\mathbf{f}} y_{\mathbf{t}}+x_{\mathbf{t}} y_{\mathbf{f}}+x_{\mathbf{f}} y_{\mathbf{f}}\right) e_{\mathbf{f}}
$$

$$
e_{i} \in\left(\mathbb{R}_{\geq 0}\right)^{\prime} \text { defined by }\left(e_{i}\right)_{j}=\delta_{i, j}
$$

## Exponential rules

Weakening and contraction:

$$
\frac{\vdash \Gamma}{\vdash \Gamma, ? A} \quad \frac{\vdash \Gamma, ? A, ? A}{\vdash \Gamma, ? A}
$$

Dereliction and promotion:

$$
\frac{\vdash \Gamma, A}{\vdash \Gamma, ? A} \quad \frac{\vdash ? A_{1}, \ldots, ? A_{k}, B}{\vdash ? A_{1}, \ldots, ? A_{k},!B}
$$

## The axiom

$$
\overline{\vdash A^{\perp}, A}
$$

There is also an echange rule

$$
\frac{\vdash A_{1}, \ldots, A_{k}}{\vdash A_{f(1)}, \ldots, A_{f(k)}}
$$

where $f:\{1, \ldots, k\} \rightarrow\{1, \ldots, k\}$ is a bijection. We keep its use implicit.

## The cut rule

$$
\frac{\vdash \Gamma, A \quad \vdash A^{\perp}, \Delta}{\vdash \Gamma, \Delta}
$$

## Theorem (Hauptsatz)

Any proof $\pi$ of $\vdash \Gamma$ can be transformed (by rewriting) into a cut-free proof $\pi_{0}$ of $\vdash \Gamma$.

Moreover $\llbracket \pi \rrbracket=\llbracket \pi_{0} \rrbracket$.

We have built a proof $\pi$ (the and function on booleans) of

$$
\vdash \perp \& \perp, \perp \& \perp, 1 \oplus 1
$$

We can "diagonalize" it as follows:

$$
\begin{gathered}
\vdots \pi \\
\frac{\vdash \perp \& \perp, \perp \& \perp, 1 \oplus 1}{\vdash ?(\perp \& \perp), \perp \& \perp, 1 \oplus 1} \operatorname{f?(\perp \& \perp ),?(\perp \& \perp ),1\oplus 1} \\
\frac{\vdash ?(\perp \& \perp), 1 \oplus 1}{} \operatorname{contr}
\end{gathered}
$$

This is a proof $\rho$ and $\llbracket \rho \rrbracket=s \in \mathbf{P c o h}(!(1 \oplus 1), 1 \oplus 1)$ such that

$$
\widehat{s}(x)=\widehat{t}(x, x)=x_{\mathbf{t}}^{2} e_{\mathbf{t}}+\left(2 x_{\mathbf{t}} x_{\mathbf{f}}+x_{\mathbf{f}}^{2}\right) e_{\mathbf{f}}
$$

## A simple use of promotion

This proof $\rho$ represents a non-linear (actually quadratic) function $1 \oplus 1 \rightarrow 1 \oplus 1$.

We should be able to "compose it with itself", this is exactly the purpose of the promotion rule (combined with cut):

$$
\begin{gathered}
\vdots \rho \\
\frac{\vdash ?(\perp \& \perp), 1 \oplus 1}{\vdash ?(\perp \& \perp),!(1 \oplus 1)} \text { prom } \quad \begin{array}{c} 
\\
\vdash ?(\perp \& \perp), 1 \oplus 1 \\
\vdash ?(\perp \& \perp), 1 \oplus 1
\end{array}
\end{gathered}
$$

getting an "homogeneous polynomial of degree 4" on booleans:

$$
x_{\mathbf{t}}^{4} e_{\mathbf{t}}+\left(4 x_{\mathbf{t}}^{3} y_{\mathbf{f}}+6 x_{\mathbf{t}}^{2} y_{\mathbf{f}}^{2}+4 x_{\mathbf{t}} y_{\mathbf{f}}^{3}+y_{\mathbf{f}}^{4}\right) e_{\mathbf{f}}
$$

The Girard translation: representing the CBN $\lambda$-calculus in LL

## Types

Let $\iota$ be a ground type.

$$
\sigma, \tau, \cdots:=\iota \mid \sigma \Rightarrow \tau
$$

We choose a formula $\iota$ of $\operatorname{LL}$ and we define $\sigma^{*}$ as a formula of LL by

$$
(\sigma \Rightarrow \tau)^{*}=\left(!\sigma^{*} \multimap \tau^{*}\right)
$$

## Terms

$$
M, N, \cdots:=x\left|\lambda x^{\sigma} M\right|(M) N
$$

Given a term $M$, a context $\Sigma=\left(x_{1}: \sigma_{1}, \ldots, x_{k}: \sigma_{k}\right)$ and a type $\tau$ such that $\Sigma \vdash M: \tau$, we can define $M_{\Sigma}^{*}$, a proof of

$$
\vdash ?\left(\sigma_{1}{ }^{*}\right)^{\perp}, \ldots, ?\left(\sigma_{k}{ }^{*}\right)^{\perp}, \tau^{*}
$$

The translation is by induction on $M$.
If $M=x_{i}$, so that $\tau=\sigma_{i}, M^{*}$ is

$$
\frac{\frac{{\stackrel{\vdash\left(\sigma_{i}{ }^{*}\right)^{\perp}, \sigma_{i}{ }^{*}}{ }} \mathrm{ax}}{\operatorname{\vdash ?~}\left(\sigma_{i}^{*}\right)^{\perp}, \sigma_{i}{ }^{*}} \mathrm{der}}{\vdash ?\left(\sigma_{1}{ }^{*}\right)^{\perp}, \ldots, ?\left(\sigma_{i}{ }^{*}\right)^{\perp}, \ldots, ?\left(\sigma_{k}{ }^{*}\right)^{\perp}, \sigma_{i}{ }^{*}} \text { weak }
$$

If $M=\lambda x^{\sigma} N$ so that $\tau=\sigma \Rightarrow \varphi$ and hence $\tau^{*}=?\left(\sigma^{*}\right)^{\perp}>\varphi^{*}$, then by inductive hypothesis we have a proof

$$
\begin{gather*}
\vdots M_{\sum, x: \sigma}^{*} \\
\frac{\vdash ?\left(\sigma_{1}^{*}\right)^{\perp}, \ldots, ?\left(\sigma_{k}^{*}\right)^{\perp}, ?\left(\sigma^{*}\right)^{\perp}, \varphi^{*}}{\vdash ?\left(\sigma_{1}^{*}\right)^{\perp}, \ldots, ?\left(\sigma_{k}^{*}\right)^{\perp}, ?\left(\sigma^{*}\right)^{\perp \mathcal{Y}} \varphi^{*}} \tag{2}
\end{gather*}
$$

If $M=(N) P$ with $\Sigma \vdash N: \varphi \Rightarrow \tau$ and $\Sigma \vdash P: \varphi$. Let $\Gamma=\left(?\left(\sigma_{1}{ }^{*}\right)^{\perp}, \ldots, ?\left(\sigma_{k}{ }^{*}\right)^{\perp}\right)$ then $M_{\Sigma}^{*}$ is

$$
\begin{aligned}
& \vdots P_{\Sigma}^{*} \\
& \vdots N_{\Sigma}^{*} \\
& \qquad \frac{\vdash \Gamma, \varphi^{*}}{\vdash \Gamma,!\varphi^{*}} \text { prom } \frac{\vdash\left(\varphi^{*}\right) \tau^{*},\left(\tau^{*}\right)^{\perp}}{\vdash \gamma \tau^{*}} \mathrm{ax} \\
& \\
& \frac{\vdash \Gamma, \tau^{*},!\varphi^{*} \otimes\left(\tau^{*}\right)^{\perp}}{\vdash \Gamma, \tau^{*}} \mathrm{fut}
\end{aligned}
$$

because all formulas of $\Gamma$ are of shape $? A$. It is only for this reason that we can use promotion and contraction.

This translation preserves $\beta$-reduction: if $M \beta M^{\prime}$ then $M_{\Sigma}^{*}$ reduces to $M_{\Sigma}^{\prime *}$ by cut elimination.

The converse is morally true.

## What can we compute in LL?

Nothing more than in the simply typed $\lambda$-calculus. . .
But we can extend LL so as to make it more expressive:

- 2nd order (or more)
- least and greatest fixed points of types
- extension allowing non-terminating "proofs": "untyped" LL à Ia Danos-Regnier, LL with a ground type of integers and general recursion analog to PCF etc.


## Conclusion (provisional)

LL allows to embed functional computations in a more symmetric world, where the input/output or program/environment dichotomy is transformed.

LL polarities are exactly about this dichotomy.

## Polarities

To be continued!

