

Evolutionary Γ -convergence

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**Analysis of Multi-Scale Systems
Driven by Functionals**

CENTRAL Workshop page: → Materials

Materials for lecture of Alexander Mielke

Survey Article

A. Mielke (2016):

On evolutionary Γ -convergence for gradient systems.

Chapter 3 (pages 187–249) in the Proceedings of Summer School 2012.

Muntean, Rademacher, Zagaris: *Macroscopic and Large Scale Phenomena: Coarse Graining, Mean Field Limits and Ergodicity.*

Lecture Notes in Applied Math. & Mechanics Vol. 3, Springer 2016.

Overview

1. Introduction
2. Gradient systems
3. Motivating examples
4. Energy-dissipation formulations
5. Evolutionary variational inequality (EVI)
6. Rate-independent systems (RIS)

Aim of these lectures:

- Evolutionary systems (time-dependent O/PDEs) with multiple scales
 $0 < \varepsilon = 1/n \ll 1$ small parameter
- Describe mathematical methods for limit passage $\varepsilon \rightarrow 0$
($\varepsilon = h$ contains the case of numerical convergence!)

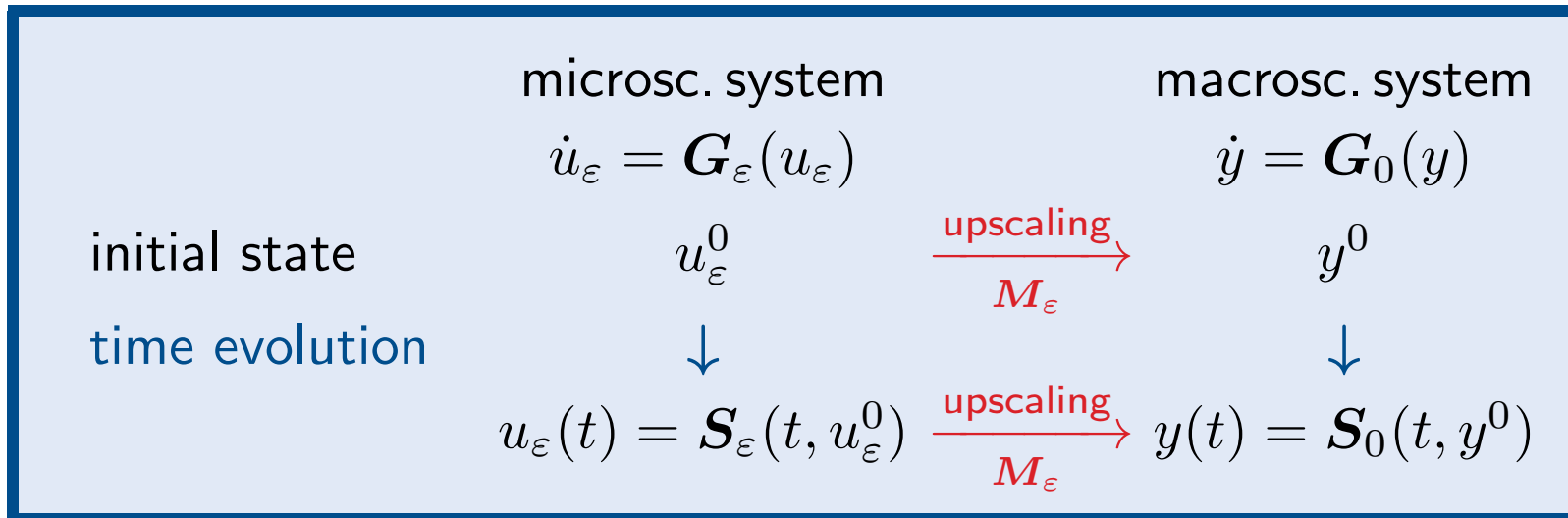
Restriction:

- only generalized gradient systems
- only very simple applications
- proofs only for the simplest results

General evolutionary equations

Multiscale limit corresponds to interchanging to limits, namely

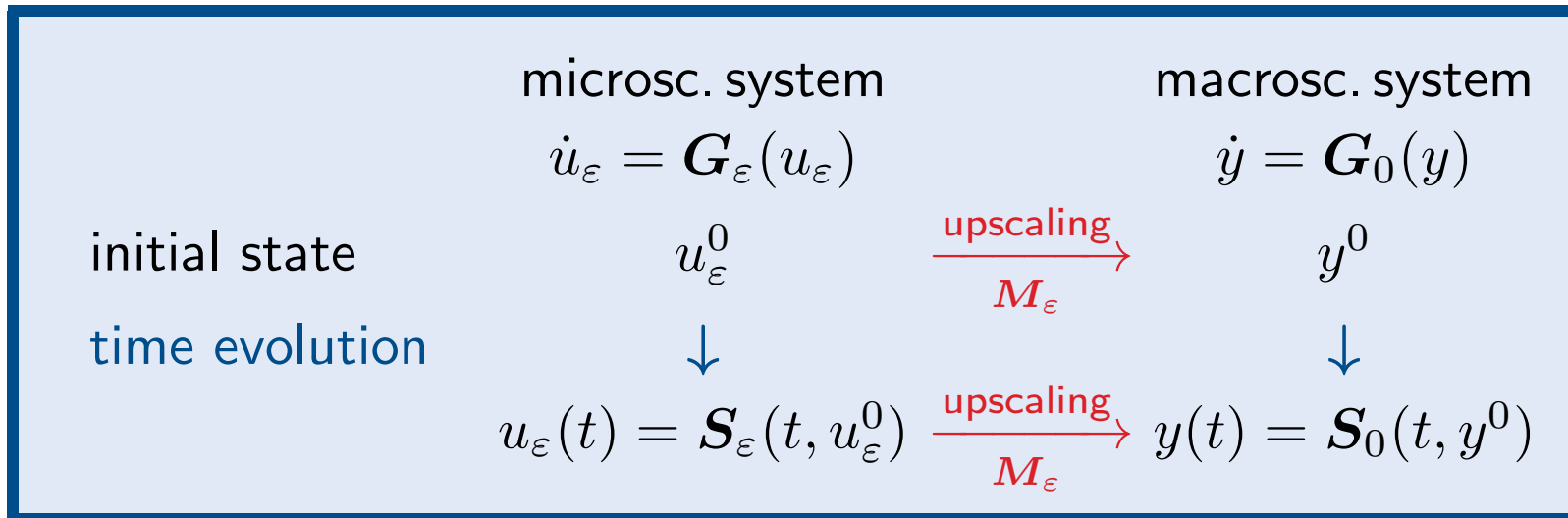
“ $\lim_{\varepsilon \rightarrow 0}$ ” and “ $u^\varepsilon(t) = u_0^\varepsilon + \int_0^t \dots ds$ ”



General evolutionary equations

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“ $\lim_{\varepsilon \rightarrow 0}$ ” and “ $u^\varepsilon(t) = u_0^\varepsilon + \int_0^t \dots ds$ ”



Mathematical task: Prove $\lim_{\varepsilon \rightarrow 0} \mathbf{M}_\varepsilon \circ \mathbf{S}_\varepsilon(t, \cdot) = \mathbf{S}_0(t, \lim_{\varepsilon \rightarrow 0} \mathbf{M}_\varepsilon(\cdot))$

We say that **the PDEs $\dot{u} = \mathbf{G}_\varepsilon(u)$ evolutionary converge to $\dot{u} = \mathbf{G}_0(u)$.**

Γ -convergence is a purely static concept

At first sight there is no relation to evolution.

- $\mathcal{J}_0, \mathcal{J}_\varepsilon : \mathbf{X} \rightarrow \mathbb{R}$ are functionals, \mathbf{X} sep./refl. Banach space

- If $\mathcal{J}_\varepsilon \xrightarrow{\Gamma} \mathcal{J}_0$, then *solutions of minimization problems converge*:

$\forall \ell \in \mathbf{X}^*$ we have

$$\left. \begin{array}{l} u_\varepsilon \in \underset{w \in \mathbf{X}}{\text{ArgMin}} \left(\mathcal{J}_\varepsilon(w) - \langle \ell, w \rangle \right) \\ \text{and } u_\varepsilon \rightarrow u \end{array} \right\} \implies u \in \underset{w \in \mathbf{X}}{\text{ArgMin}} \left(\mathcal{J}_0(w) - \langle \ell, w \rangle \right)$$

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Evolutionary Γ -convergence means “evolutionary convergence” for time-dependent O/PDEs that are given in terms of functionals.

\rightsquigarrow evolution/dynamics driven by functionals

An example for **evolution driven by functionals**

■ the damped wave equation

$$\rho(x)\ddot{u}(t, x) + \delta(x)\dot{u}(t, x) = \operatorname{div}(A(x)\nabla u(t, x)) + f(t, x) \text{ in } \Omega + \text{Dir.B.C.}$$

What are the relevant functionals?

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What are the relevant functionals?

- kinetic energy $\mathcal{K}(\dot{u}) = \int_{\Omega} \frac{\rho}{2} \dot{u}^2 dx$
- potential energy $\mathcal{E}(t, u) = \int_{\Omega} \left(\frac{1}{2} \nabla u \cdot A \nabla u - u f(t) \right) dx$

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- dissipation potential $\mathcal{R}(\dot{u}) = \int_{\Omega} \frac{\delta}{2} \dot{u}^2 dx$

The PDE is given in terms of the three functionals \mathcal{K} , \mathcal{E} , and \mathcal{R} via the force balance (cf. lectures by T. Roubíček or E. Davoli)

$$0 = \underbrace{\frac{\partial}{\partial t} \left(D_{\dot{u}} \mathcal{K}(\dot{u}) \right)}_{\text{inertial terms}} + \underbrace{D_{\dot{u}} \mathcal{R}(\dot{u})}_{\text{dissipation}} + \underbrace{D_u \mathcal{E}(t, u)}_{\text{potential force}}$$

Slightly more general O/PDE driven by functionals

$$(DE)_\varepsilon \quad 0 = \frac{\partial}{\partial t} \left(D_{\dot{u}} \mathcal{K}_\varepsilon(u, \dot{u}) \right) + D_{\dot{u}} \mathcal{R}_\varepsilon(u, \dot{u}) + D_u \mathcal{E}_\varepsilon(t, u)$$

Naïve hope of evolutionary Γ convergence

$$\left. \begin{array}{l} \mathcal{K}_\varepsilon \xrightarrow{\Gamma} \mathcal{K}_0 \\ \mathcal{R}_\varepsilon \xrightarrow{\Gamma} \mathcal{R}_0 \\ \mathcal{E}_\varepsilon \xrightarrow{\Gamma} \mathcal{E}_0 \end{array} \right\} \implies (DE)_\varepsilon \xrightarrow{\text{evol}} (DE)_0$$

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- In general, this is wrong since the convergences need to be “compatible”.
(In numerics: discretizations of different parts need to be compatible.)
- True goal: Find sufficient compatibility conditions for the convergences.

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2.1. General definitions for gradient systems

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2. Gradient systems

Gradient flows = evolution driven by **gradient systems** $(\mathbf{X}, \mathcal{E}, \mathbb{G})$

- $u \in \mathbf{X}$ = state space (closed convex subset of a reflexive Banach space)
- $\mathcal{E} : [0, T] \times \mathbf{X} \rightarrow \mathbb{R}_\infty := \mathbb{R} \cup \{\infty\}$ energy functional
- $\mathbb{G}(u) : T_u \mathbf{X} = \mathbf{X} \rightarrow T_u^* \mathbf{X} = \mathbf{X}^*$ metric structure
(Riemannian tensor with $\mathbb{G}(u) = \mathbb{G}(u)^* \geq 0$)

A gradient system induces a DE via (the force balance)

$$0 = \underbrace{\mathbb{G}(u)\dot{u}}_{\text{visc. force}} + \underbrace{D_u \mathcal{E}(t, u)}_{\text{rest. force}} \in T_u^* \mathbf{X} = \mathbf{X}^*$$

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This gives the equivalent formulation (gradient flow)

$$\dot{u} = -\nabla_{\mathbb{G}} \mathcal{E}(t, u) = -\mathbb{G}(u)^{-1} D_u \mathcal{E}(t, u) \in T_u \mathbf{X} = \mathbf{X}$$

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2. Gradient systems

$$0 = \mathbb{G}(u)\dot{u} + D_u \mathcal{E}(t, u) \in \mathbf{X}^* \iff \dot{u} = -\mathbb{K}(u)D_u \mathcal{E}(t, u) \in \mathbf{X}$$

The metric tensor \mathbb{G} is uniquely characterized by a quadratic form, namely the **(primal) dissipation potential**

$$\mathcal{R}(u, \dot{u}) := \frac{1}{2} \left\langle \underbrace{\mathbb{G}(u)\dot{u}}_{\in \mathbf{X}^*}, \underbrace{\dot{u}}_{\in \mathbf{X}} \right\rangle \implies D_{\dot{u}} \mathcal{R}(u, \dot{u}) = \mathbb{G}(u)\dot{u} \in \mathbf{X}^*$$

We introduce the short-hand $\mathbb{K}(u) := \mathbb{G}(u)^{-1}$ and define the **dual dissipation potential**

$$\mathcal{R}^*(u, \xi) := \frac{1}{2} \left\langle \underbrace{\xi}_{\in \mathbf{X}^*}, \underbrace{\mathbb{K}(u)\xi}_{\in \mathbf{X}} \right\rangle \implies D_{\xi} \mathcal{R}^*(u, \xi) = \mathbb{K}(u)\xi \in \mathbf{X}$$

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$$0 = \mathbb{G}(u)\dot{u} + D_u \mathcal{E}(t, u) \in \mathbf{X}^* \quad \Longleftrightarrow \quad \dot{u} = -\mathbb{K}(u)D_u \mathcal{E}(t, u) \in \mathbf{X}$$

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- Gradient systems can also be denoted by

$$(\mathbf{X}, \mathcal{E}, \mathbb{G}) = (\mathbf{X}, \mathcal{E}, \mathcal{R}) = (\mathbf{X}, \mathcal{E}, \mathcal{R}^*) = (\mathbf{X}, \mathcal{E}, \mathbb{K})$$

- The induced equation can be written as

$$0 = D_{\dot{u}} \mathcal{R}(u, \dot{u}) + D_u \mathcal{E}(t, u) \in \mathbf{X}^* \quad \Longleftrightarrow \quad \dot{u} = D_{\xi} \mathcal{R}^*(u, -D_u \mathcal{E}(t, u)) \in \mathbf{X}$$

2. Gradient systems

Generalized gradient systems $(\mathbf{X}, \mathcal{E}, \mathcal{R})$

$\mathcal{R}(u, \dot{u})$ general dissipation potential, which means that

$\mathcal{R}(u, \cdot): \mathbf{X} \rightarrow [0, \infty]$ is convex, lower semi-continuous, and $\mathcal{R}(u, 0) = 0$.

The possible dissipative forces are given by the (set-valued) convex subdifferential $\partial_{\dot{u}} \mathcal{R}(u, \dot{u}) = \{ \xi \in \mathbf{X}^* \mid \forall w \in \mathbf{X}: \mathcal{R}(u, w) \geq \mathcal{R}(u, \dot{u}) + \langle \xi, w - \dot{u} \rangle \}$.

$$(DE) \quad 0 \in \partial_{\dot{u}} \mathcal{R}(u, \dot{u}) + D_u \mathcal{E}(t, u)$$

Classical gradient system $\mathcal{R}(u, v) = \frac{1}{2} \langle \mathbb{G}(u)v, v \rangle$ (quadratic)

More general $\mathcal{R}(u, v) = \|\mathbb{A}(u)v\|_B + \frac{1}{2} \|\mathbb{V}(u)v\|_H^2 + \frac{1}{p} \|\mathbb{M}(u)\|_Z^p$

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In **multiscale modeling** one is interested in

Γ -convergence for families of gradient systems $(\mathbf{X}, \mathcal{E}_\varepsilon, \mathcal{R}_\varepsilon)$

- homogenization
- dimension reductions (plates, interfaces, ...)
- singular perturbations
- numerical approximation $\varepsilon = h \rightarrow 0$

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Our working definition for this course:

Definition (Γ -convergence of generalized gradient systems
= **evolutionary Γ -convergence**)

We write $(\mathbf{X}, \mathcal{E}_\varepsilon, \mathcal{R}_\varepsilon) \xrightarrow{\text{evol}} (\mathbf{X}, \mathcal{E}_0, \mathcal{R}_0)$ if and only if

$$\left. \begin{array}{l} u^\varepsilon : [0, T] \rightarrow \mathbf{X} \\ \text{solves } (\mathbf{X}, \mathcal{E}_\varepsilon, \mathcal{R}_\varepsilon) \\ u^\varepsilon(0) \rightharpoonup u_0, \\ \mathcal{E}_\varepsilon(u^\varepsilon(0)) \rightarrow \mathcal{E}_0(u_0) \end{array} \right\} \implies \left\{ \begin{array}{l} \exists u \text{ sln. of } (\mathbf{X}, \mathcal{E}_0, \mathcal{R}_0) \text{ with } u(0) = u_0 \\ \text{and a subsequence } \varepsilon_k \rightarrow 0 : \\ \forall t \in [0, T] : u^{\varepsilon_k}(t) \rightharpoonup u(t) \\ \mathcal{E}_\varepsilon(u^{\varepsilon_k}(t)) \rightarrow \mathcal{E}_0(u(t)) \end{array} \right.$$

Aim: Find conditions of $(\mathcal{E}_\varepsilon, \mathcal{R}_\varepsilon) \rightsquigarrow (\mathcal{E}_0, \mathcal{R}_0)$
to guarantee evolutionary Γ -convergence.

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3. Motivating examples

3.1. Possible applications

3.2. Γ -convergence for (static) functionals

3.3. An ODE problem

3.4. Homogenization

4. Energy-dissipation formulations

5. Evolutionary variational inequality (EVI)

6. Rate-independent systems (RIS)

3. Motivating examples

Why do we want to use gradient structures?

- They displays the physics behind:
 - one DE may have several gradient structures
- The gradient structure defines function space
 - energy space $u \in \mathbf{Z} \Leftrightarrow \mathcal{E}(t, u) < \infty$
 - dynamic space $\dot{u} \in \mathbf{X} \Leftrightarrow \mathcal{R}(\dot{u}) < \infty$
- Using the gradient structure may simplify the proof of showing evolutionary convergence

However: Evol. Γ -conv. for $(\mathbf{X}, \mathcal{E}_\varepsilon, \mathbb{G}_\varepsilon) \xRightarrow{\neq} \text{Evol. conv. for } \dot{u} = -\nabla_{\mathbb{G}_\varepsilon} \mathcal{E}_\varepsilon(u)$

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However: Evol. Γ -conv. for $(\mathbf{X}, \mathcal{E}_\varepsilon, \mathbb{G}_\varepsilon) \not\Rightarrow$ Evol. conv. for $\dot{u} = -\nabla_{\mathbb{G}_\varepsilon} \mathcal{E}_\varepsilon(u)$

- **Most importantly:** We will see an example, where one equation has different gradient structures having evolutionary Γ -limits that do not coincide

$$\begin{array}{l}
 \dot{u} = \mathbf{G}_\varepsilon(u) \begin{array}{l} \nearrow \dot{u} = -\mathbb{K}_\varepsilon(u) D\mathcal{E}_\varepsilon(u) \xrightarrow{\text{evol. } \Gamma\text{-conv.}} \dot{u} = -\mathbb{K}_0(u) D\mathcal{E}_0 = \mathbf{G}_0(u) \\ \searrow \dot{u} = -\tilde{\mathbb{K}}_\varepsilon(u) D\tilde{\mathcal{E}}_\varepsilon(u) \xrightarrow{\text{evol. } \Gamma\text{-conv.}} \dot{u} = -\tilde{\mathbb{K}}_0(u) D\tilde{\mathcal{E}}_0 = \tilde{\mathbf{G}}_0(u) \end{array} \quad \text{different!!}
 \end{array}$$

3. Motivating examples

$$\text{Heat equation } \dot{\theta} = \kappa \Delta \theta \quad \neq \quad \text{Diffusion equation } \dot{u} = m \Delta u$$

(This will be important for coupling reaction-diffusion and heat transfer in one thermodynamic framework. \rightsquigarrow Tutorial)

3. Motivating examples

Heat equation for temperature \neq diffusion equation

Diffusion equation $\dot{v} = m\Delta v = -\mathbb{K}_{\text{diff}}(v)D\mathcal{E}_{\text{diff}}(v)$ with
 $\mathcal{E}_{\text{diff}}(v) = \int_{\Omega} v \log v - v \, dx$ and $\mathbb{K}_{\text{diff}}(v)\xi = -m \operatorname{div}(v\nabla\xi)$ JKO/Wasserstein

Pure heat equation for temperature

total entropy $\mathcal{S}(\theta) := \int_{\Omega} S(\theta(x)) \, dx$

total energy $\mathcal{E}(\theta) := \int_{\Omega} E(\theta(x)) \, dx$ with Gibbs relation $0 < c(\theta) = E'(\theta) = \frac{1}{\theta} S'(\theta)$
 (e.g. $S(\theta) = s_0 \log \theta$ and $E(\theta) = s_0 \theta$)

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Physical heat equation: $c(\theta)\dot{\theta} = \operatorname{div}(\kappa(\theta)\nabla\theta)$

Physical gradient structure $\dot{\theta} = +\mathbb{K}_{\text{heat}}(\theta)D\mathcal{S}(\theta)$

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Physical gradient structure $\dot{\theta} = +\mathbb{K}_{\text{heat}}(\theta)D\mathcal{S}(\theta)$

Only choice: $\mathbb{K}_{\text{heat}}(\theta)\xi = -\frac{1}{E'(\theta)} \operatorname{div}\left(\mu(\theta)\nabla\left(\frac{\xi}{E'(\theta)}\right)\right)$ (note $\mathbb{K}_{\text{heat}}(\theta)D\mathcal{E}(\theta) \equiv 0!$)

Using $\frac{S'(\theta)}{E'(\theta)} = \frac{1}{\theta}$ we have to choose $\mu(\theta) = \theta^2 \kappa(\theta)$

3. Motivating examples

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Physical heat equation: $c(\theta)\dot{\theta} = \operatorname{div}(\kappa(\theta)\nabla\theta)$

Physical gradient structure $\dot{\theta} = +\mathbb{K}_{\text{heat}}(\theta)D\mathcal{S}(\theta)$

For coupling it is better to use the **internal energy** $u = E(\theta)$ as variable

$\widehat{\mathcal{E}}(u) = \int_{\Omega} u(x) \, dx$ and $\widehat{\mathcal{S}}(u) = \int_{\Omega} \widehat{S}(u(x)) \, dx$

Equivalent gradient structure $\dot{u} = \widehat{\mathbb{K}}(u)D\widehat{\mathcal{S}}(u)$ with $\widehat{\mathbb{K}}(u)\eta = -\operatorname{div}(\widehat{\mu}(u)\nabla\eta)$