

Adaptivity analysis

Expr.	e	$::=$	$x \mid e_1 \ e_2 \mid \lambda x.e$ $c \mid \delta(e)$
Value	v	$::=$	$c \mid \lambda x.e$
Adaptivity	R	$::=$	n
Environment	θ	$::=$	$x_1 \mapsto (v_1, R_1), \dots, x_n \mapsto (v_n, R_n)$
Linear type	A	$::=$	$\tau \multimap \tau \mid \mathbf{b}$
Nonlinear Type	τ	$::=$	$!_I A$

$$\begin{array}{c}
\frac{\theta(x) = (v, R)}{\theta, x \Downarrow^R v, \theta} \text{ var} \qquad \frac{}{\theta, c \Downarrow^0 c, \theta} \text{ const} \qquad \frac{}{\theta, \lambda x.e \Downarrow^0 \lambda x.e, \theta} \text{ lambda} \\
\\
\frac{\theta, e_1 \Downarrow^{R_1} \lambda x.e, \theta_1 \quad \theta, e_2 \Downarrow^{R_2} v_2, \theta_2 \quad (\theta_1 \uplus \theta_2)[x \rightarrow (v_2, R_2)], e \Downarrow^{R_3} v, \theta_3}{\theta, e_1 \ e_2 \Downarrow^{R_1+R_3} v, \theta_3} \text{ app} \\
\\
\frac{\theta, e \Downarrow^R v', \theta_1 \quad \delta(v') = v}{\theta, \delta(e) \Downarrow^{R+1} v, \theta_1} \text{ delta} \\
\\
\begin{array}{lll}
\theta_1 \uplus \emptyset & \triangleq & \theta_1 \\
\emptyset \uplus \theta_2 & \triangleq & \theta_2 \\
(\theta_1, [x \rightarrow (v, R_1)]) \uplus (\theta_2, [x \rightarrow (v, R_2)]) & \triangleq & (\theta_1 \uplus \theta_2), [x \rightarrow (v, \max(R_1, R_2))] \\
\text{adap}(e, \emptyset) & ::= & 0 \\
\text{adap}(e, [x \rightarrow (v, R)] \uplus \theta) & ::= & \max(R, \text{adap}(e[v/x], \theta)) \quad x \in \text{FV}(e). \\
& ::= & \text{adap}(e, \theta) \quad x \notin \text{FV}(e)
\end{array}
\end{array}$$

Figure 1: Big-step semantics

$$\begin{array}{c}
\frac{}{\Gamma, x : !_Z A, \Gamma' \vdash_Z x : !_Z A} \mathbf{Ax} \qquad \frac{}{\Gamma \vdash_Z c : !_Z \mathbf{b}} \mathbf{b} \\[10pt]
\frac{\Gamma, x : \tau_1 \vdash_Z e : \tau_2}{k + \Gamma \vdash_{k+Z} \lambda x. e : !_k(\tau_1 \multimap \tau_2)} \mathbf{lambda} \\[10pt]
\frac{\Gamma_1 \vdash_{Z_1} e_1 : !_0(\tau_1 \multimap \tau_2) \quad \Gamma_2 \vdash_{Z_2} e_2 : \tau_1}{\max(\Gamma_1, \Gamma_2) \vdash_{\max(Z_1, Z_2)} e_1 e_2 : \tau_2} \mathbf{app} \\[10pt]
\frac{\Gamma \vdash_Z e : !_k A}{\Gamma', 1 + \Gamma \vdash_{1+Z} \delta(e) : !_k A} \mathbf{delta} \\[10pt]
\frac{\Gamma' \vdash_{Z'} e : \tau' \quad \Gamma' \leq \Gamma \quad Z' \leq Z \quad \tau' <: \tau \quad \Gamma \vdash_Z e : !_k A}{\Gamma \vdash_Z e : \tau} \mathbf{subtype} \\[10pt]
\frac{\Gamma, y : \tau', x : \tau, \Gamma' \vdash_Z e : \tau}{\Gamma, x : \tau, y : \tau', \Gamma' \vdash_Z e : \tau} \mathbf{exchange} \\[10pt]
\begin{array}{ll}
k + !_r A & \triangleq \quad !_k A \\
k + \emptyset & \triangleq \quad \emptyset \\
k + ([x : \tau], \Gamma) & \triangleq \quad [x : k + \tau], k + \Gamma \\
\max(!_k A, !_k A) & \triangleq \quad !_k A \\
\max(\Gamma, \emptyset) & \triangleq \quad \Gamma \\
\max(\emptyset, \Gamma) & \triangleq \quad \Gamma \\
\max\left([x : \tau], \Gamma, ([x : \tau'], \Delta)\right) & \triangleq \quad [x : \max(\tau, \tau')], \max(\Gamma, \Delta) \\
\Gamma <: \Delta & \triangleq \quad \text{dom}(\Gamma) = \text{dom}(\Delta) \wedge \forall x \in \text{dom}(\Gamma), \Delta(x) <: \Gamma(x)
\end{array}
\end{array}$$

Figure 2: Typing rules, first version

$$\begin{array}{ccc}
\frac{k_1 \leq k \quad A <: A_1}{!_k A <: !_k A_1} \mathbf{bang} & \frac{\tau_1 <: \tau \quad \tau' <: \tau'_1}{\tau \multimap \tau' <: \tau_1 \multimap \tau'_1} \mathbf{arrow} & \frac{}{\mathbf{b} <: \mathbf{b}} \mathbf{base}
\end{array}$$

Figure 3: subtyping

Theorem 1 (Substitution). 1. If $\Gamma, x : \tau' \vdash_Z e : \tau$ and $\vdash_{Z'} v : \tau'$, then $\Gamma \vdash_{\max(Z, Z')} e[v/x] : \tau$.

Proof. By induction on the typing derivation. \square

Lemma 2 (Parameter Decreasing). if $k + \Gamma \vdash_Z v : k + \tau$, then exists Z' so that $\Gamma \vdash_{Z'} v : \tau$ and $Z' \leq Z - k$.

Proof. if v is a constant, then it is trivial, assume $\tau = !_r \mathbf{b}$, choose $Z' = r, k' = k$, from the rule b .

If $v = \lambda x.e$. Assume $\tau = !_r \tau_1 \multimap A_2$, then $k + \tau = !_k \tau_1 \multimap A_2$. From its typing derivation, we know: $\Gamma - r, x : \tau_1, \vdash_{Z-(k+r)} e : \tau_2$ (1). Choose $Z' = Z - r$, we know that $\Gamma \vdash_{Z'} v : !_r \tau_1 \multimap A_2$ from the rule lambda . \square

$$\begin{aligned} \theta \models \Gamma &\triangleq \forall x_i \in \text{dom}(\Gamma). \theta(x_i) = (v_i, R_i) \wedge \vdash_{R_i} v_i : \Gamma(x_i) \\ F(\theta, e) &::= \max(R_i) \\ &\text{where } \forall x_i \in \text{FV}(e). \theta(x_i) = (v_i, R_i). \end{aligned}$$

Theorem 3 (Soundness). If $\Gamma \vdash_Z e : \tau$, $\forall \theta$ that $\theta \models \Gamma$, exists θ' and v so that $\theta, e \Downarrow^R v, \theta'$, then $R + \text{adap}(v, \theta') \leq Z + F(\theta, e)$.

Proof. By Induction on the typing derivation.

$\frac{}{\Gamma, x : !_Z A \vdash_Z x : !_Z A} \mathbf{Ax}$
 Assume $\theta = (\theta_1, [x \rightarrow (v, R)]) \models (\Gamma, x : !_Z A)$ where $\theta_1 \models \Gamma$. We know that $\vdash_R v : !_Z A$. From the evaluation rule var , we know $\theta, x \Downarrow^R v, \theta$. TS: $R + \text{adap}(v, \theta) \leq Z + F(\theta) \implies R + 0 \leq Z + \max(R, F(\theta_1))$. It is trivially true.

$\frac{\Gamma, x : \tau_1 \vdash_Z e : \tau_2}{k + \Gamma \vdash_{k+Z} \lambda x.e : !_k(\tau_1 \multimap \tau_2)} \mathbf{lambda}$
 Choose $\theta \models (k + \Gamma)$ so that $\forall x_i \in (\Gamma). \theta(x_i) = (v_i, R_i) \wedge \vdash_{R_i} v_i : k + \Gamma(x_i)$. By the evaluation rule we know $\theta, \lambda x.e \Downarrow^0 \lambda x.e, \theta$, TS: $0 + \text{adap}(\lambda x.e, \theta) \leq k + Z + F(\theta)$, which is trivially true because $\text{adap}(\lambda x.e, \theta) \leq F(\theta)$.

$\frac{\Gamma_1 \vdash_{Z_1} e_1 : !_0(\tau_1 \multimap \tau_2) \quad \Gamma_2 \vdash_{Z_2} e_2 : \tau_1}{\max(\Gamma_1, \Gamma_2) \vdash_{\max(Z_1, Z_2)} e_1 e_2 : \tau_2} \mathbf{app}$
 Choose $\theta = [x_i \rightarrow (v_i, 0)]$ for all x_i in $\text{dom}(\max(\Gamma_1, \Gamma_2))$ so that $\vdash_{Z_i} v_i : (\max(\Gamma_1, \Gamma_2)(x_i))$. From the definition, we know that $\theta \models \Gamma_1$ and $\theta \models \Gamma_2$. Because e_1 has the arrow type and will be evaluated to a function, assume exists θ_1 so that $\theta, e_1 \Downarrow^{R_1} \lambda x.e, \theta_1$. By induction hypothesis on the first premise, we know that: $R_1 + \text{adap}(\lambda x.e, \theta_1) \leq Z_1 + F(\theta, \Gamma_1)$ (1). Assume exists θ_2 so that e_2 is evaluated to an arbitrary value $v_2 : \theta, e_2 \Downarrow^{R_2} v_2, \theta_2$, by induction hypothesis, we conclude that: $R_2 + \text{adap}(v, \theta_2) \leq Z_2 + F(\theta, \Gamma_2)$ (2).

$$\frac{\theta, e_1 \Downarrow^{R_1} \lambda x.e, \theta_1 \quad \theta, e_2 \Downarrow^{R_2} v_2, \theta_2 \quad (\theta_1 \uplus \theta_2)[x \rightarrow (v_2, R_2)], e \Downarrow^{R_3} v, \theta_3}{\theta, e_1 \ e_2 \Downarrow^{R_1+R_3} v, \theta_3} \text{app}$$

□

Theorem 4 (Subject Reduction). If $\Gamma \vdash_Z e : \tau$, $\forall \theta. \theta \models \Gamma$, exists θ' and v , $\theta, e \Downarrow^R v, \theta'$, then $\Gamma \vdash_{\tau} v : \tau$.

By induction on the typing derivation.