
Linearity: syntax vs. semantics

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- **Coherence Spaces** has been introduced by Jean-Yves Girard after a fine analysis of stable semantics: stable functions were decomposed in linear functions and exponential. Such a decomposition is patently reflected in linear logic syntax.
- In the context of programming languages, linearity were quickly adopted, at first to eliminate garbage collection and shortly thereafter to handle mutable state.
- Variants, refinements, and improvements on linear type systems have been proposed for many applications, including explicit memory management and control of aliasing, capabilities, tracking state changes for program analysis, tpestates for well-behaving API calls, and session types of a channel use agreement.

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Linearity Classification

1. **Syntactical linearity** claims a linear use of variables in terms.

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3. **Reduction linearity** claims that a reduction cannot duplicate/erase redex-occurrences (apart to consume itself).
4. **Operational linearity** claims that redexes are not duplicated during the operational evaluation.
5. **Denotational linearity** claims that programs correspond to linear functions between domains.

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- The starting point is the least full subcategory of coherence spaces endowed with linear functions as morphisms and including as object **infinite flat domain** and closed under **linear-function** spaces.

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- The starting point is the least full subcategory of coherence spaces endowed with linear functions as morphisms and including as object **infinite flat domain** and closed under **linear-function** spaces.
- We avoided the use of exponential domain constructors, thus it should be clear that the considered linear model is not correct (w.r.t. standard interpretation) for a wide number of languages inspired to linear logic.
- We study **PCF**-like languages able to program only functions of such a purely linear model.

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All stable endofunctions in $\iota \rightarrow \iota$ are **linear**, but the constant-functions (having trace $\{(\emptyset, n) \mid n \in \mathbb{N}\}$).

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Novelties

All stable endofunctions in $\iota \rightarrow \iota$ are **linear**, but the constant-functions (having trace $\{(\emptyset, n) \mid n \in \mathbb{N}\}$).

$$\text{As instance, } \frac{\{(n, \underbrace{n + \dots + n}_n) \mid n \in \mathbb{N}\},$$

$$\frac{\{(n, \underbrace{n * \dots * n}_n) \mid n \in \mathbb{N}\}}$$

and $\{(n, \underbrace{n \dots n}_n) \mid n \in \mathbb{N}\}$ are linear traces.

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and $\{(n, \underbrace{n \dots n}_n) \mid n \in \mathbb{N}\}$ are linear traces.

In order to represent all functions in $\mathbb{N} \multimap \mathbb{N}$, \mathcal{SLPCF}_{tk} does not put syntactical-linear constraints on occurrences of ground variables.

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Note that this liberality on the management of ground variables entails that we may also use a high-order term many times, provided that we apply it always to the same sequence of arguments.

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Take into account,

$$\lambda F^{\sigma_0 \rightarrow \dots \rightarrow \sigma_k \rightarrow \iota} . (\lambda x^\iota . x \text{ Op } x) (FM_1^{\sigma_0} \dots M_k^{\sigma_k})^\iota .$$

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$$\lambda F^{\sigma_0 \rightarrow \dots \rightarrow \sigma_k \rightarrow \iota} . (\lambda x^\iota . x \text{Op } x) (FM_1^{\sigma_0} \dots M_k^{\sigma_k})^\iota .$$

The previous term is expected to be equivalent to

$$\lambda F^{\sigma_0 \rightarrow \dots \rightarrow \sigma_k \rightarrow \iota} . (FM_1^{\sigma_0} \dots M_k^{\sigma_k}) \text{Op } (FM_1^{\sigma_0} \dots M_k^{\sigma_k})$$

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The previous term is expected to be equivalent to

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Now, if $N_i^{\sigma_i}$ is the evaluation of $M_i^{\sigma_i}$ then the previous term is expected to be equivalent to

$$\lambda F^{\sigma_0 \rightarrow \dots \rightarrow \sigma_k \rightarrow \iota} . (FN_1^{\sigma_0} \dots N_k^{\sigma_k}) \text{ Op } (FM_1^{\sigma_0} \dots M_k^{\sigma_k})$$

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All linear functions are **strict**, in case:

- Ground Variables: **call-by-value parameter passing**.
- High-Order Variables: **syntactical linearity**.

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- Ground Variables: **call-by-value parameter passing**.
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In order to add all first-order strict stable-functions to our linear-language, we need **fixpoints**. Unfortunately, the least fixpoint of a linear function is always the **bottom** of the considered domain, because strictness.

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$\mathcal{L}PCF_{tk}$ will contain a special kind of variables: **stable variables**. We don't permit to λ -abstract those variables, they will be used only in order to obtain fixpoints.

Typing the Core of $\mathcal{S}\ell\text{PCF}_{\text{tk}}$

$$\frac{}{\mathbf{x}^\iota \vdash \mathbf{x} : \iota} \text{ (gv)}$$

$$\frac{}{\mathbf{f}^{\sigma \multimap \tau} \vdash \mathbf{f} : \sigma \multimap \tau} \text{ (hv)}$$

$$\frac{}{F^\sigma \vdash F : \sigma} \text{ (sv)}$$

Typing the Core of $\mathcal{S}lPCF_{tk}$

$$\frac{}{\mathbf{x}^\ell \vdash \mathbf{x} : \iota} \text{ (gv)}$$

$$\frac{}{\mathbf{f}^{\sigma \multimap \tau} \vdash \mathbf{f} : \sigma \multimap \tau} \text{ (hv)}$$

$$\frac{}{F^\sigma \vdash F : \sigma} \text{ (sv)}$$

$$\frac{\Gamma \vdash M : \tau}{\Gamma, \mathbf{x}^\ell \vdash M : \tau} \text{ (gw)}$$

$$\frac{\Gamma, \mathbf{x}_1^\ell, \mathbf{x}_2^\ell \vdash M : \tau}{\Gamma, \mathbf{x}^\ell \vdash M[\mathbf{x}/\mathbf{x}_1, \mathbf{x}_2] : \tau} \text{ (gc)}$$

$$\frac{\Gamma, \mathcal{X}_2^{\sigma_2}, \mathcal{X}_1^{\sigma_1}, \Delta \vdash M : \tau}{\Gamma, \mathcal{X}_1^{\sigma_1}, \mathcal{X}_2^{\sigma_2}, \Delta \vdash M : \tau} \text{ (ex)}$$

$$\frac{\Gamma, F_1^\sigma, F_2^\sigma \vdash M : \tau}{\Gamma, F^\sigma \vdash M[F/F_1, F_2] : \tau} \text{ (sc)}$$

$$\frac{\Gamma \vdash M : \tau}{\Gamma, F^\sigma \vdash M : \tau} \text{ (sw)}$$

Typing the Core of $\mathcal{S}lPCF_{tk}$

$$\frac{}{\mathbf{x}^\iota \vdash \mathbf{x} : \iota} \text{ (gv)}$$

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$$\frac{\Gamma, \mathcal{X}_2^{\sigma_2}, \mathcal{X}_1^{\sigma_1}, \Delta \vdash M : \tau}{\Gamma, \mathcal{X}_1^{\sigma_1}, \mathcal{X}_2^{\sigma_2}, \Delta \vdash M : \tau} \text{ (ex)}$$

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$$\frac{\Gamma \vdash M : \tau}{\Gamma, F^\sigma \vdash M : \tau} \text{ (sw)}$$

$$\frac{\Gamma, \mathbf{x}^\sigma \vdash M : \tau}{\Gamma \vdash \lambda \mathbf{x}^\sigma. M : \sigma \multimap \tau} \text{ (\lambda)}$$

$$\frac{\Gamma \cap \Delta = \emptyset \quad \Gamma \vdash M : \sigma \multimap \tau \quad \Delta \vdash N : \sigma}{\Gamma, \Delta \vdash MN : \tau} \text{ (ap)}$$

$$\frac{}{\vdash \underline{0} : \iota} \text{ (z)}$$

$$\frac{}{\vdash \text{succ} : \iota \multimap \iota} \text{ (s)}$$

$$\frac{}{\vdash \text{pred} : \iota \multimap \iota} \text{ (p)}$$

$$\frac{\Gamma \cap \Delta = \emptyset \quad \Gamma \vdash M : \iota \quad \Delta \vdash L : \iota \quad \Delta \vdash R : \iota}{\Gamma, \Delta \vdash \text{lif } M L R : \iota} \text{ (lif)}$$

$$\frac{\Gamma^\iota, \Delta^*, F^\sigma \vdash M : \sigma}{\Gamma^\iota, \Delta^* \vdash \mu F.M : \sigma} \text{ (\mu)}$$

Evaluating the Core of $\mathcal{S}\ell\text{PCF}_{\text{tk}}$

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$$\frac{}{\underline{0} \Downarrow \underline{0}} \quad (0)$$

$$\frac{M \Downarrow \underline{n}}{\text{succ } M \Downarrow \text{succ } \underline{n}} \quad (s)$$

$$\frac{M \Downarrow \text{succ } \underline{n}}{\text{pred } M \Downarrow \underline{n}} \quad (p)$$

$$\frac{M \Downarrow \underline{0} \quad L \Downarrow \underline{m}}{\text{elif } M \ L \ R \Downarrow \underline{m}} \quad (\text{ifl})$$

$$\frac{M \Downarrow \text{succ}(\underline{n}) \quad R \Downarrow \underline{m}}{\text{elif } M \ L \ R \Downarrow \underline{m}} \quad (\text{ifr})$$

$$\frac{N \Downarrow \underline{m} \quad M[\underline{m}/x]P_1 \cdots P_i \Downarrow \underline{v}}{(\lambda x^{\iota}.M)NP_1 \cdots P_i \Downarrow \underline{v}} \quad (\lambda^{\iota})$$

$$\frac{M[N/f]P_1 \cdots P_i \Downarrow \underline{v}}{(\lambda f^{\sigma \rightarrow \sigma^{\tau}}.M)NP_1 \cdots P_i \Downarrow \underline{v}} \quad (\lambda^{\circ})$$

$$\frac{M[\mu F.M/F]P_1 \cdots P_i \Downarrow \underline{v}}{(\mu F.M)P_1 \cdots P_i \Downarrow \underline{v}} \quad (\mu)$$

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If $M^{(\iota \rightarrow \circ \iota) \rightarrow \circ \iota}$ program a linear function, then
 $M(\lambda x^\iota . x) \Downarrow \underline{n}$ implies that there is \underline{k} such that

$$M\left(\lambda x^\iota . \text{if}(x \doteq k) \underline{k} \Omega^\iota\right) \Downarrow \underline{n}$$

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A further operator is **which?** : $((\iota \multimap \circ \iota) \multimap \circ \iota) \multimap \circ \iota$

$$\frac{}{\text{which? } M^{(\iota \multimap \circ \iota) \multimap \circ \iota} \Downarrow \langle \underline{n}, \underline{k} \rangle} (w^{\underline{k}})$$

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A further operator is **which?** : $((\iota \multimap \circ \iota) \multimap \circ \iota) \multimap \circ \iota$

$$\frac{M\left(\lambda x^\iota. \text{if}(\underline{k} \doteq x) \underline{k} \Omega^\iota\right) \Downarrow \underline{n}}{\text{which? } M^{(\iota \multimap \circ \iota) \multimap \circ \iota} \Downarrow \langle \underline{n}, \underline{k} \rangle} \quad (w^{\underline{k}})$$

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$\mathcal{S}\ell\text{PCF}_{tk}$ was introduced in [Gaboardi and Paolini 2007, Paolini and Piccolo 2008] as the syntactical counterpart of Linear Stable Functions among Coherence Spaces.

- Formalization of a **Turing-complete** linear language

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- Language assures **token-definability**

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- **Restricted Full Abstraction** for Linear Stable Functions

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- Introduction of **novel linear operators**

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- Formalization of a **Turing-complete** linear language
- Language assures **token-definability**
- **Restricted Full Abstraction** for Linear Stable Functions
- Introduction of **novel linear operators**
- New knowledge on **higher-type computation**

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$$\left\{ \begin{array}{l} ((0, 0), (1, 0), (0, 1), 0) \\ ((0, 1), (0, 0), (1, 0), 1) \\ ((1, 0), (0, 1), (0, 0), 2) \\ ((1, 1), (1, 1), (1, 1), 3) \end{array} \right\}$$

in $(\iota \multimap \iota) \multimap (\iota \multimap \iota) \multimap (\iota \multimap \iota) \multimap \iota$ is linear,

but it cannot be defined in \mathcal{SLPCF}_{tk} .

$$\left\{ \begin{array}{l} ((0, 0), (1, 0), (0, 1), 0) \\ ((0, 1), (0, 0), (1, 0), 1) \\ ((1, 0), (0, 1), (0, 0), 2) \\ ((1, 1), (1, 1), (1, 1), 3) \end{array} \right\}$$

in $(\iota \multimap \iota) \multimap (\iota \multimap \iota) \multimap (\iota \multimap \iota) \multimap \iota$ is linear,

but it cannot be defined in $\mathcal{S}\ell\text{PCF}_{\text{tk}}$.

We can add $\text{Gor}^2 : (\iota \multimap \iota) \multimap (\iota \multimap \iota) \multimap (\iota \multimap \iota) \multimap \iota$ to $\mathcal{S}\ell\text{PCF}_{\text{tk}}^+$ with the following operational semantics

Extensions of $\mathcal{S}lPCF_{tk}$

$$\left\{ \begin{array}{l} ((0, 0), (1, 0), (0, 1), 0) \\ ((0, 1), (0, 0), (1, 0), 1) \\ ((1, 0), (0, 1), (0, 0), 2) \\ ((1, 1), (1, 1), (1, 1), 3) \end{array} \right\}$$

in $(\iota \multimap \iota) \multimap (\iota \multimap \iota) \multimap (\iota \multimap \iota) \multimap \iota$ is linear,

but it cannot be defined in $\mathcal{S}lPCF_{tk}$.

We can add $Gor^2 : (\iota \multimap \iota) \multimap (\iota \multimap \iota) \multimap (\iota \multimap \iota) \multimap \iota$ to $\mathcal{S}lPCF_{tk}^+$ with the following operational semantics

$$\frac{P_1 \underline{0} \Downarrow \underline{0} \quad P_1 \underline{1} \Downarrow \underline{0} \quad P_2 \underline{0} \Downarrow \underline{1}}{Gor^2 P_0 P_1 P_2 \Downarrow \underline{0}} \quad (Gor_0^2)$$

$$\frac{P_1 \underline{0} \Downarrow \underline{1} \quad P_1 \underline{0} \Downarrow \underline{0} \quad P_2 \underline{1} \Downarrow \underline{0}}{Gor^2 P_0 P_1 P_2 \Downarrow \underline{1}} \quad (Gor_1^2)$$

$$\frac{P_1 \underline{1} \Downarrow \underline{0} \quad P_1 \underline{0} \Downarrow \underline{1} \quad P_2 \underline{0} \Downarrow \underline{0}}{Gor^2 P_0 P_1 P_2 \Downarrow \underline{2}} \quad (Gor_2^2)$$

$$\frac{P_1 \underline{1} \Downarrow \underline{1} \quad P_1 \underline{1} \Downarrow \underline{1} \quad P_2 \underline{1} \Downarrow \underline{1}}{Gor^2 P_0 P_1 P_2 \Downarrow \underline{3}} \quad (Gor_3^2)$$

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Tokens are **Trees of Integers** respecting the type-trees!

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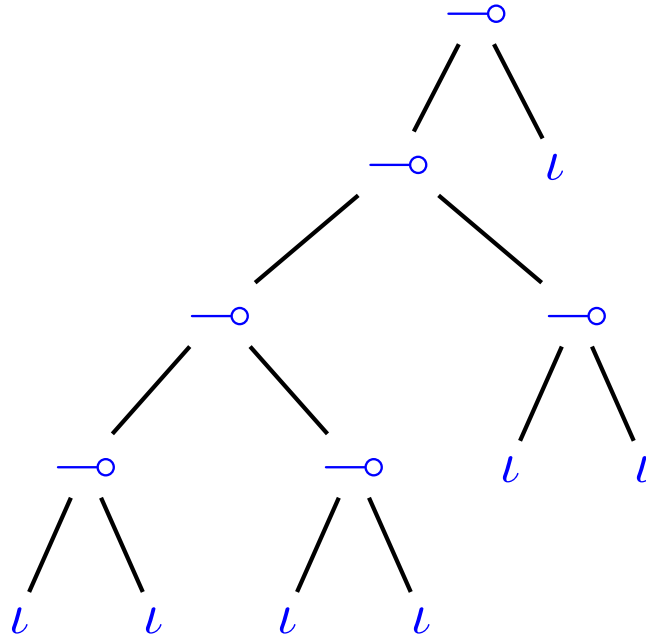
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Tokens are **Trees of Integers** respecting the type-trees!

As instance, the first token of \mathbb{G}^2_{or} is $((0, 0), (1, 0), (0, 1), 0)$

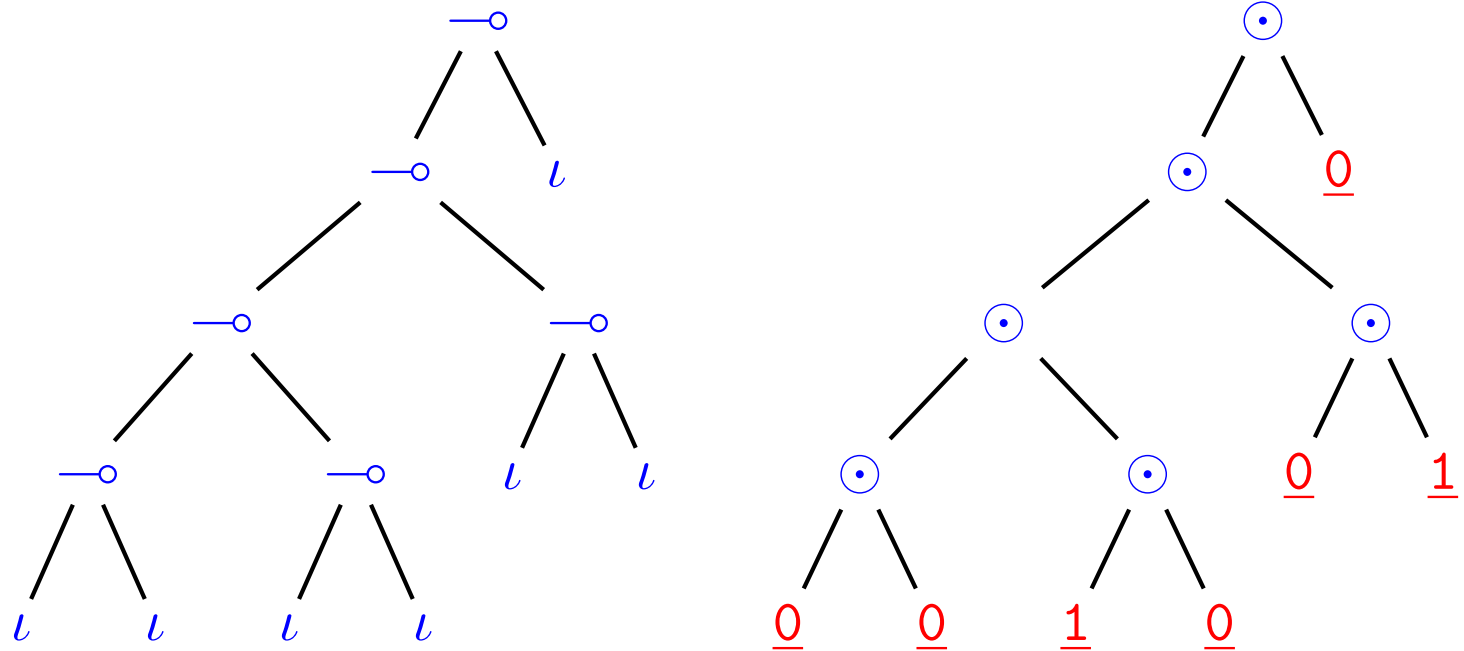


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As instance, the first token of G_{or}^2 is $((0, 0), (1, 0), (0, 1), 0)$



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$\text{Tkn}_n^\sigma : \sigma$ and $\text{Chk}_n^{(\sigma)} : \sigma \multimap \iota$ are defined by mutual induction on σ .

□ case $\sigma = \iota$.

$$\text{Tkn}_n^\iota = \underline{n} \quad \text{Chk}_n^{(\iota)} = \lambda y^\iota. \text{lif } \underline{n} \doteq y \ \underline{0} \ \Omega^\iota$$

□ case $\sigma = \sigma_1 \multimap \sigma_2$ (where $\sigma_2 = \tau_1 \multimap \dots \multimap \tau_k \multimap \iota$).

$$\begin{aligned} \text{Tkn}_n^\sigma &= \lambda f^{\sigma_1}. \lambda g_1^{\tau_1} \dots \lambda g_k^{\tau_k}. \\ &\quad \text{lif } (\text{Chk}_{\pi_1(n)}^{(\sigma_1)} f) (\text{Tkn}_{\pi_2(n)}^{\sigma_2} g_1 \dots g_k) (\Omega^{\sigma_2} g_1 \dots g_k) \end{aligned}$$

$$\text{Chk}_n^{(\sigma)} = \lambda f^\sigma. \text{lif } (\text{Chk}_{\pi_2(n)}^{(\sigma_2)} (f \text{Tkn}_{\pi_1(n)}^{(\sigma_1)})) \ \underline{0} \ \Omega^\iota$$

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Let $a \in \llbracket \sigma_1 \multimap \dots \sigma_m \multimap \iota \rrbracket$ and let $n = \langle (a_1, \dots, a_m, k) \rangle$.

1. $\llbracket \text{Tkn}^{(\sigma)}(\underline{n}) \rrbracket \rho = \{(a_1, \dots, a_m, k)\}$

2. If $N^\sigma \in \mathcal{S}\ell\text{PCF}_{\text{tk}}$ and $\llbracket \text{Chk}_{\underline{n}}^{(\sigma)} N \rrbracket \rho = \llbracket \underline{0} \rrbracket \rho$

then $(a_1, \dots, a_m, k) \in \llbracket N \rrbracket \rho$

Linear Non-Determinism: *let-lor*

$$\frac{\Delta_l = \mathbf{f}_1^{\sigma_1}, \dots, \mathbf{f}_k^{\sigma_k} \quad \Gamma_1 \vdash N_1 : \sigma_1 \dots \Gamma_k \vdash N_k : \sigma_k \quad \Delta \vdash M_i : \iota \quad (i \in \{1, 2, 3\})}{\Gamma_1, \dots, \Gamma_k, \Delta_s, \Delta_l \vdash \underline{\text{let}} \mathbf{f}_1 = N_1, \dots, \mathbf{f}_k = N_k \underline{\text{in}} \text{lor } M_1 M_2 M_3 : \iota} \text{ (let-lor)}$$

Linear Non-Determinism: *let-lor*

$$\frac{\Delta \downarrow_{\ell} = \mathbf{f}_1^{\sigma_1}, \dots, \mathbf{f}_k^{\sigma_k} \quad \Gamma_1 \vdash N_1 : \sigma_1 \quad \dots \quad \Gamma_k \vdash N_k : \sigma_k \quad \Delta \vdash M_i : \iota \quad (i \in \{1, 2, 3\})}{\Gamma_1, \dots, \Gamma_k, \Delta \downarrow_{\mathcal{S}}, \Delta \downarrow_{\ell} \vdash \underline{\text{let}} \mathbf{f}_1 = N_1, \dots, \mathbf{f}_k = N_k \underline{\text{inlor}} M_1 M_2 M_3 : \iota} \quad (\text{let-lor})$$

$$\frac{\begin{array}{l} \exists n_1 \dots n_k \quad M_1[\text{Tkn}_{n_1}^{\sigma_1}/\mathbf{f}_1, \dots, \text{Tkn}_{n_k}^{\sigma_k}/\mathbf{f}_k] \Downarrow \underline{0} \quad (j \in \{1, \dots, k\}) \\ M_2[\text{Tkn}_{n_1}^{\sigma_1}/\mathbf{f}_1, \dots, \text{Tkn}_{n_k}^{\sigma_k}/\mathbf{f}_k] \Downarrow \text{succ } \underline{m} \quad \text{Chk}_{n_j}^{(\sigma_k)} N_j \Downarrow \underline{0} \end{array}}{\underline{\text{let}} \mathbf{f}_1^{\sigma_1} = N_1, \dots, \mathbf{f}_k^{\sigma_k} = N_k \underline{\text{inlor}} M_1 M_2 M_3 \Downarrow \underline{m}} \quad (\text{llgor})$$

Linear Non-Determinism: *let-lor*

$$\frac{\Delta_{\mathcal{L}} = \mathbf{f}_1^{\sigma_1}, \dots, \mathbf{f}_k^{\sigma_k} \quad \Gamma_1 \vdash N_1 : \sigma_1 \quad \dots \quad \Gamma_k \vdash N_k : \sigma_k \quad \Delta \vdash M_i : \iota \quad (i \in \{1, 2, 3\})}{\Gamma_1, \dots, \Gamma_k, \Delta_{\mathcal{S}}, \Delta_{\mathcal{L}} \vdash \underline{\text{let}} \mathbf{f}_1 = N_1, \dots, \mathbf{f}_k = N_k \underline{\text{inlor}} M_1 M_2 M_3 : \iota} \text{ (let-lor)}$$

$$\frac{\begin{array}{l} \exists n_1 \dots n_k \quad M_1[\text{Tkn}_{n_1}^{\sigma_1}/\mathbf{f}_1, \dots, \text{Tkn}_{n_k}^{\sigma_k}/\mathbf{f}_k] \Downarrow \underline{0} \quad (j \in \{1, \dots, k\}) \\ M_2[\text{Tkn}_{n_1}^{\sigma_1}/\mathbf{f}_1, \dots, \text{Tkn}_{n_k}^{\sigma_k}/\mathbf{f}_k] \Downarrow \text{succ } \underline{m} \quad \text{Chk}_{n_j}^{(\sigma_k)} N_j \Downarrow \underline{0} \end{array}}{\underline{\text{let}} \mathbf{f}_1^{\sigma_1} = N_1, \dots, \mathbf{f}_k^{\sigma_k} = N_k \underline{\text{inlor}} M_1 M_2 M_3 \Downarrow \underline{m}} \text{ (1lgor)}$$

$$\frac{\begin{array}{l} \exists n_1 \dots n_k \quad M_2[\text{Tkn}_{n_1}^{\sigma_1}/\mathbf{f}_1, \dots, \text{Tkn}_{n_k}^{\sigma_k}/\mathbf{f}_k] \Downarrow \underline{0} \quad (j \in \{1, \dots, k\}) \\ M_3[\text{Tkn}_{n_1}^{\sigma_1}/\mathbf{f}_1, \dots, \text{Tkn}_{n_k}^{\sigma_k}/\mathbf{f}_k] \Downarrow \text{succ } \underline{m} \quad \text{Chk}_{n_j}^{(\sigma_k)} N_j \Downarrow \underline{0} \end{array}}{\underline{\text{let}} \mathbf{f}_1^{\sigma_1} = N_1, \dots, \mathbf{f}_k^{\sigma_k} = N_k \underline{\text{inlor}} M_1 M_2 M_3 \Downarrow \underline{m}} \text{ (2lgor)}$$

$$\frac{\begin{array}{l} \exists n_1 \dots n_k \quad M_3[\text{Tkn}_{n_1}^{\sigma_1}/\mathbf{f}_1, \dots, \text{Tkn}_{n_k}^{\sigma_k}/\mathbf{f}_k] \Downarrow \underline{0} \quad (j \in \{1, \dots, k\}) \\ M_1[\text{Tkn}_{n_1}^{\sigma_1}/\mathbf{f}_1, \dots, \text{Tkn}_{n_k}^{\sigma_k}/\mathbf{f}_k] \Downarrow \text{succ } \underline{m} \quad \text{Chk}_{n_j}^{(\sigma_k)} N_j \Downarrow \underline{0} \end{array}}{\underline{\text{let}} \mathbf{f}_1^{\sigma_1} = N_1, \dots, \mathbf{f}_k^{\sigma_k} = N_k \underline{\text{inlor}} M_1 M_2 M_3 \Downarrow \underline{m}} \text{ (3lgor)}$$

Pre-orders Coincidence

Let $x_1^\iota, \dots, x_n^\iota, f_1^{\tau_1}, \dots, f_m^{\tau_m}, F_1^{\mu_1}, \dots, F_l^{\mu_l} \vdash M, N : \sigma_1 \multimap \dots \sigma_k \multimap \iota$.

1. $M \lesssim_\sigma N$ whenever, for all $C[\cdot^\sigma]$ s.t. $C[M], C[N] \in \mathcal{P}$,
if $C[M] \Downarrow \underline{n}$ then $C[N] \Downarrow \underline{n}$.

Pre-orders Coincidence

Let $x_1^\iota, \dots, x_n^\iota, f_1^{\tau_1}, \dots, f_m^{\tau_m}, F_1^{\mu_1}, \dots, F_l^{\mu_l} \vdash M, N : \sigma_1 \multimap \dots \sigma_k \multimap \iota$.

1. $M \lesssim_\sigma N$ whenever, for all $C[\cdot^\sigma]$ s.t. $C[M], C[N] \in \mathcal{P}$,
if $C[M] \Downarrow \underline{n}$ then $C[N] \Downarrow \underline{n}$.
2. $M \lesssim_\sigma N$ whenever, for all closed term $P_1^{\sigma_1}, \dots, P_n^{\sigma_n}$, for all $C[\cdot^\sigma]$ s.t.
 $C[M[P_1/F_1, \dots, P_n/F_n]], C[N[P_1/F_1, \dots, P_n/F_n]] \in \mathcal{P}$,
if $C[M[P_1/F_1, \dots, P_n/F_n]] \Downarrow \underline{n}$ then $C[N[P_1/F_1, \dots, P_n/F_n]] \Downarrow \underline{n}$.

Pre-orders Coincidence

Let $x_1^\iota, \dots, x_n^\iota, f_1^{\tau_1}, \dots, f_m^{\tau_m}, F_1^{\mu_1}, \dots, F_l^{\mu_l} \vdash M, N : \sigma_1 \multimap \dots \sigma_k \multimap \iota$.

1. $M \lesssim_\sigma N$ whenever, for all $C[\cdot^\sigma]$ s.t. $C[M], C[N] \in \mathcal{P}$,
if $C[M] \Downarrow \underline{n}$ then $C[N] \Downarrow \underline{n}$.

2. $M \lesssim_\sigma N$ whenever, for all closed term $P_1^{\sigma_1}, \dots, P_n^{\sigma_n}$, for all $C[\cdot^\sigma]$ s.t.
 $C[M[P_1/F_1, \dots, P_n/F_n]], C[N[P_1/F_1, \dots, P_n/F_n]] \in \mathcal{P}$,
if $C[M[P_1/F_1, \dots, P_n/F_n]] \Downarrow \underline{n}$ then $C[N[P_1/F_1, \dots, P_n/F_n]] \Downarrow \underline{n}$.

3. $M \lesssim_\sigma^A N$ whenever,
 for all closed terms $P_1^\iota, \dots, P_n^\iota, Q_1^{\tau_1}, \dots, Q_m^{\tau_m}, R_1^{\mu_1}, \dots, R_l^{\mu_l}, L_1^{\sigma_1}, \dots, L_k^{\sigma_k}$
if $(\lambda \vec{x}. \lambda \vec{f}. M[\vec{R}/\vec{F}]) \vec{P} \vec{Q} \vec{L} \Downarrow \underline{p}$ then $(\lambda \vec{x}. \lambda \vec{f}. N[\vec{R}/\vec{F}]) \vec{P} \vec{Q} \vec{L} \Downarrow \underline{p}$.

Constructive Semantics

$$\frac{}{\langle \underline{0} | e \rangle \Rightarrow \langle \underline{0} | e \rangle} \quad (0) \quad \frac{\langle M | e_0 \rangle \Rightarrow \langle \underline{n} | e_1 \rangle}{\langle \text{succ } M | e_0 \rangle \Rightarrow \langle \text{succ } \underline{n} | e_1 \rangle} \quad (s) \quad \frac{\langle M | e_0 \rangle \Rightarrow \langle \text{succ } \underline{n} | e_1 \rangle}{\langle \text{pred } M | e_0 \rangle \Rightarrow \langle \underline{n} | e_1 \rangle} \quad (p)$$

$$\frac{\langle M | e_0 \rangle \Rightarrow \langle \underline{0} | e_1 \rangle \quad \langle L | e_1 \rangle \Rightarrow \langle \underline{m} | e_2 \rangle}{\langle \text{lif } M \text{ L } R | e_0 \rangle \Rightarrow \langle \underline{m} | e_2 \rangle} \quad (\text{ifl}) \quad \frac{\langle M | e_0 \rangle \Rightarrow \langle \text{succ}(\underline{n}) | e_1 \rangle \quad \langle R | e_1 \rangle \Rightarrow \langle \underline{m} | e_2 \rangle}{\langle \text{lif } M \text{ L } R | e_0 \rangle \Rightarrow \langle \underline{m} | e_2 \rangle} \quad (\text{ifr})$$

$$\frac{\langle M[\mathbf{h}/\mathbf{f}]P_1 \dots P_k | e_0[\mathbf{h} := \mathbf{N}] \rangle \Rightarrow \langle \underline{n} | e_1 \rangle}{\langle (\lambda \mathbf{f}^{\sigma \rightarrow \tau} . M)NP_1 \dots P_k | e_0 \rangle \Rightarrow \langle \underline{n} | e_1 \downarrow_{\{\mathbf{h}\}} \rangle} \quad (\lambda^\circ)$$

$$\frac{\langle \mathbf{N} | e_0 \rangle \Rightarrow \langle \underline{m} | e_1 \rangle \quad \langle M[\mathbf{z}/\mathbf{x}]P_1 \dots P_k | e_1[\mathbf{z} := \underline{m}] \rangle \Rightarrow \langle \underline{n} | e_2 \rangle}{\langle (\lambda \mathbf{x}^\iota . M)NP_1 \dots P_k | e_0 \rangle \Rightarrow \langle \underline{n} | e_2 \downarrow_{\{\mathbf{z}\}} \rangle} \quad (\lambda^\iota)$$

Constructive Semantics

$$\frac{\langle \mathbf{g}P_2 \dots P_k | e_1[\mathbf{h} := P_1, \mathbf{g} := M[\mathbf{h}/\mathbf{x}]] - \mathbf{f} \rangle \Rightarrow \langle \underline{\mathbf{n}} | e_2 \rangle}{\langle \mathbf{f}P_1 \dots P_k | e_0[\mathbf{f} := \lambda \mathbf{x}^{\sigma \rightarrow \sigma} . M] \rangle \Rightarrow \langle \underline{\mathbf{n}} | e_2[\mathbf{f} = e_2(\mathbf{h}) \odot e_2(\mathbf{g})] \upharpoonright_{\{\mathbf{h}, \mathbf{g}\}} \rangle} \text{ (abs}^{-\circ}\text{)}$$

$$\frac{\langle \mathbf{h}NP_1 \dots P_k | e_0[\mathbf{h} := M] - \mathbf{f} \rangle \Rightarrow \langle \underline{\mathbf{n}} | e_1 \rangle}{\langle \mathbf{f}P_1 \dots P_k | e_0[\mathbf{f} := MN] \rangle \Rightarrow \langle \underline{\mathbf{n}} | e_1[\mathbf{f} := \pi_2(e_1(\mathbf{h}))] \upharpoonright_{\{\mathbf{h}\}} \rangle} \text{ (app)}$$

$$\frac{\langle \mathbf{h}P_1 \dots P_k | e_0 - \mathbf{f} \rangle \Rightarrow \langle \underline{\mathbf{n}} | e_1 \rangle}{\langle \mathbf{f}P_1 \dots P_k | e_0[\mathbf{f} := \mathbf{h}] \rangle \Rightarrow \langle \underline{\mathbf{n}} | e_1[\mathbf{f} := \mathbf{h}] \rangle} \text{ (var)}$$

$$\frac{\langle \mathbf{h}I | e_0[\mathbf{h} := M] \rangle \Rightarrow \langle \underline{\mathbf{n}} | e_1 \rangle}{\langle \text{which?}M^{(\iota \rightarrow \iota) \rightarrow \iota} | e_0 \rangle \Rightarrow \langle \underline{\mathbf{n}} \odot \pi_1(e_1(\mathbf{h})) | e_1 \rangle} \text{ (w)}$$

Constructive Semantics

$$\frac{\langle P_1 | e_0 - f \rangle \Rightarrow \langle \underline{m} | e_1 \rangle \quad \langle f P_2 \dots P_k | e_1 [z := \underline{m}, g := M[z/x]] \rangle \Rightarrow \langle \underline{n} | e_2 \rangle}{\langle f P_1 \dots P_k | e_0 [f := \lambda x^t . M] \rangle \Rightarrow \langle \underline{n} | e_2 [f = e_2(\underline{h}) \odot e_2(\underline{g})] \upharpoonright_{\{z,g\}} \rangle} \text{ (abs}^t\text{)}$$

$$\frac{\langle M[\mu F . M / F] P_1 \dots P_k | e_0 \rangle \Rightarrow \langle \underline{n} | e_1 \rangle}{\langle (\mu F . M) P_1 \dots P_k | e_0 \rangle \Rightarrow \langle \underline{n} | e_1 \rangle} \text{ } (\mu) \quad \frac{}{\langle x^t | e[x := \underline{n}] \rangle \Rightarrow \langle \underline{n} | e[x := \underline{n}] \rangle} \text{ (gvar)}$$

$$\frac{\langle M_1 | e_0 \rangle \Rightarrow \langle \underline{n}_1 | e_1 \rangle \quad \langle M_2 | e_1 \rangle \Rightarrow \langle \underline{n}_2 | e_2 \rangle}{\langle M_1 \odot M_2 | e_0 \rangle \Rightarrow \langle \underline{n}_1 \odot \underline{n}_2 | e_2 \rangle} \text{ } (\odot)$$

$$\frac{\langle M | e_0 \rangle \Rightarrow \langle \underline{n}_1 \odot \underline{n}_2 | e_1 \rangle \quad \langle M[z_1/x_1, z_2/x_2] P_1 \dots P_k | e_1 [z_1 := \underline{n}_1, z_2 := \underline{n}_2] \rangle \Rightarrow \langle \underline{n} | e_2 \rangle}{\langle (\underline{\text{let}} \ x_1, x_2 = M \ \underline{\text{in}} \ N) P_1 \dots P_n | e_0 \rangle \Rightarrow \langle \underline{n} | e_2 \rangle} \text{ (let)}$$

Constructive Semantics

$$\begin{array}{l}
 \langle M_1 | e_0 [f_1 := N_1, \dots, f_k := N_k] \rangle \Rightarrow \langle \underline{0} | e_1 \rangle \\
 \langle M_2 | e_0 [f_1 := N_1, \dots, f_k := N_k] \rangle \Rightarrow \langle \underline{m+1} | e_2 \rangle \\
 \hline
 \langle \underline{\text{let}} f_1^{\sigma_1} = N_1, \dots, f_k^{\sigma_k} = N_k \underline{\text{inlor}} M_1 M_2 M_3 | e_0 \rangle \Rightarrow \langle \underline{m} | e_1 \setminus \{f_1, \dots, f_k\} \rangle
 \end{array} \quad \forall i \leq k, e_1(f_i) = e_2(f_i) \quad (1\text{lgor})$$

$$\begin{array}{l}
 \langle M_2 | e_0 [f_1 := N_1, \dots, f_k := N_k] \rangle \Rightarrow \langle \underline{0} | e_2 \rangle \\
 \langle M_3 | e_0 [f_1 := N_1, \dots, f_k := N_k] \rangle \Rightarrow \langle \underline{m+1} | e_3 \rangle \\
 \hline
 \langle \underline{\text{let}} f_1^{\sigma_1} = N_1, \dots, f_k^{\sigma_k} = N_k \underline{\text{inlor}} M_1 M_2 M_3 | e_0 \rangle \Rightarrow \langle \underline{m} | e_2 \setminus \{f_1, \dots, f_k\} \rangle
 \end{array} \quad \forall i \leq k, e_2(f_i) = e_3(f_i) \quad (2\text{lgor})$$

$$\begin{array}{l}
 \langle M_3 | e_0 [f_1 := N_1, \dots, f_k := N_k] \rangle \Rightarrow \langle \underline{0} | e_3 \rangle \\
 \langle M_1 | e_0 [f_1 := N_1, \dots, f_k := N_k] \rangle \Rightarrow \langle \underline{m+1} | e_1 \rangle \\
 \hline
 \langle \underline{\text{let}} f_1^{\sigma_1} = N_1, \dots, f_k^{\sigma_k} = N_k \underline{\text{inlor}} M_1 M_2 M_3 | e_0 \rangle \Rightarrow \langle \underline{m} | e_3 \setminus \{f_1, \dots, f_k\} \rangle
 \end{array} \quad \forall i \leq k, e_3(f_i) = e_1(f_i) \quad (3\text{lgor})$$

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