

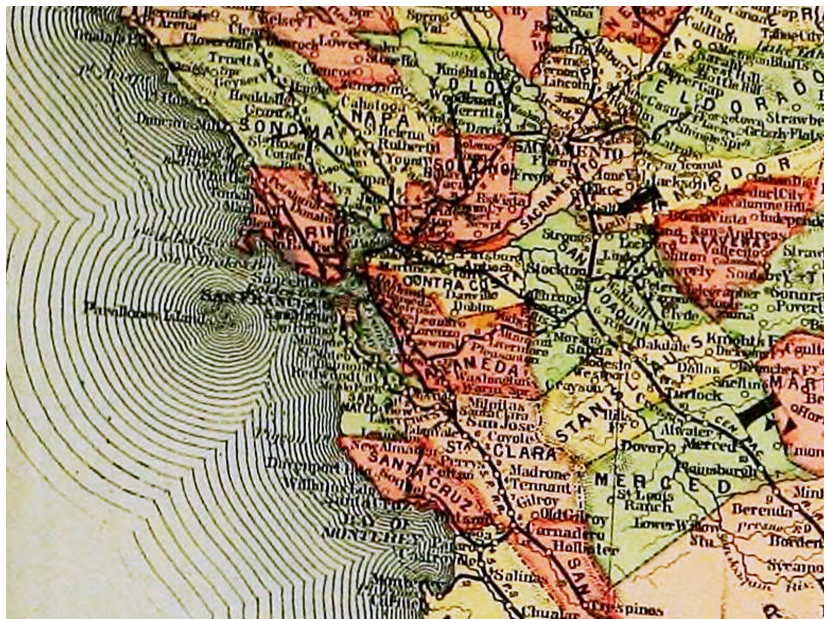
Type-checking Linear Dependent Types

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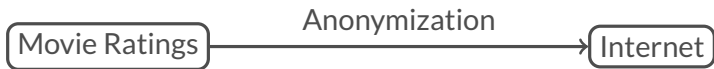
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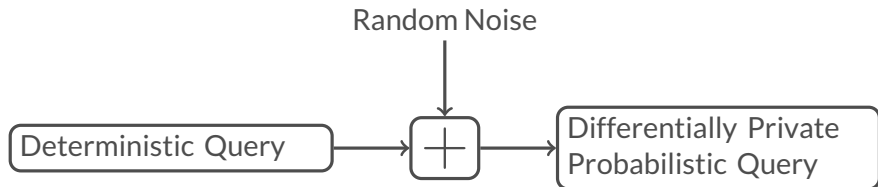


How to allow database queries and retain privacy guarantees?

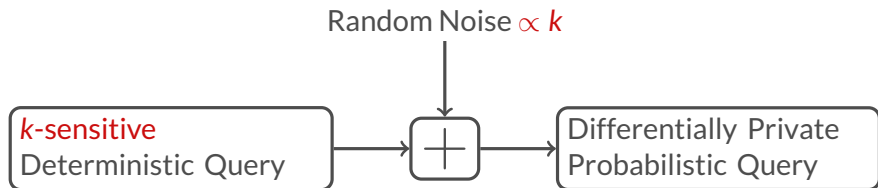
Differential Privacy

- Rigorous bound on “privacy loss” [Dwork, 2006]
- Informally: adding one’s data doesn’t change query results by much
- Many available algorithms
 - Statistical analyses, combinatorial optimizations, machine learning, ...

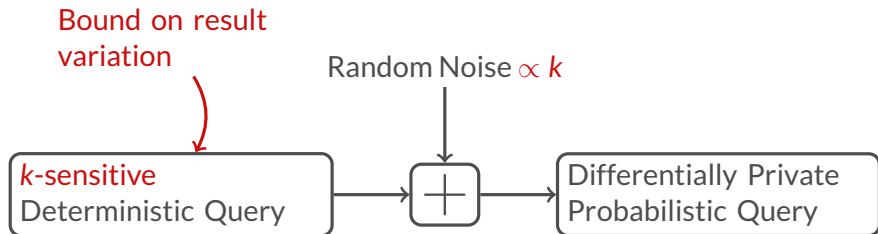
Ensuring Differential Privacy



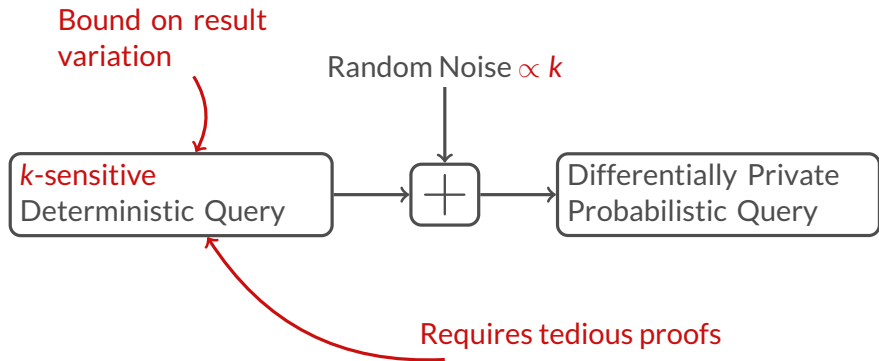
Ensuring Differential Privacy



Ensuring Differential Privacy



Ensuring Differential Privacy



Types to the Rescue

- **DFuzz** [Reed&Pierce10,Gaboardi13] is a type system for function sensitivity (hence, differential privacy)
- Capable of expressing many differentially private algorithms
- Metatheory ensures differential privacy
- Type-checking algorithm: proof automation

Challenge: Checking and Inference

The DFuzz type system combines interesting features:

- Linear indexed types
- Dependent types
- Subtyping

Their interplay makes it difficult to reuse existing techniques directly

Our Contributions

- A **type-checking** and **type-inference** algorithm for a system combining **linear** and **dependent types** in the presence of **subtyping**
- Showing how ideas from the type-checking literature for those domains can be **adapted** to a type system built around a special-purpose **index language**

Outline

- DFuzz and function sensitivity
- Type checking and inference for DFuzz

DFuzz and Function Sensitivity

Function Sensitivity

Bound output variation based on input variation

f is *k-sensitive*: $d(f(x), f(y)) \leq k \cdot d(x, y)$

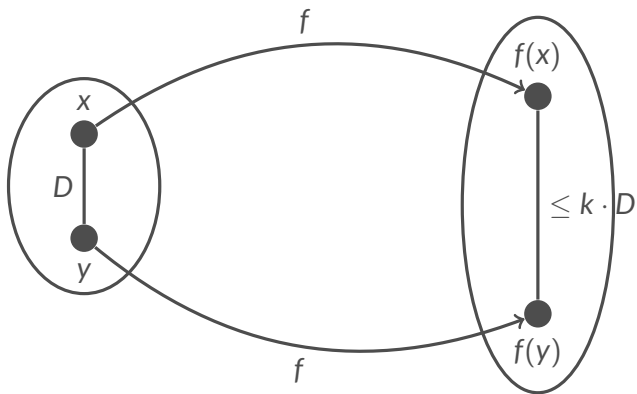
Function Sensitivity

Bound output variation based on input variation

f is k -sensitive: $d(f(x), f(y)) \leq k \cdot d(x, y)$

Distance functions

Function Sensitivity



DFuzz in a Nutshell

- $!_k\sigma \dashv\circ \tau$: k -sensitive function (= linear)

DFuzz in a Nutshell

- $!_k\sigma \rightarrow \pi$: k -sensitive function (= linear)

Multivariate polynomial



DFuzz in a Nutshell

- $!_k \sigma \rightarrow \tau$: k -sensitive function (= linear)
- $list_n \sigma$: list of length n (mechanisms that depend on **input size**)

DFuzz in a Nutshell

- $!_k\sigma \multimap \tau$: k -sensitive function (= linear)
- $list_n \sigma$: list of length n (mechanisms that depend on **input size**)
- $\sigma \sqsubseteq \tau$: sensitivity weakening (e.g. $(!_1\sigma \multimap \tau) \sqsubseteq (!_2\sigma \multimap \tau)$)

A Basic Example

Consider the standard map function

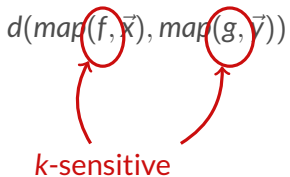
```
function map f l {  
  case l of {  
    | []      => []  
    | x :: l' => f x :: map f l'  
  }  
}
```

How to bound “distance” between results of two calls?

Analyzing Example

$$d(\text{map}(f, \vec{x}), \text{map}(g, \vec{y}))$$

Analyzing Example

$$d(\text{map}(f, \vec{x}), \text{map}(g, \vec{y}))$$


k -sensitive

Analyzing Example

$$d(\text{map}(f, \vec{x}), \text{map}(g, \vec{y}))$$

length n

Analyzing Example

$$d(\text{map}(f, \vec{x}), \text{map}(g, \vec{y}))$$

Distance for lists

$$= \sum_{i=1}^n d(f(x_i), g(y_i))$$

Analyzing Example

$$d(\text{map}(f, \vec{x}), \text{map}(g, \vec{y}))$$

$$= \sum_{i=1}^n d(f(x_i), g(y_i))$$

Triangle inequality

$$\leq \sum_{i=1}^n [d(f(x_i), g(x_i)) + d(g(x_i), g(y_i))]$$

Analyzing Example

$$\begin{aligned} & d(\text{map}(f, \vec{x}), \text{map}(g, \vec{y})) \\ &= \sum_{i=1}^n d(f(x_i), g(y_i)) \\ &\leq \sum_{i=1}^n [d(f(x_i), g(x_i)) + d(g(x_i), g(y_i))] \\ &\leq \sum_{i=1}^n [d(f, g) + k \cdot d(x_i, y_i)] \end{aligned}$$

Max difference
between f and g



Analyzing Example

$$d(\text{map}(f, \vec{x}), \text{map}(g, \vec{y}))$$

$$= \sum_{i=1}^n d(f(x_i), g(y_i))$$

$$\leq \sum_{i=1}^n [d(f(x_i), g(x_i)) + d(g(x_i), g(y_i))]$$

$$\leq \sum_{i=1}^n [d(f, g) + k \cdot d(x_i, y_i)]$$

Definition of
sensitivity

Analyzing Example

$$\begin{aligned} & d(\text{map}(f, \vec{x}), \text{map}(g, \vec{y})) \\ &= \sum_{i=1}^n d(f(x_i), g(y_i)) \\ &\leq \sum_{i=1}^n [d(f(x_i), g(x_i)) + d(g(x_i), g(y_i))] \\ &\leq \sum_{i=1}^n [d(f, g) + k \cdot d(x_i, y_i)] \\ &= n \cdot d(f, g) + k \cdot d(\vec{x}, \vec{y}) \end{aligned}$$

Typing Example in DFuzz

$$\text{map} : !_n(!_k\sigma \multimap \tau) \multimap !_k\text{list}_n \sigma \multimap \text{list}_n \tau$$

Typing Example in DFuzz

$map : !_n (!_k \sigma \multimap \tau) \multimap !_k list_n \sigma \multimap list_n \tau$

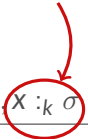
Argument sensitivities

Some Rules

$$\frac{\Gamma, x :_k \sigma \vdash e : \tau}{\Gamma \vdash \lambda x :_k \sigma. e : !_k \sigma \multimap \tau} \quad (\multimap I)$$

Some Rules

Keep track of sensitivity


$$\frac{\Gamma, x :_k \sigma \vdash e : \tau}{\Gamma \vdash \lambda x :_k \sigma. e : !_k \sigma \multimap \tau} \quad (\multimap I)$$

Some Rules

$$\frac{\Gamma, x :_k \sigma \vdash e : \tau}{\Gamma \vdash \lambda x :_k \sigma. e : !_k \sigma \multimap \tau} \quad (\multimap I)$$

Propagate sensitivity to type

Some Rules

$$\frac{\Gamma \vdash e_1 : !_k \sigma \multimap \tau \quad \Delta \vdash e_2 : \sigma}{\Gamma + k \cdot \Delta \vdash e_1 e_2 : \tau} \quad (\multimap E)$$

Context split, combine sensitivities

Some Rules

$$\frac{\Gamma \vdash e_1 : !_k \sigma \multimap \tau \quad \Delta \vdash e_2 : \sigma}{\Gamma \vdash k \cdot \Delta \vdash e_1 e_2 : \tau} \quad (\multimap E)$$

Composition: multiply sensitivities

Some Rules

$$\frac{\begin{array}{c} \Delta \vdash e : list_n \sigma \\ \Gamma \vdash e_{nil} : \tau \\ \Gamma, h :_k \sigma, t :_k list_i \sigma \vdash e_{cons} : \tau \end{array}}{\Gamma + k \cdot \Delta \vdash case\ e\ of\ [] \rightarrow e_{nil} \mid h :: t \rightarrow e_{cons} : \tau} \quad (list\ E)$$

Some Rules

Assuming $n = 0$

$$\frac{\begin{array}{c} \Delta \vdash e : list_n \sigma \\ \Gamma \vdash e_{nil} : \tau \\ \Gamma, h :_k \sigma, t :_k list_j \sigma \vdash e_{cons} : \tau \end{array}}{\Gamma + k \cdot \Delta \vdash case\ e\ of\ [] \rightarrow e_{nil} \mid h :: t \rightarrow e_{cons} : \tau} \quad (list\ E)$$

Some Rules

Assuming $n = i + 1$

$$\frac{\begin{array}{l} \Delta \vdash e : list_n \sigma \\ \Gamma \vdash e_{nil} : \tau \\ \Gamma, h :_k \sigma, t :_k list_i \sigma \vdash e_{cons} : \tau \end{array}}{\Gamma + k \cdot \Delta \vdash case\ e\ of\ [] \rightarrow e_{nil} \mid h :: t \rightarrow e_{cons} : \tau} \quad (list\ E)$$

Some Rules

$$\frac{\begin{array}{c} \Delta \vdash e : \text{list}_n \sigma \\ \Gamma \vdash e_{\text{nil}} : \tau \\ \Gamma, h :_k \sigma, t :_k \text{list}_i \sigma \vdash e_{\text{cons}} : \tau \end{array}}{\Gamma \vdash k \cdot \Delta \vdash \text{case } e \text{ of } [] \rightarrow e_{\text{nil}} \mid h :: t \rightarrow e_{\text{cons}} : \tau} \quad (\text{list E})$$

Track sensitivity on list

Type Checking and Inference

Plan

DFuzz Program

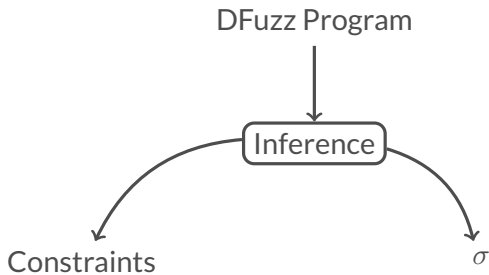
Plan

DFuzz Program

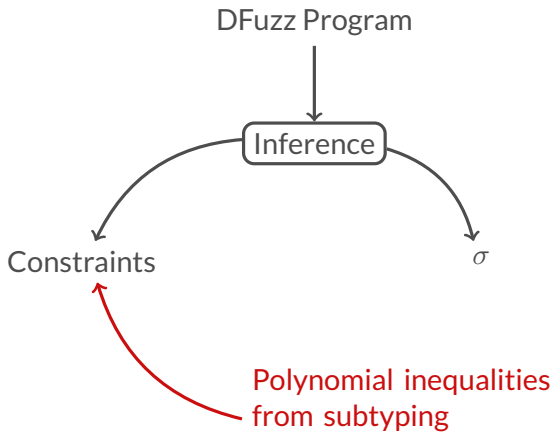


Inference

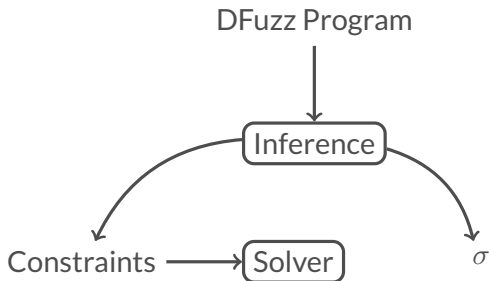
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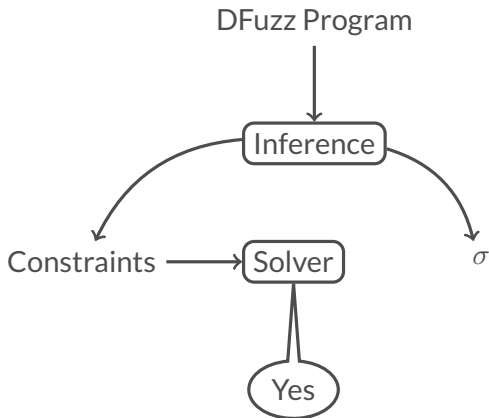
Plan



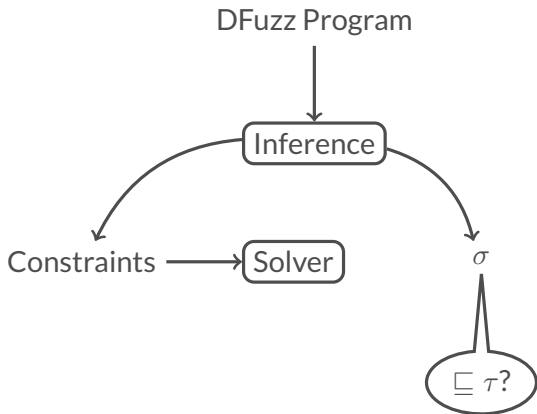
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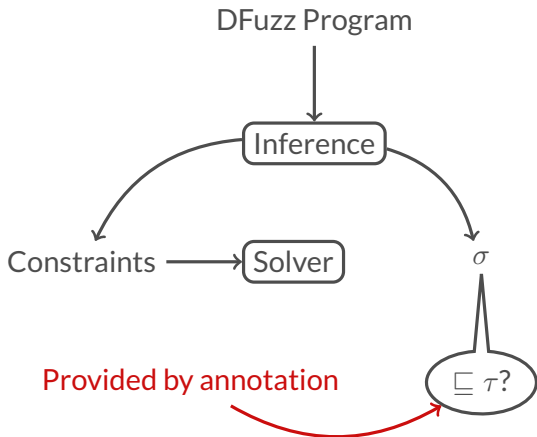
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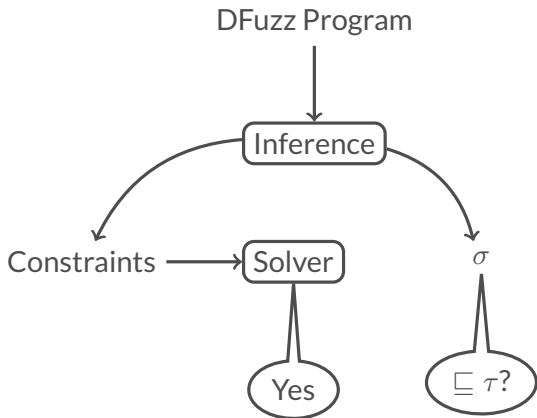
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Plan



Plan



Important Points

- Context splitting imposes a **bottom-up** strategy: start with leaves, combine sensitivities progressively

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... e_1 ... e_2 ...

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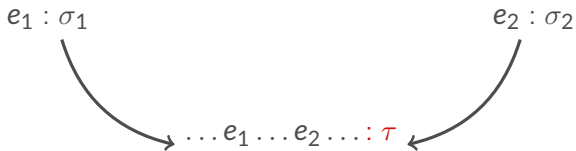
$e_1 : \sigma_1$

$e_2 : \sigma_2$

$\dots e_1 \dots e_2 \dots$

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- Restrict subtyping to essential places (e.g. application)
- Assume sensitivities on higher-order types are given
E.g $!_{\mathbf{k}}(!_{\mathbf{k}}\alpha \multimap \alpha) \multimap \alpha$

Problems

- Language not rich enough to express minimal sensitivities
 - E.g. point-wise maximum of two polynomials is not a polynomial
 - Solution: enrich sensitivity language with new operators

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cf. literature on subtyping

Problems

- Language not rich enough to express minimal sensitivities
 - E.g. point-wise maximum of two polynomials is not a polynomial
 - Solution: enrich sensitivity language with new operators
- Type checking is undecidable
 - Can encode equality of integer polynomials (Hilbert's tenth problem)
 - Completeness relative to a decider of sensitivity inequalities

Syntax-Directed Rules

Equivalent to previous ones, but directly translatable to algorithm

Input Term, argument type annotations

Output Minimal sensitivities, minimal type

Syntax-Directed Rules

$$\frac{\Gamma \vdash e_1 : !_k \sigma \multimap \tau \quad \Delta \vdash e_2 : \sigma' \quad \sigma' \sqsubseteq \sigma}{\Gamma + k \cdot \Delta \vdash e_1 e_2 : \tau} \quad (\multimap E)$$

Syntax-Directed Rules

Not necessarily equal

$$\frac{\Gamma \vdash e_1 : k\sigma \multimap \tau \quad \Delta \vdash e_2 : \sigma' \quad \sigma' \sqsubseteq \sigma}{\Gamma + k \cdot \Delta \vdash e_1 e_2 : \tau} \quad (\multimap E)$$

Syntax-Directed Rules

$$\frac{\Gamma \vdash e_1 : !_k \sigma \multimap \tau \quad \Delta \vdash e_2 : \sigma' \quad \sigma' \sqsubseteq \sigma}{\Gamma + k \cdot \Delta \vdash e_1 e_2 : \tau} \quad (\multimap E)$$

Subtype check

Syntax-Directed Rules

$$\frac{\begin{array}{l} \Delta \vdash e : \text{list}_n \sigma \\ \Gamma \vdash e_{\text{nil}} : \tau_{\text{nil}} \\ \Gamma, h :_k \sigma, t :_k \text{list}_i \sigma \vdash e_{\text{cons}} : \tau_{\text{cons}} \\ \tau = \text{case}(n, \tau_{\text{nil}}, i, \tau_{\text{cons}}) \end{array}}{\Gamma + k \cdot \Delta \vdash \text{case } e \text{ of } [] \rightarrow e_{\text{nil}} \mid h :: t \rightarrow e_{\text{cons}} : \tau} \quad (\text{list E})$$

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Sensitivity-level case lifted to types

Solver Integration

- Need to convert constraints so that standard solvers understand them
- Avoid alternating quantifiers

Solver Integration

$$k \geq \text{case}(n, k_0, i, k_s)$$



$$(n = 0 \Rightarrow k \geq k_0) \wedge (\forall i, n = i + 1 \Rightarrow k \geq k_s)$$

Wrapping Up

Conclusion

- Type-checking system with **linear** and **dependent types**
- Standard ideas adapted to exploit **application domain** and **index structure**
- Recover minimal sensitivities by **extending index language**
 - As simple as possible, no need for much expressive power (cf [DalLago&Petit13])

Implementation

Available at

<http://cis.upenn.edu/~emilioga/dFuzz.tar.gz>

Capable of checking most of the original DFuzz examples

Future Directions

- Let-generalization for sensitivities (remove higher-order annotations)
- Identify decidable fragment of DFuzz

Questions?

Some Metric Spaces

$$d_{\mathbb{R}}(x, y) = |x - y|$$

$$d_{\sigma \rightarrow \tau}(f, g) = \sup_{x \in \sigma} d_{\tau}(f(x), g(x))$$

$$d_{list \sigma}(l_1, l_2) = \begin{cases} \infty & \text{if } \text{length}(l_1) \neq \text{length}(l_2) \\ \sum_i d_{\sigma}(l_1[i], l_2[i]) & \text{otherwise} \end{cases}$$

$$d_{set \sigma}(s_1, s_2) = |s_1 \setminus s_2 \cup s_2 \setminus s_1|$$

$$d_{\mathcal{P}(\sigma)}(\mu, \nu) = \int_{\sigma} \log \left(\frac{d\mu}{d\nu} \right) d\mu$$

More Typing Rules

$$\frac{}{\Gamma, x :_1 \sigma \vdash x : \sigma} \text{ (Var)}$$

More Typing Rules

$$\frac{\Gamma, x :_{\infty} \sigma \vdash e : \sigma}{\infty \cdot \Gamma \vdash \mathit{fix} x : \sigma. e : \sigma} \quad (\mathit{Fix})$$

More Typing Rules

$$\frac{\Delta \sqsubseteq \Gamma \quad \Gamma \vdash e : \sigma \quad \sigma \sqsubseteq \tau}{\Delta \vdash e : \tau} \quad (\sqsubseteq)$$

Metric Preservation

Suppose

$$\vdash e : !_k \sigma \multimap \tau$$

$$\vdash v_1 : \sigma$$

$$\vdash v_2 : \sigma$$

$$e v_1 \rightarrow^* v'_1$$

There exists v'_2 such that $e v_2 \rightarrow^* v'_2$ and

$$d_\tau(v'_1, v'_2) \leq k \cdot d_\sigma(v_1, v_2)$$