

CS 247 – Scientific Visualization

Lecture 25: Vector / Flow Visualization, Pt. 4

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Reading Assignment #13 (until Apr 25)



Read (required):

- Data Visualization book
 - Chapter 6.1 (Divergence and Vorticity)
- Diffeomorphisms / smooth deformations
<https://en.wikipedia.org/wiki/Diffeomorphism>
- Integral curves: Stream lines, path lines, streak lines
https://en.wikipedia.org/wiki/Integral_curve
https://en.wikipedia.org/wiki/Streamlines,_streaklines,_and_pathlines
- Paper:
Bruno Jobard and Wilfrid Lefer
Creating Evenly-Spaced Streamlines of Arbitrary Density,
<http://citeseerx.ist.psu.edu/viewdoc/summary?doi=10.1.1.29.9498>

Quiz #3: Apr 25



Organization

- First 30 min of lecture
- No material (book, notes, ...) allowed

Content of questions

- Lectures (both actual lectures and slides)
- Reading assignments (except optional ones)
- Programming assignments (algorithms, methods)
- Solve short practical examples

Vector fields

A **static vector field** $\mathbf{v}(\mathbf{x})$ is a vector-valued function of space.

A **time-dependent vector field** $\mathbf{v}(\mathbf{x}, t)$ depends also on time.

In the case of **velocity** fields, the terms **steady** and **unsteady flow** are used.

The dimensions of \mathbf{x} and \mathbf{v} are equal, often 2 or 3, and we denote components by x, y, z and u, v, w :

$$\mathbf{x} = (x, y, z), \quad \mathbf{v} = (u, v, w)$$

Sometimes a vector field is defined on a surface $\mathbf{x}(i, j)$. The vector field is then a function of parameters and time:

$$\mathbf{v}(i, j, t)$$

Steady vs. Unsteady Flow



- Steady flow: time-independent

- Flow itself is static over time: $\mathbf{v}(\mathbf{x})$

$$\mathbf{v}: \mathbb{R}^n \rightarrow \mathbb{R}^n,$$
$$x \mapsto \mathbf{v}(x).$$

- Example: laminar flows

- Unsteady flow: time-dependent

- Flow itself changes over time: $\mathbf{v}(\mathbf{x}, t)$

$$\mathbf{v}: \mathbb{R}^n \times \mathbb{R} \rightarrow \mathbb{R}^n,$$
$$(x, t) \mapsto \mathbf{v}(x, t).$$

- Example: turbulent flows

(here just for Euclidean domain; analogous on general manifolds)

Steady vs. Unsteady Flow



- Steady flow: time-independent

- Flow itself is static over time: $\mathbf{v}(\mathbf{x})$

$$\mathbf{v}: M \rightarrow \mathbb{R}^n,$$
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- Flow itself changes over time: $\mathbf{v}(\mathbf{x}, t)$

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- Example: turbulent flows

(here just for Euclidean domain; analogous on general manifolds)

Vector fields as ODEs

For simplicity, the vector field is now interpreted as a **velocity** field.

Then the field $\mathbf{v}(\mathbf{x}, t)$ describes the connection between location and velocity of a (massless) particle.

It can equivalently be expressed as an **ordinary differential equation**

$$\dot{\mathbf{x}}(t) = \mathbf{v}(\mathbf{x}(t), t)$$

This ODE, together with an **initial condition**

$$\mathbf{x}(t_0) = \mathbf{x}_0 ,$$

is a so-called **initial value problem** (IVP).

Its solution is the **integral curve** (or **trajectory**)

$$\mathbf{x}(t) = \mathbf{x}_0 + \int_{t_0}^t \mathbf{v}(\mathbf{x}(\tau), \tau) d\tau$$

Vector fields as ODEs

The integral curve is a **pathline**, describing the **path** of a massless **particle** which was released at time t_0 at position x_0 .

Remark: $t < t_0$ is allowed.

For static fields, the ODE is **autonomous**:

$$\dot{\mathbf{x}}(t) = \mathbf{v}(\mathbf{x}(t))$$

and its integral curves

$$\mathbf{x}(t) = \mathbf{x}_0 + \int_{t_0}^t \mathbf{v}(\mathbf{x}(\tau)) d\tau$$

are called **field lines**, or (in the case of velocity fields) **streamlines**.

Vector fields as ODEs

In **static** vector fields, pathlines and streamlines are **identical**.

In **time-dependent** vector fields, **instantaneous streamlines** can be computed from a "snapshot" at a fixed time T (which is a static vector field)

$$\mathbf{v}_T(\mathbf{x}) = \mathbf{v}(\mathbf{x}, T)$$

In practice, time-dependent fields are often given as a dataset per time step. Each dataset is then a snapshot.

Streamline integration

Outline of algorithm for numerical streamline integration
(with obvious extension to pathlines):

Inputs:

- static vector field $\mathbf{v}(\mathbf{x})$
- seed points with time of release (\mathbf{x}_0, t_0)
- control parameters:
 - step size (temporal, spatial, or in local coordinates)
 - step count limit, time limit, etc.
 - order of integration scheme

Output:

- streamlines as "polylines", with possible attributes
(interpolated field values, time, speed, arc length, etc.)

Streamline integration

Preprocessing:

- set up search structure for point location
- for each seed point:
 - **global point location**: Given a point \mathbf{x} , find the cell containing \mathbf{x} and the local coordinates (ξ, η, ζ) or if the grid is structured:
find the computational space coordinates $(i + \xi, j + \eta, k + \zeta)$
 - If \mathbf{x} is not found in a cell, remove seed point

Streamline integration

Integration loop, for each seed point \mathbf{x} :

- interpolate \mathbf{v} trilinearly to local coordinates (ξ, η, ζ)
- do an integration step, producing a new point \mathbf{x}'
- **incremental point location**: For position \mathbf{x}' find cell and local coordinates (ξ', η', ζ') making use of information (coordinates, local coordinates, cell) of old point \mathbf{x}

Termination criteria:

- grid boundary reached
- step count limit reached
- optional: velocity close to zero
- optional: time limit reached
- optional: arc length limit reached

Streamline integration

Integration step: widely used integration methods:

- **Euler** (used only in special speed-optimized techniques, e.g. GPU-based texture advection)

$$\mathbf{x}_{new} = \mathbf{x} + \mathbf{v}(\mathbf{x}, t) \cdot \Delta t$$

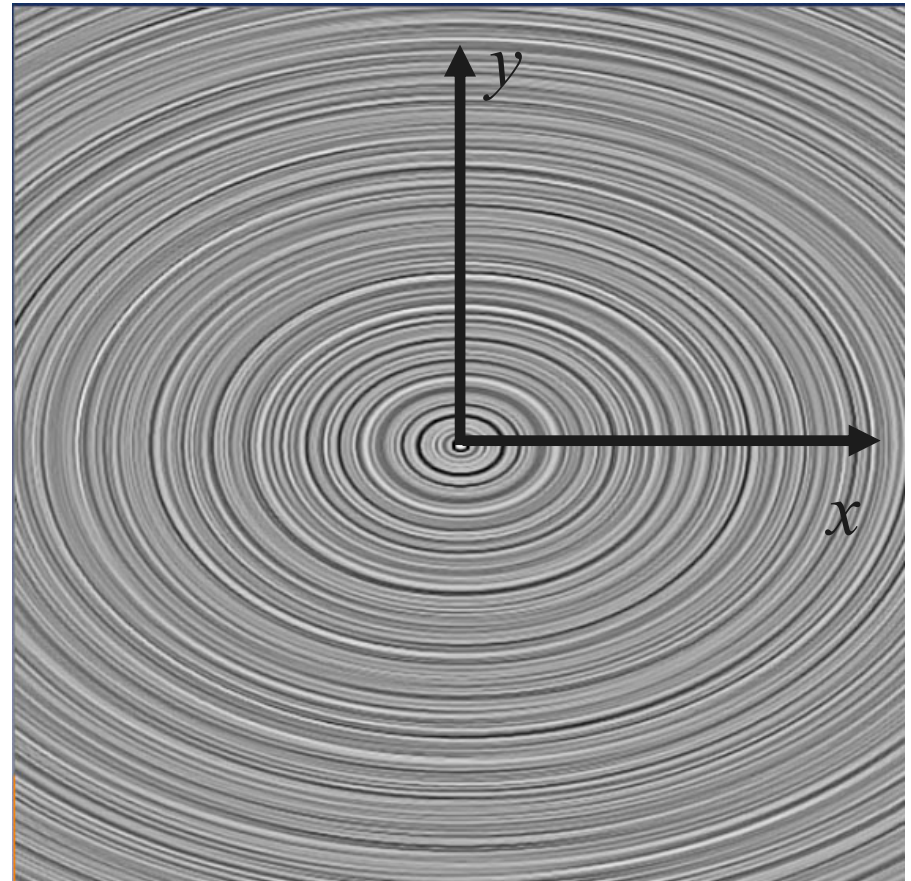
- **Runge-Kutta**, 2nd or 4th order

Higher order than 4th?

- often too slow for visualization
- study (Yeung/Pope 1987) shows that, when using standard trilinear interpolation, **interpolation errors** dominate **integration errors**.

Numerical Integration

- **Numerical integration of stream lines:**
- approximate streamline by polygon \mathbf{x}_i
- **Testing example:**
 - $\mathbf{v}(x,y) = (-y, x/2)^T$
 - exact solution: ellipses
 - starting integration from $(0,-1)$



- Basic approach:

- theory: $\mathbf{s}(t) = \mathbf{s}_0 + \int_{0 \leq u \leq t} \mathbf{v}(\mathbf{s}(u)) du$

- practice: numerical integration

- idea:

- (very) locally, the solution is (approx.) linear

- Euler integration:

- follow the current flow vector $\mathbf{v}(\mathbf{s}_i)$ from the current streamline point \mathbf{s}_i for a very small time (dt) and therefore distance

- Euler integration: $\mathbf{s}_{i+1} = \mathbf{s}_i + dt \cdot \mathbf{v}(\mathbf{s}_i)$,
integration of small steps (dt very small)

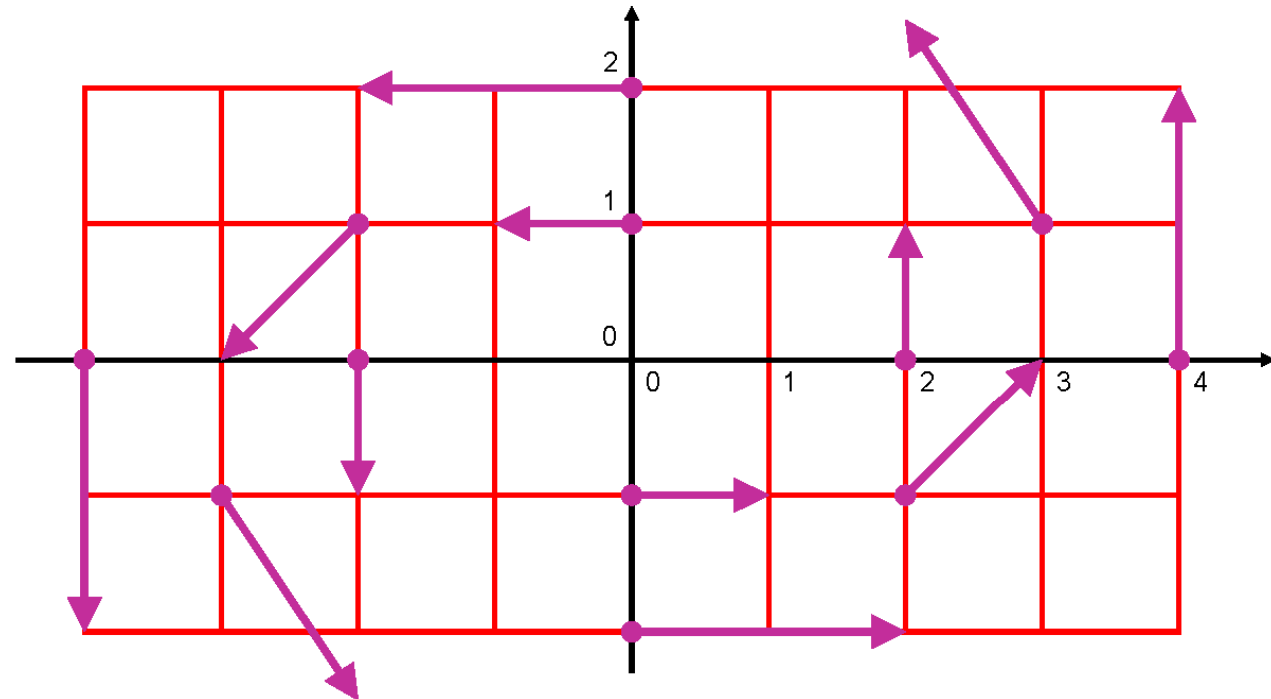
Euler Integration – Example

- 2D model data:

$$v_x = dx/dt = -y$$

$$v_y = dy/dt = x/2$$

- Sample arrows:

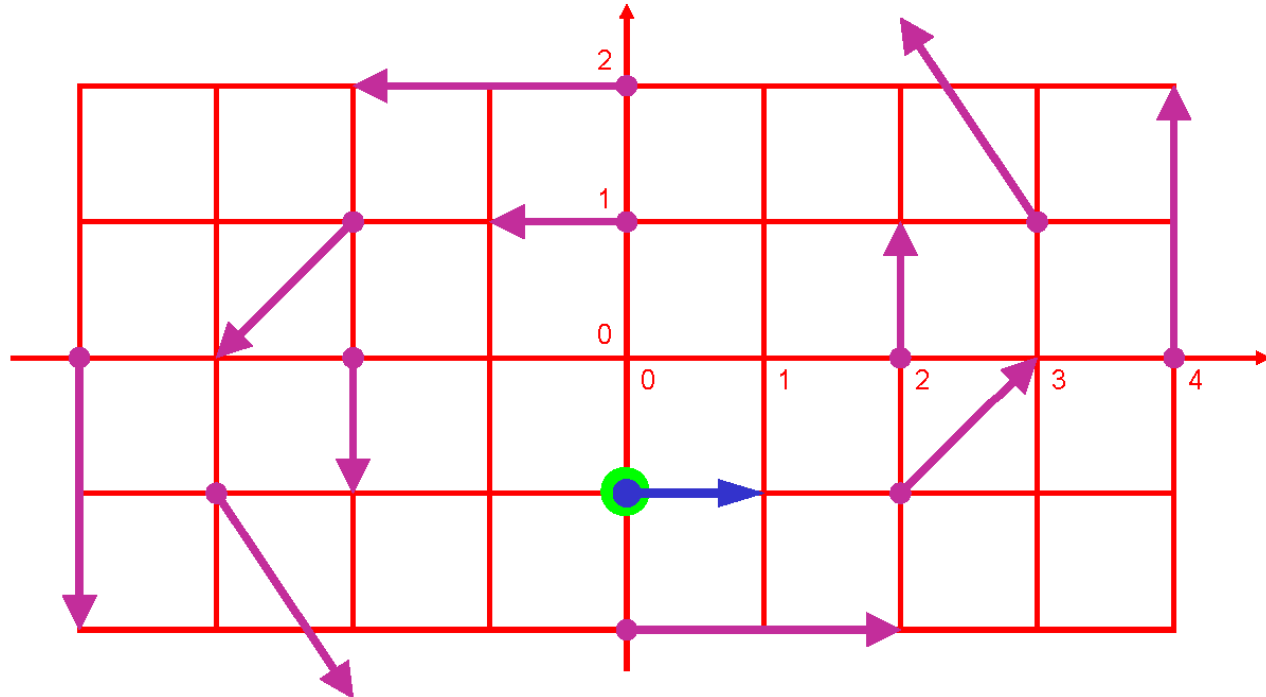


- True solution: ellipses!

Euler Integration – Example

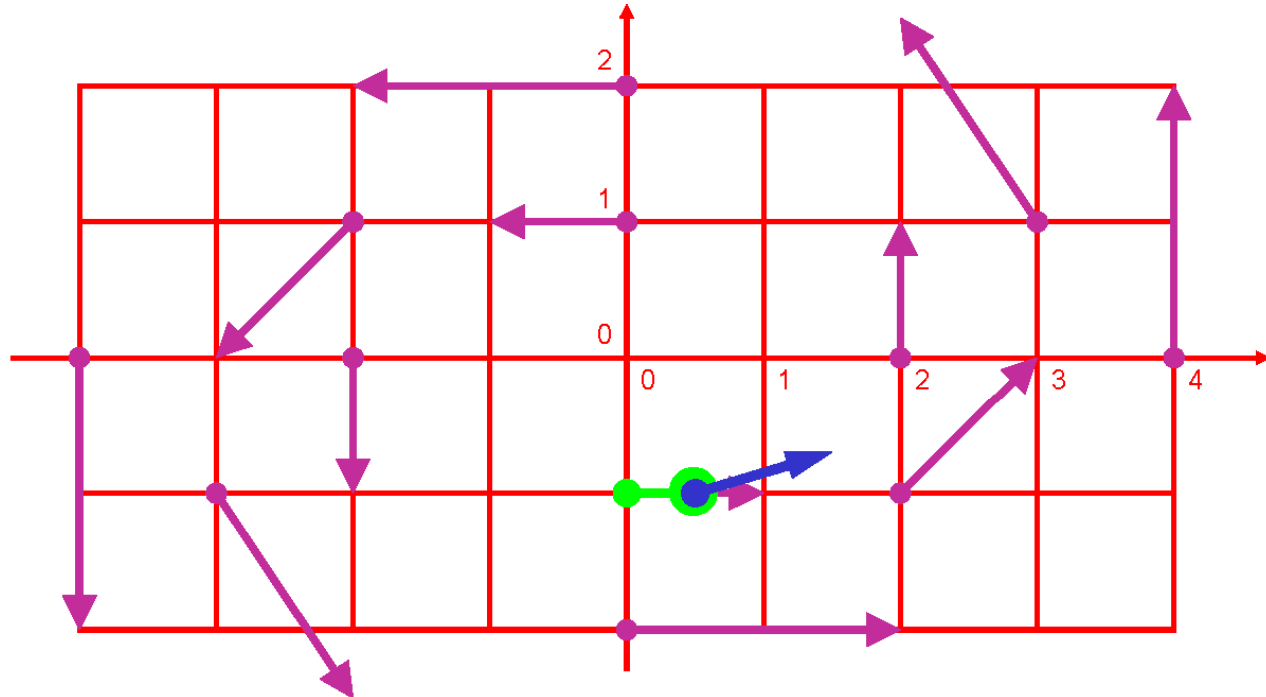


- Seed point $\mathbf{s}_0 = (0 \mid -1)^T$;
 current flow vector $\mathbf{v}(\mathbf{s}_0) = (1 \mid 0)^T$;
 $dt = 1/2$



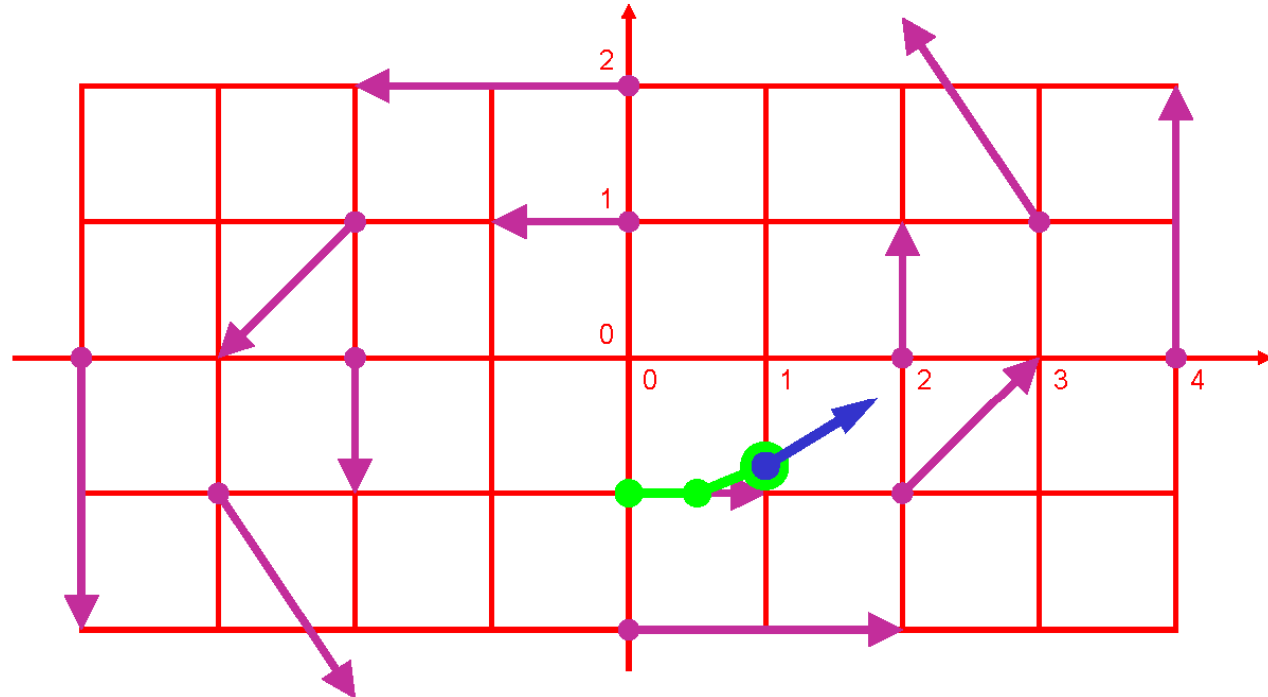
Euler Integration – Example

- New point $\mathbf{s}_1 = \mathbf{s}_0 + \mathbf{v}(\mathbf{s}_0) \cdot dt = (1/2 \mid -1)^T$;
 current flow vector $\mathbf{v}(\mathbf{s}_1) = (1 \mid 1/4)^T$;



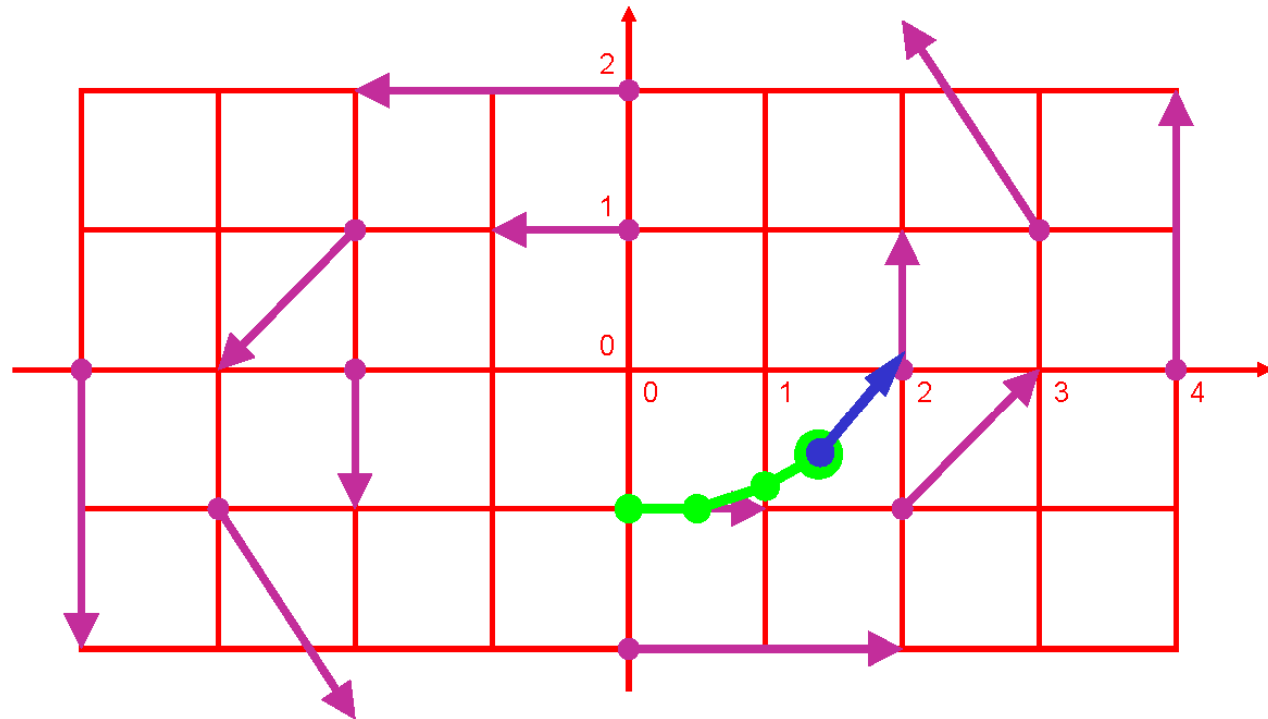
Euler Integration – Example

- New point $\mathbf{s}_2 = \mathbf{s}_1 + \mathbf{v}(\mathbf{s}_1) \cdot dt = (1 \mid -7/8)^T$;
 current flow vector $\mathbf{v}(\mathbf{s}_2) = (7/8 \mid 1/2)^T$;



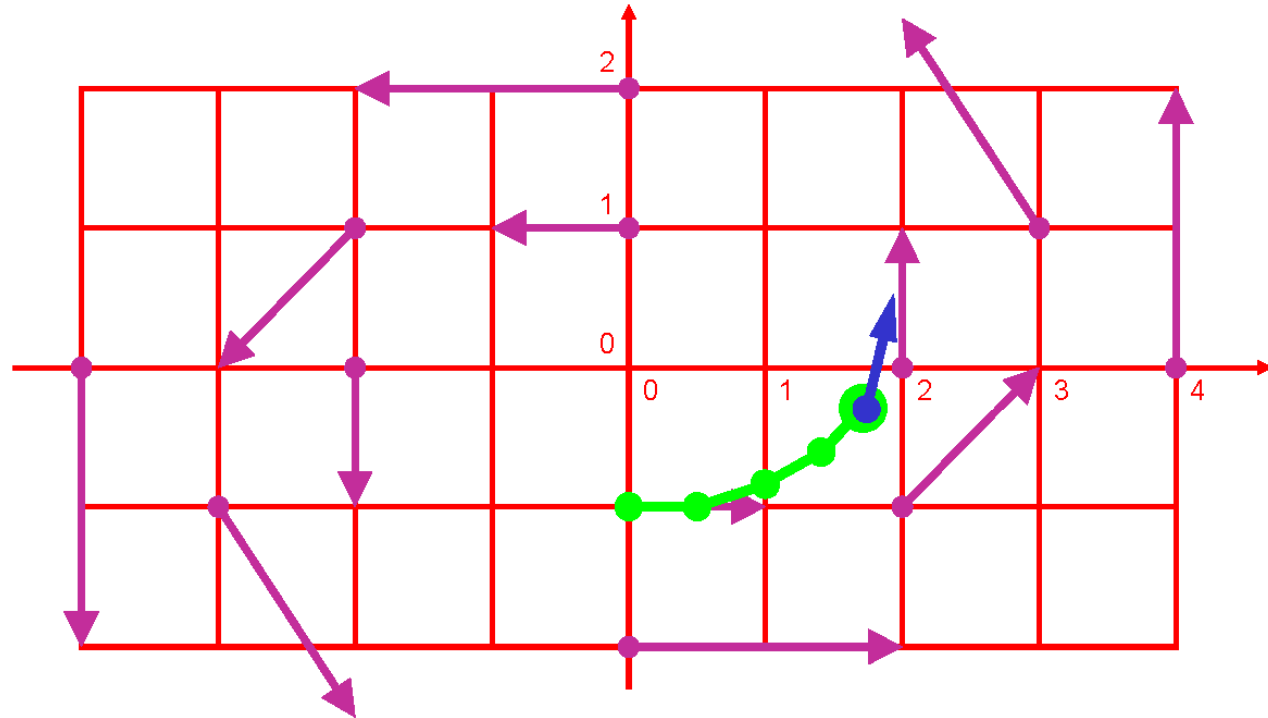
Euler Integration – Example

$$\begin{aligned} \blacksquare \mathbf{s}_3 &= (23/16 | -5/8)^T \approx (1.44 | -0.63)^T; \\ \mathbf{v}(\mathbf{s}_3) &= (5/8 | 23/32)^T \approx (0.63 | 0.72)^T; \end{aligned}$$



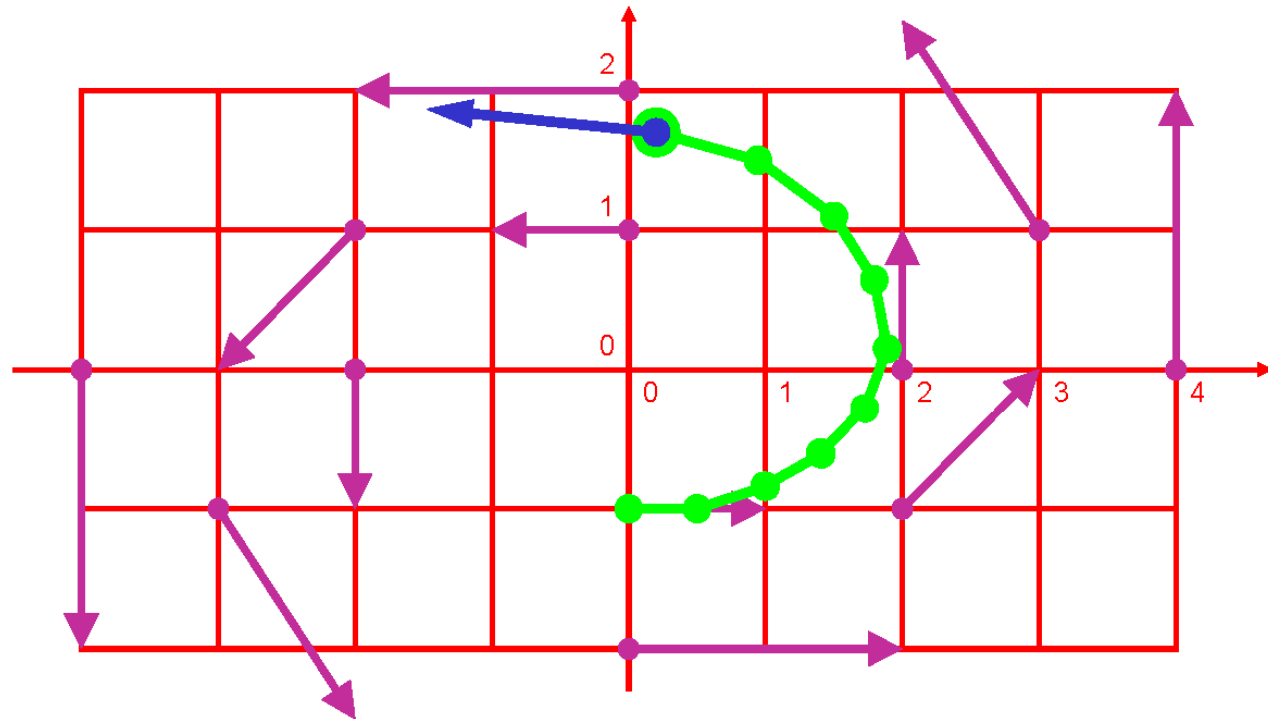
Euler Integration – Example

$$\begin{aligned}
 \blacksquare \mathbf{s}_4 &= (7/4 \mid -17/64)^T \approx (1.75 \mid -0.27)^T; \\
 \mathbf{v}(\mathbf{s}_4) &= (17/64 \mid 7/8)^T \approx (0.27 \mid 0.88)^T;
 \end{aligned}$$



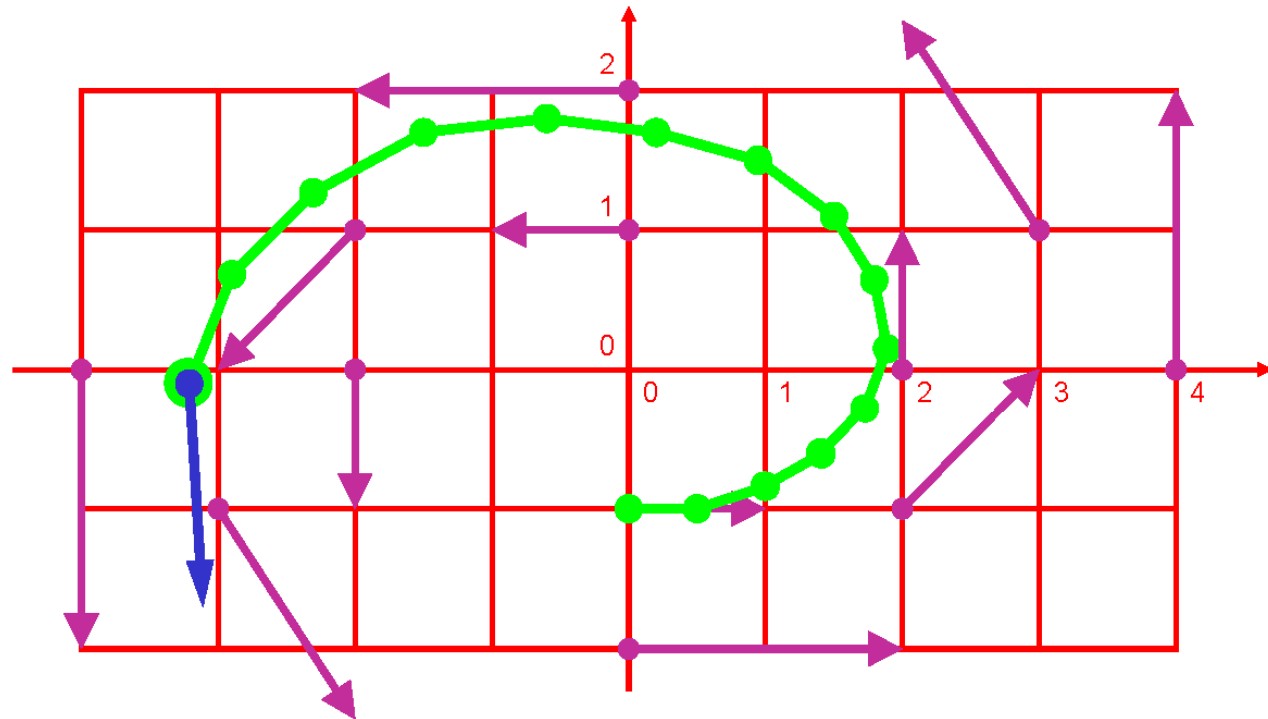
Euler Integration – Example

■ $\mathbf{s}_9 \approx (0.20 \mid 1.69)^T;$
 $\mathbf{v}(\mathbf{s}_9) \approx (-1.69 \mid 0.10)^T;$



Euler Integration – Example

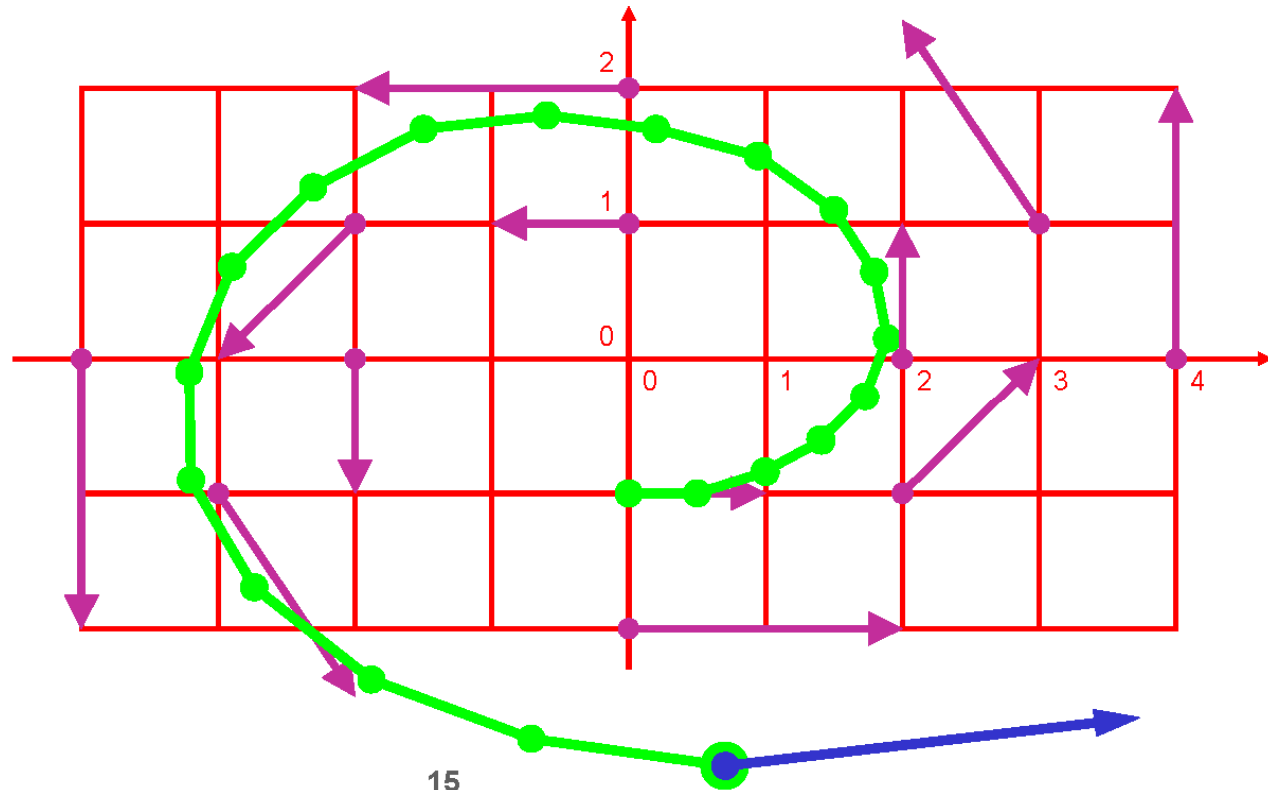
■ $\mathbf{s}_{14} \approx (-3.22 \mid -0.10)^T$;
 $\mathbf{v}(\mathbf{s}_{14}) \approx (0.10 \mid -1.61)^T$



Euler Integration – Example

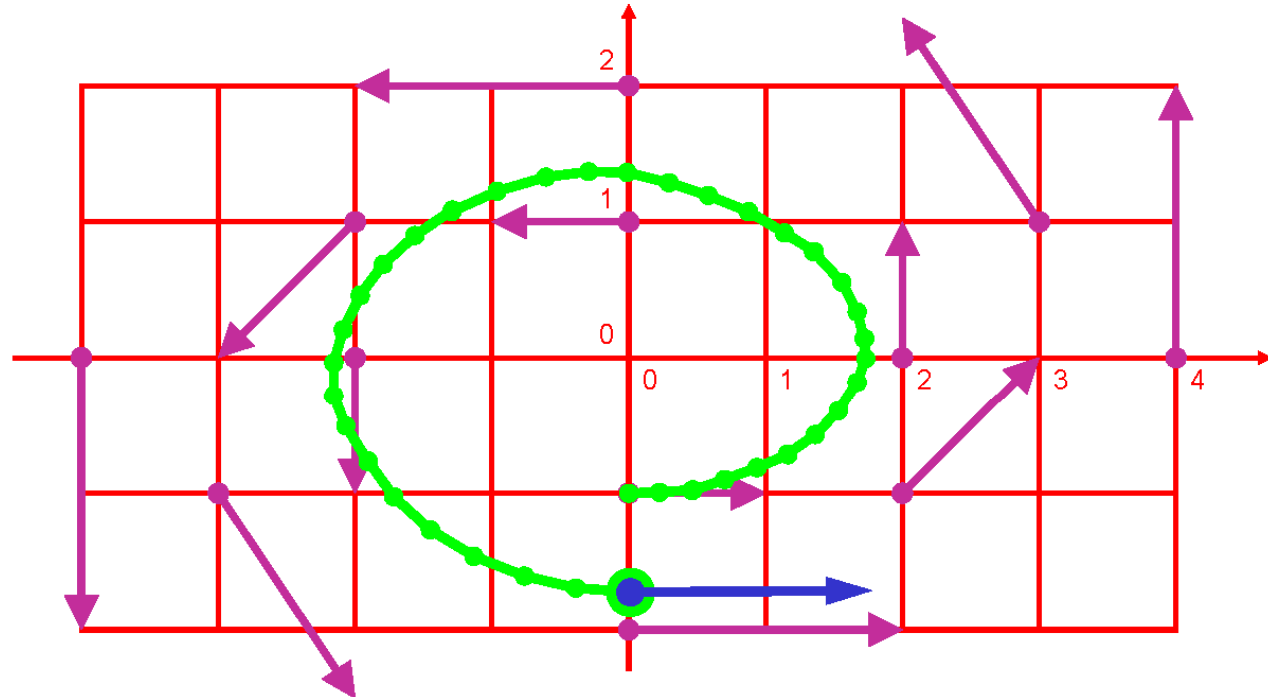


- $\mathbf{s}_{19} \approx (0.75 | -3.02)^T$; $\mathbf{v}(\mathbf{s}_{19}) \approx (3.02 | 0.37)^T$;
clearly: large integration error, dt too large!
19 steps



Euler Integration – Example

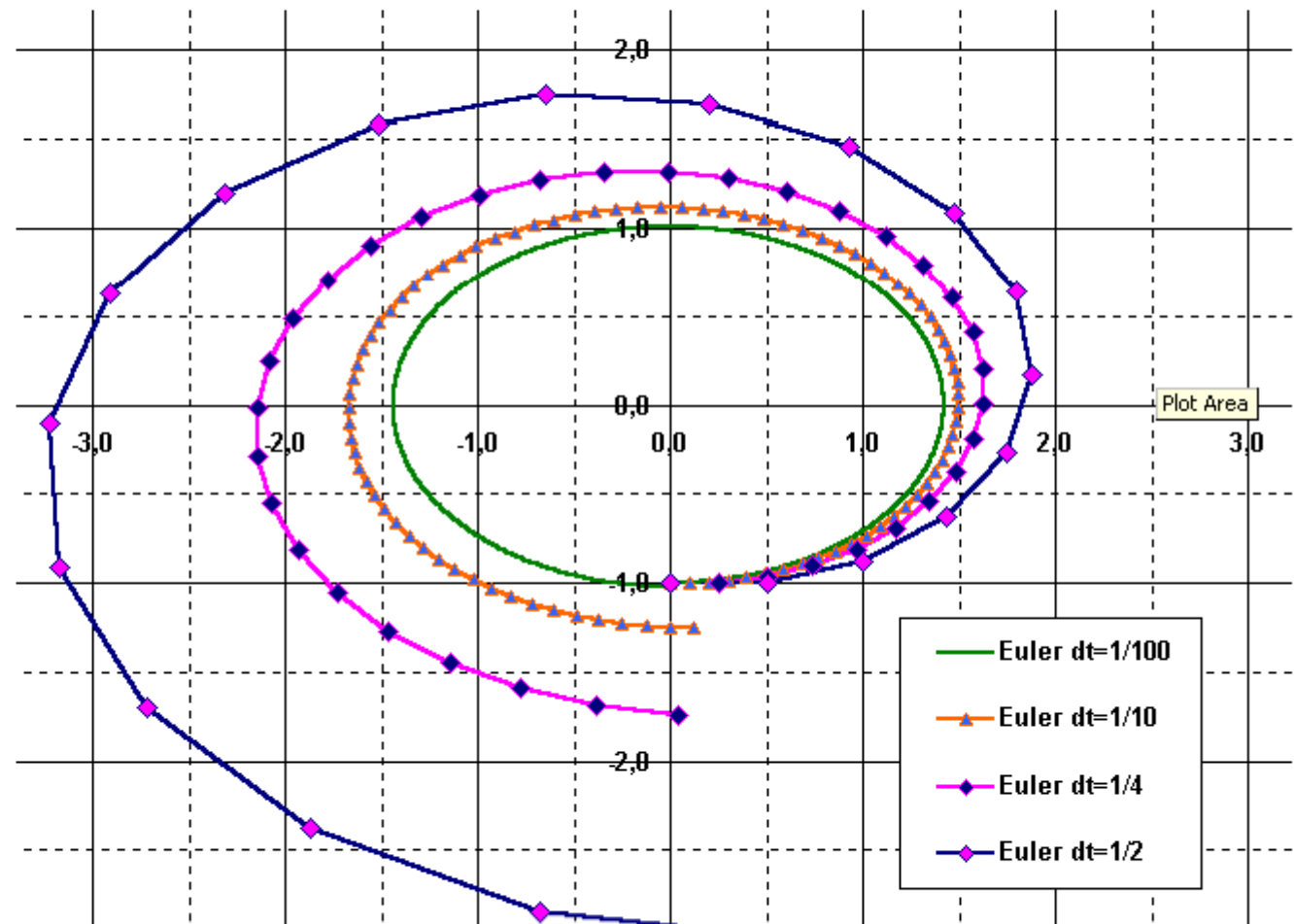
- dt smaller ($1/4$): more steps, more exact!
 $\mathbf{s}_{36} \approx (0.04 \mid -1.74)^T$; $\mathbf{v}(\mathbf{s}_{36}) \approx (1.74 \mid 0.02)^T$;
- 36 steps



Comparison Euler, Step Sizes



Euler
is getting
better
propor-
tionally
to dt



Better than Euler Integr.: RK



■ Runge-Kutta Approach:

- theory: $\mathbf{s}(t) = \mathbf{s}_0 + \int_{0 \leq u \leq t} \mathbf{v}(\mathbf{s}(u)) du$

- Euler: $\mathbf{s}_j = \mathbf{s}_0 + \sum_{0 \leq u < j} \mathbf{v}(\mathbf{s}_u) \cdot dt$

- Runge-Kutta integration:

- idea: cut short the curve arc

- RK-2 (second order RK):

- 1.: do half a Euler step

- 2.: evaluate flow vector there

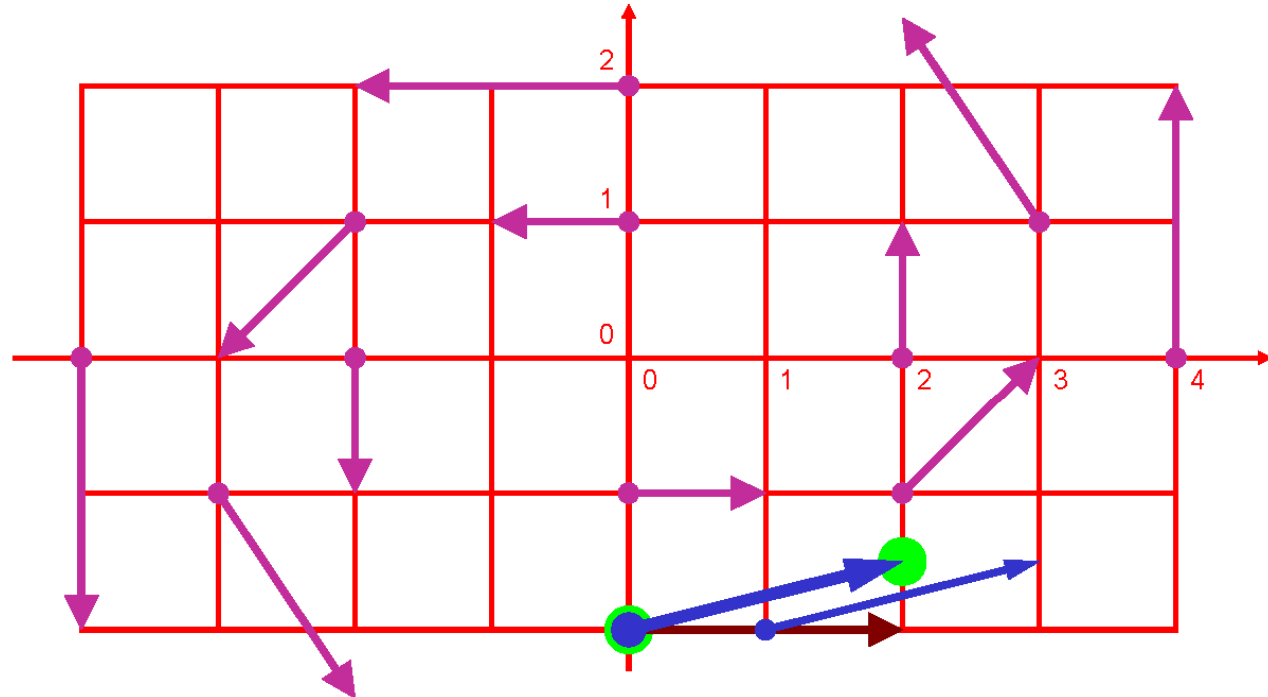
- 3.: use it in the origin

- RK-2 (two evaluations of \mathbf{v} per step):

- $\mathbf{s}_{j+1} = \mathbf{s}_j + \mathbf{v}(\mathbf{s}_j + \mathbf{v}(\mathbf{s}_j) \cdot dt/2) \cdot dt$

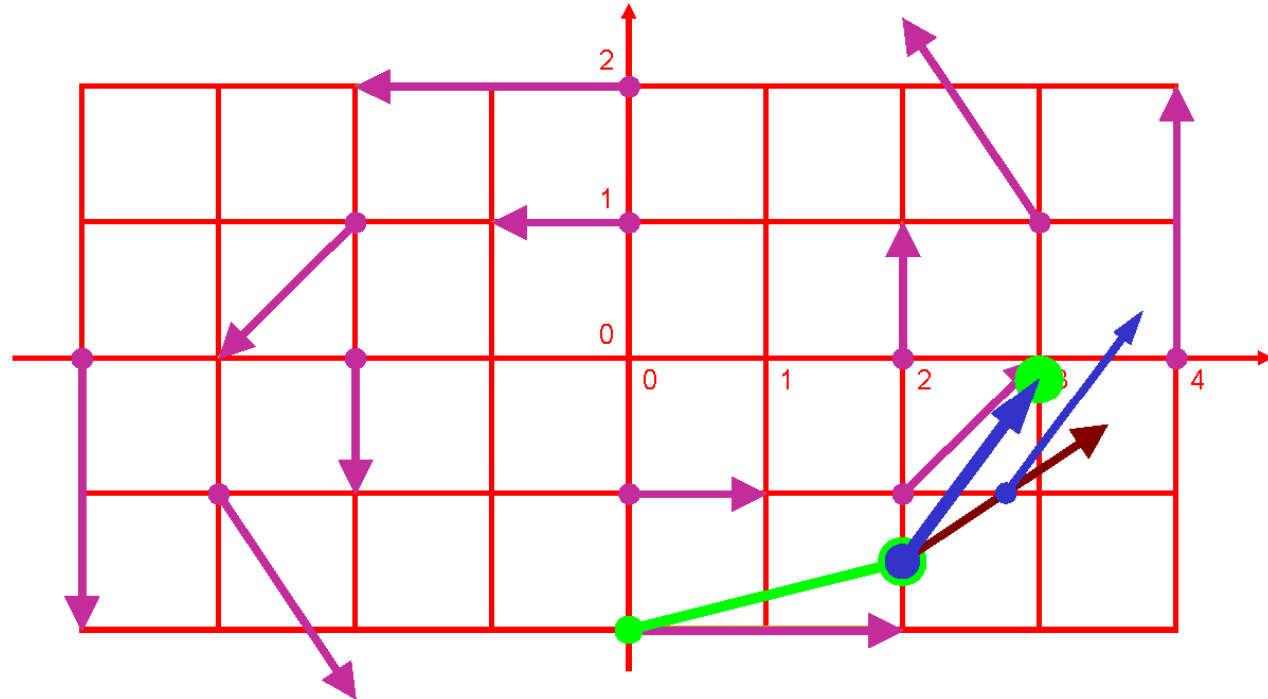
RK-2 Integration – One Step

- Seed point $\mathbf{s}_0 = (0 \mid -2)^T$;
- current flow vector $\mathbf{v}(\mathbf{s}_0) = (2 \mid 0)^T$;
- preview vector $\mathbf{v}(\mathbf{s}_0 + \mathbf{v}(\mathbf{s}_0) \cdot dt/2) = (2 \mid 0.5)^T$;
- $dt = 1$



RK-2 – One more step

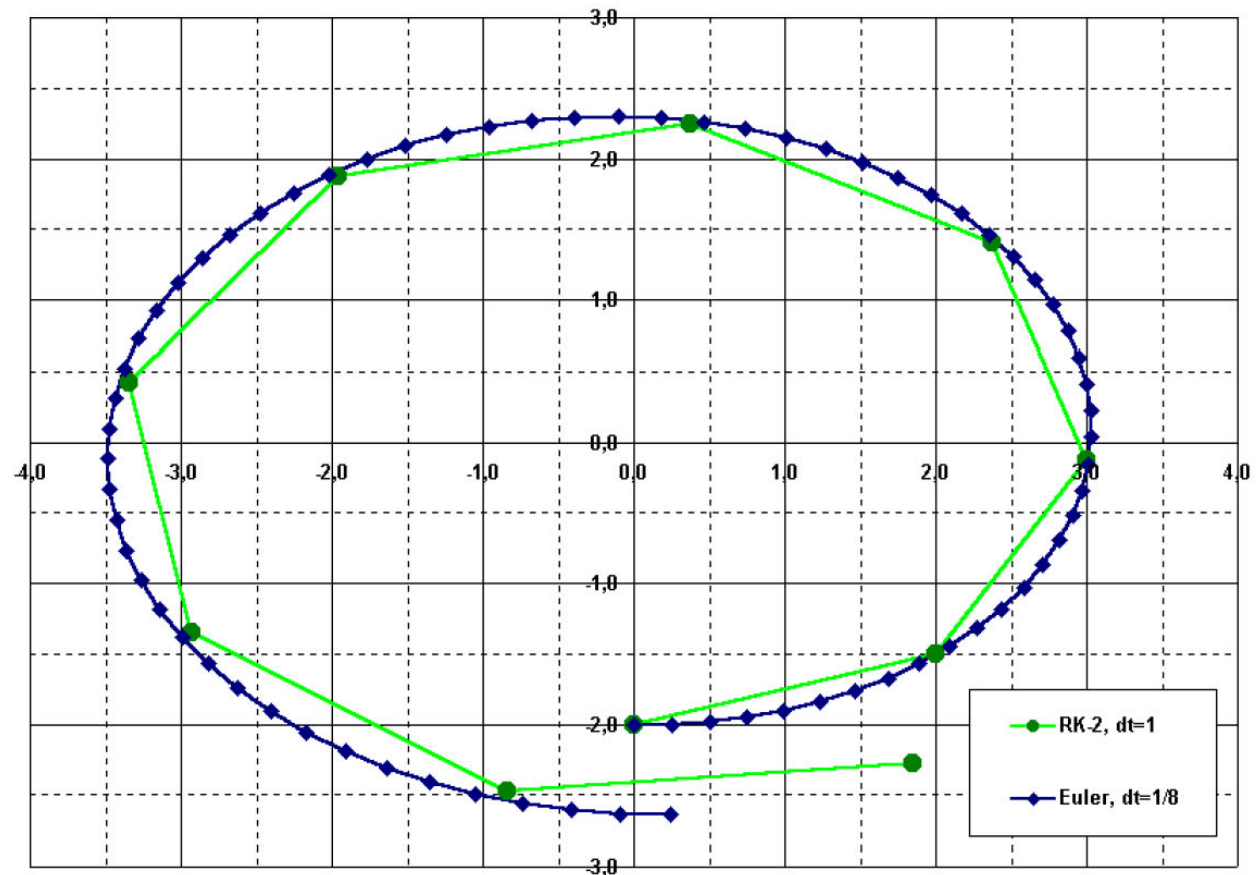
- Seed point $\mathbf{s}_1 = (2 \mid -1.5)^T$;
- current flow vector $\mathbf{v}(\mathbf{s}_1) = (1.5 \mid 1)^T$;
- preview vector $\mathbf{v}(\mathbf{s}_1 + \mathbf{v}(\mathbf{s}_1) \cdot dt/2) \approx (1 \mid 1.4)^T$;
- $dt = 1$



RK-2 – A Quick Round



- RK-2: even with $dt=1$ (9 steps) better than Euler with $dt=1/8$ (72 steps)



RK-4 vs. Euler, RK-2

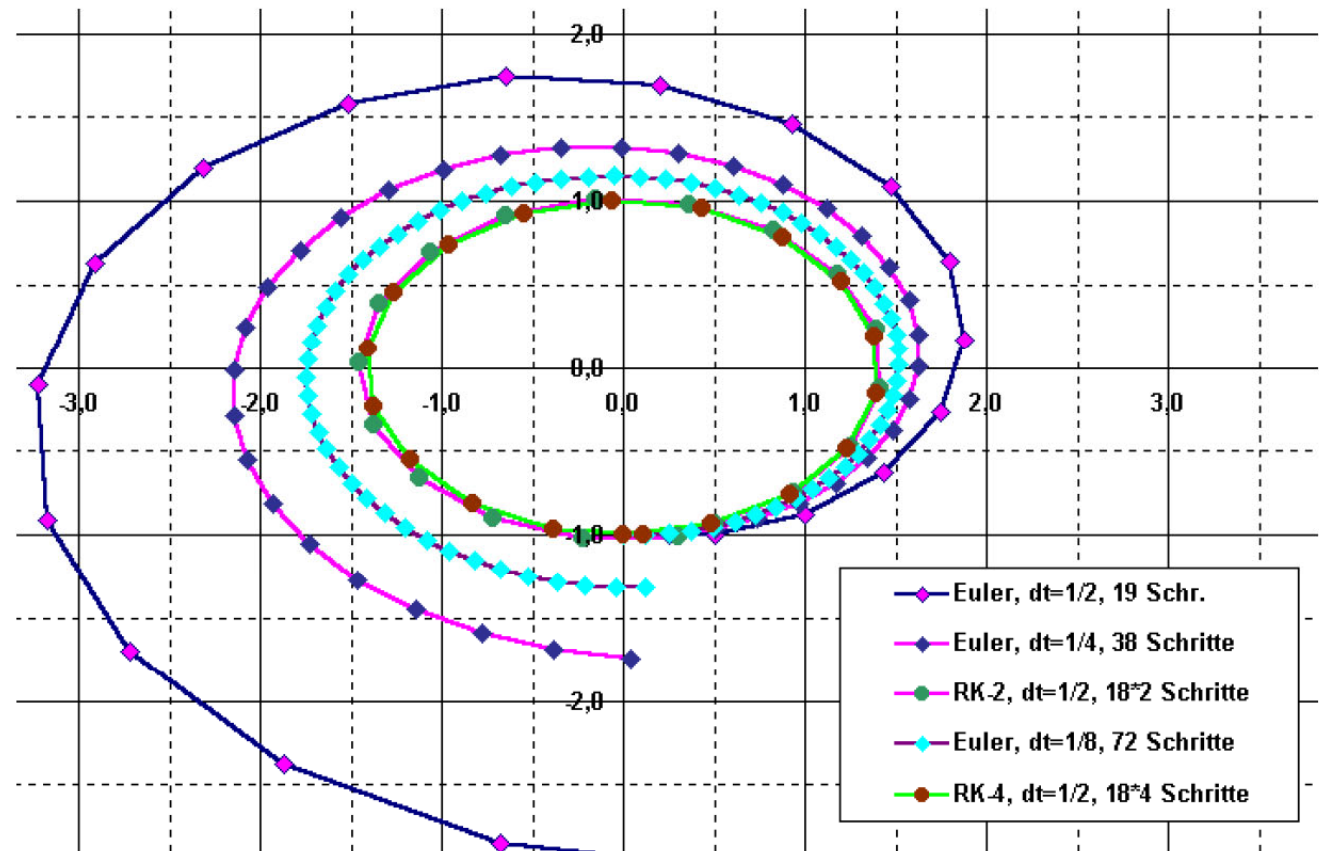


- Even better: fourth order RK:
 - four vectors **a**, **b**, **c**, **d**
 - one step is a convex combination:
$$\mathbf{s}_{i+1} = \mathbf{s}_i + (\mathbf{a} + 2 \cdot \mathbf{b} + 2 \cdot \mathbf{c} + \mathbf{d})/6$$
 - vectors:
 - $\mathbf{a} = dt \cdot \mathbf{v}(\mathbf{s}_i)$... original vector
 - $\mathbf{b} = dt \cdot \mathbf{v}(\mathbf{s}_i + \mathbf{a}/2)$... RK-2 vector
 - $\mathbf{c} = dt \cdot \mathbf{v}(\mathbf{s}_i + \mathbf{b}/2)$... use RK-2 ...
 - $\mathbf{d} = dt \cdot \mathbf{v}(\mathbf{s}_i + \mathbf{c})$... and again!

Euler vs. Runge-Kutta



- RK-4: pays off only with complex flows
- Here approx. like RK-2





■ Summary:

- analytic determination of streamlines usually not possible
- hence: numerical integration
- several methods available (Euler, Runge-Kutta, etc.)
- Euler: simple, imprecise, esp. with small dt
- RK: more accurate in higher orders
- furthermore: adaptive methods, implicit methods, etc.

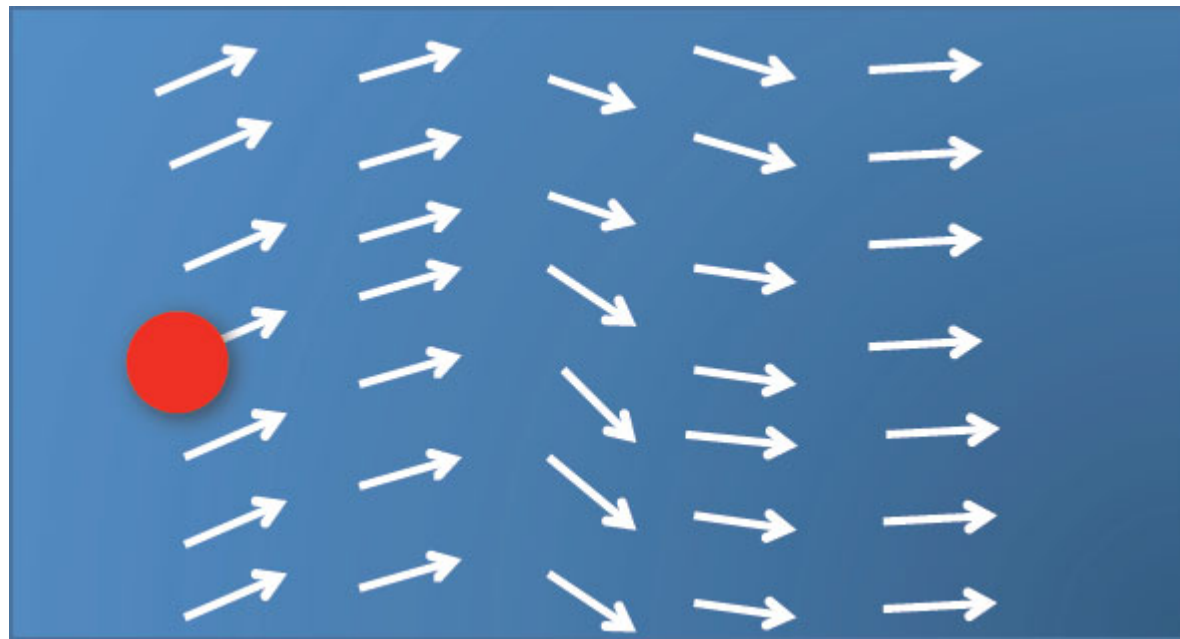
Integral Curves, Pt. 2

Particle Trajectories



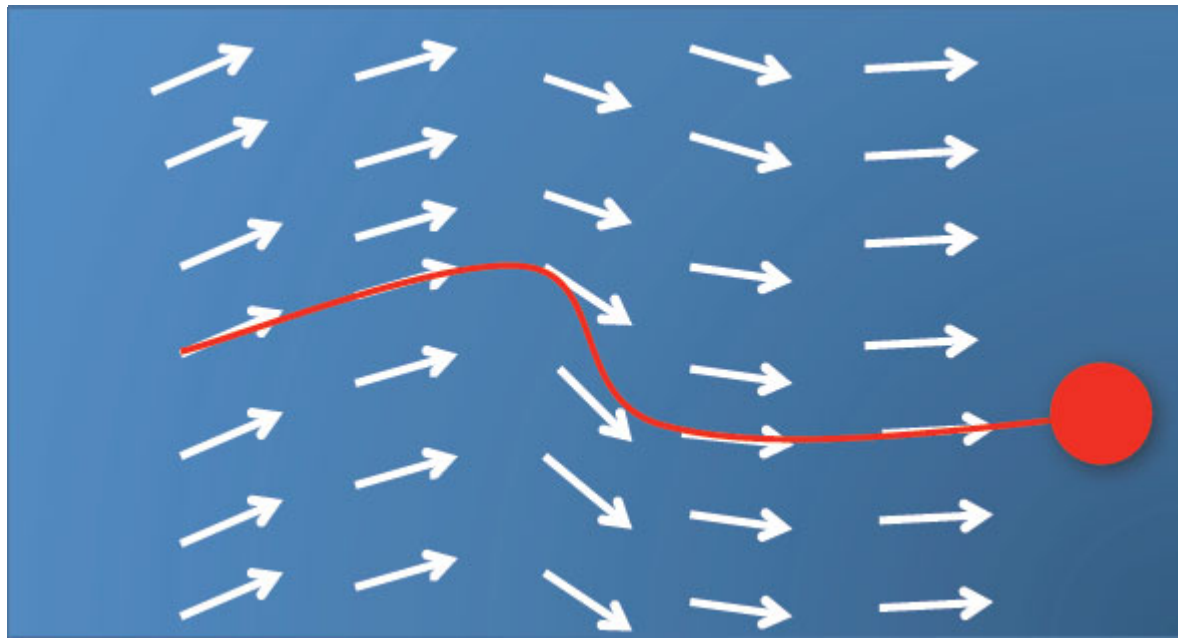
Courtesy Jens Krüger

Particle Trajectories



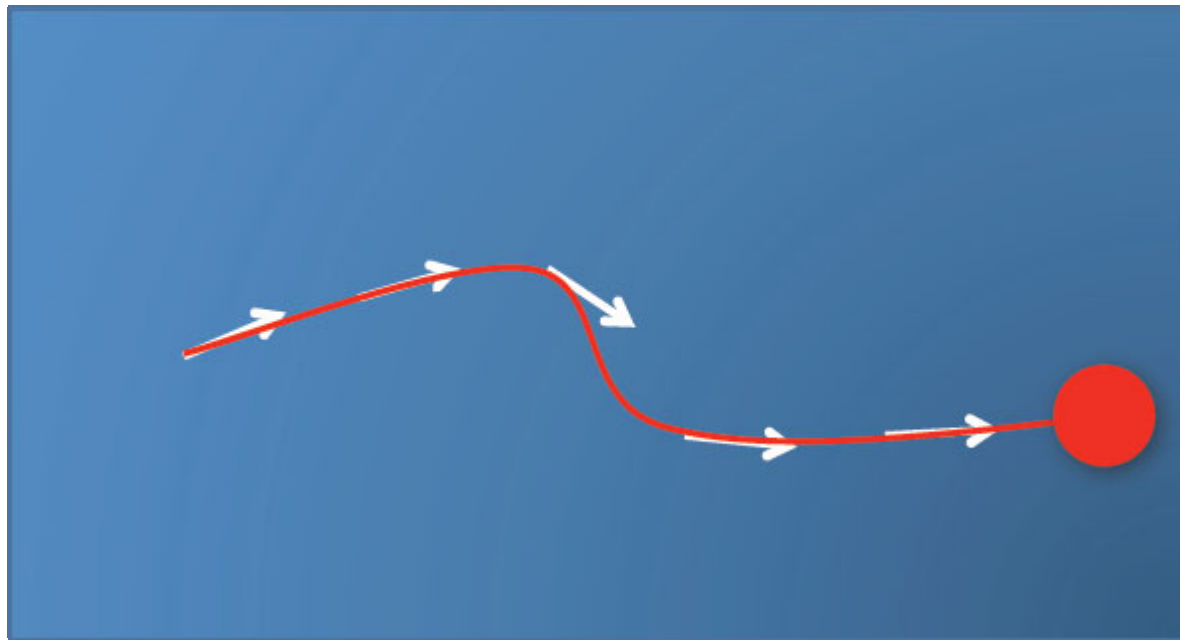
Courtesy Jens Krüger

Particle Trajectories



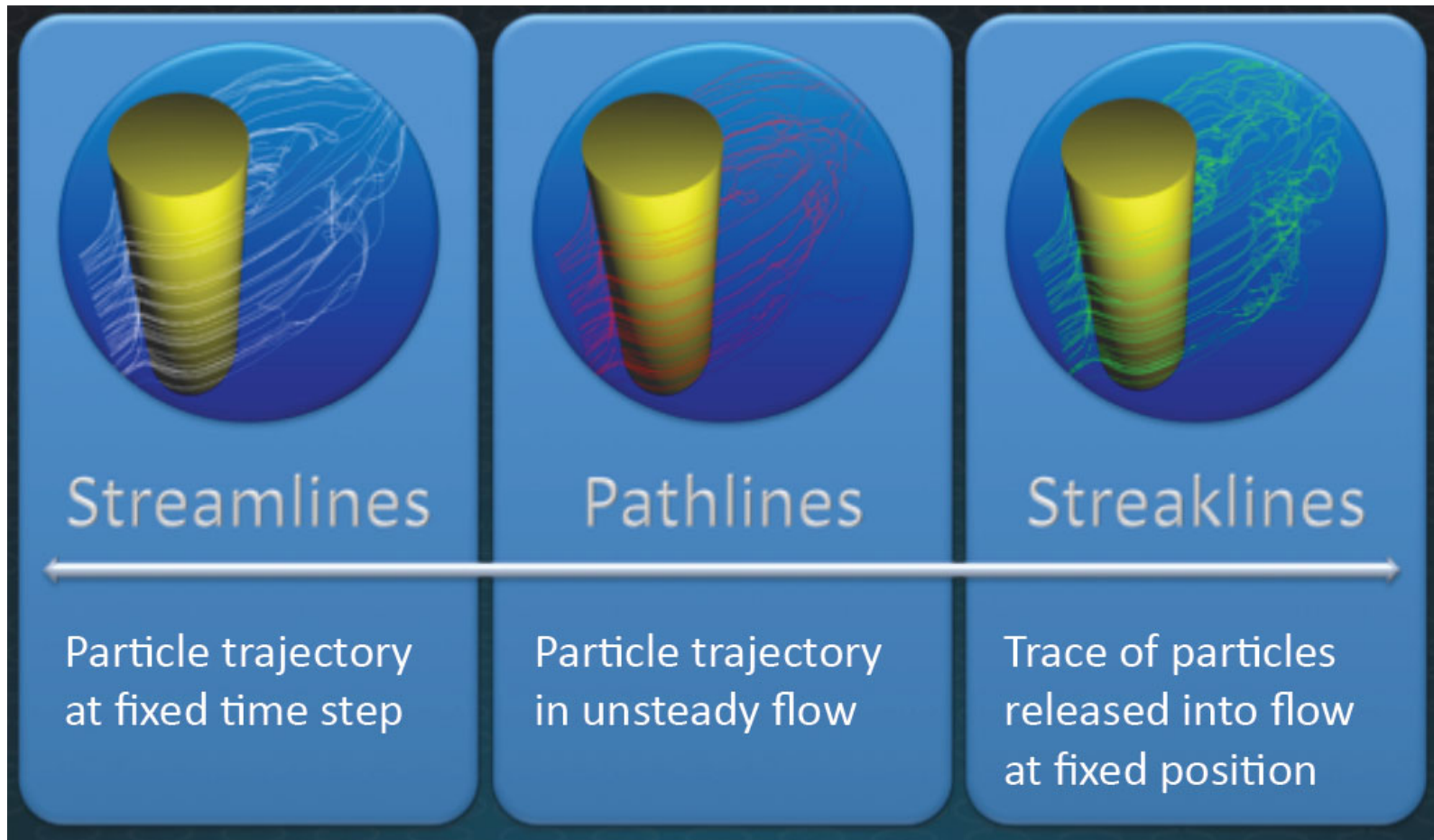
Courtesy Jens Krüger

Particle Trajectories



Courtesy Jens Krüger

Integral Curves



Streamline

- Curve parallel to the vector field in each point for a fixed time

Pathline

- Describes motion of a massless particle over time

Streakline

- Location of all particles released at a *fixed position* over time

Timeline

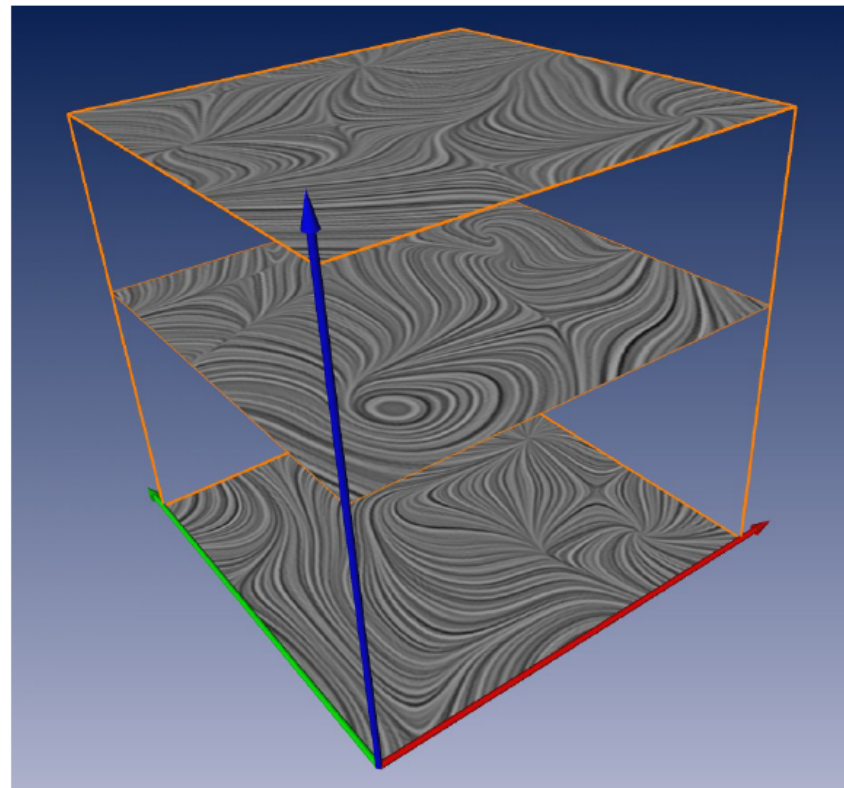
- Location of all particles released along a line at a *fixed time*

Streamlines Over Time



Defined only for steady flow or for a fixed time step (of unsteady flow)

Different tangent curves in every time step for time-dependent vector fields (unsteady flow)

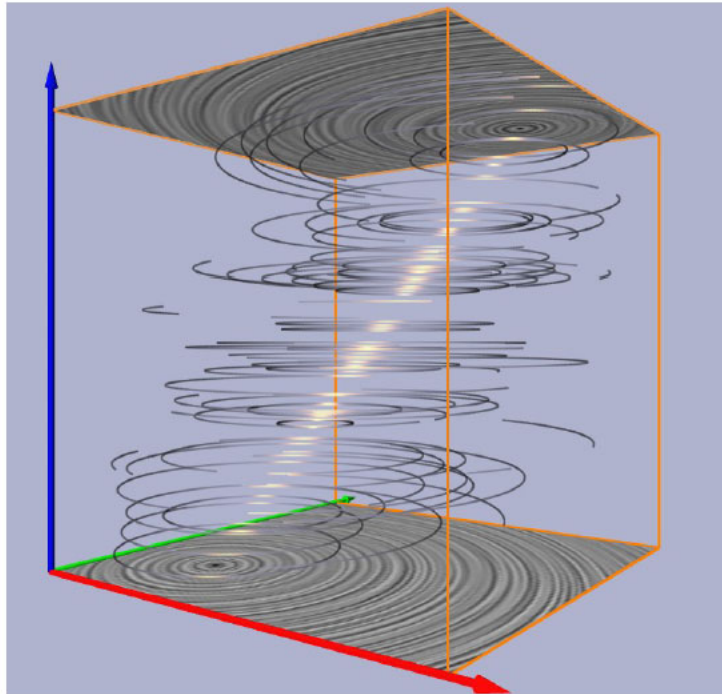


Stream Lines vs. Path Lines Viewed Over Time

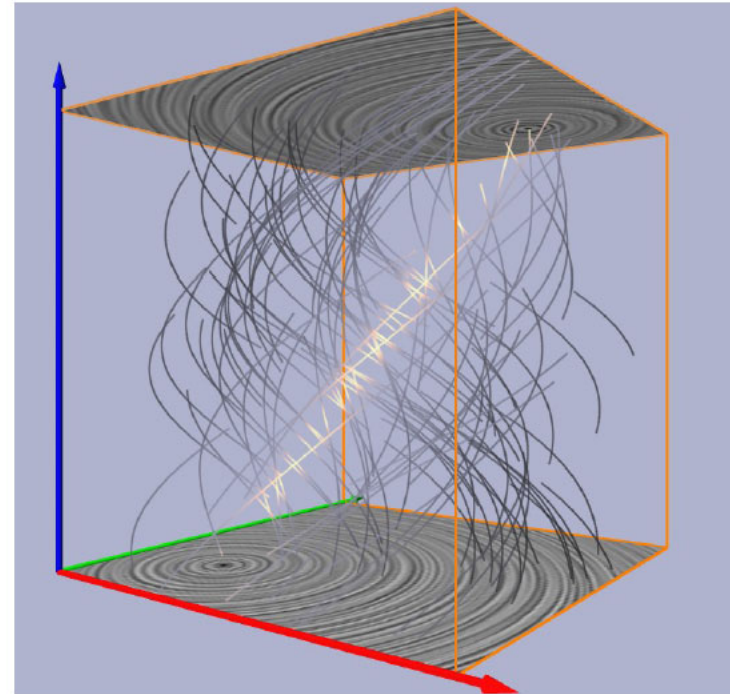


Plotted with time as third dimension

- Tangent curves to a $(n + 1)$ -dimensional vector field



Stream Lines



Path Lines

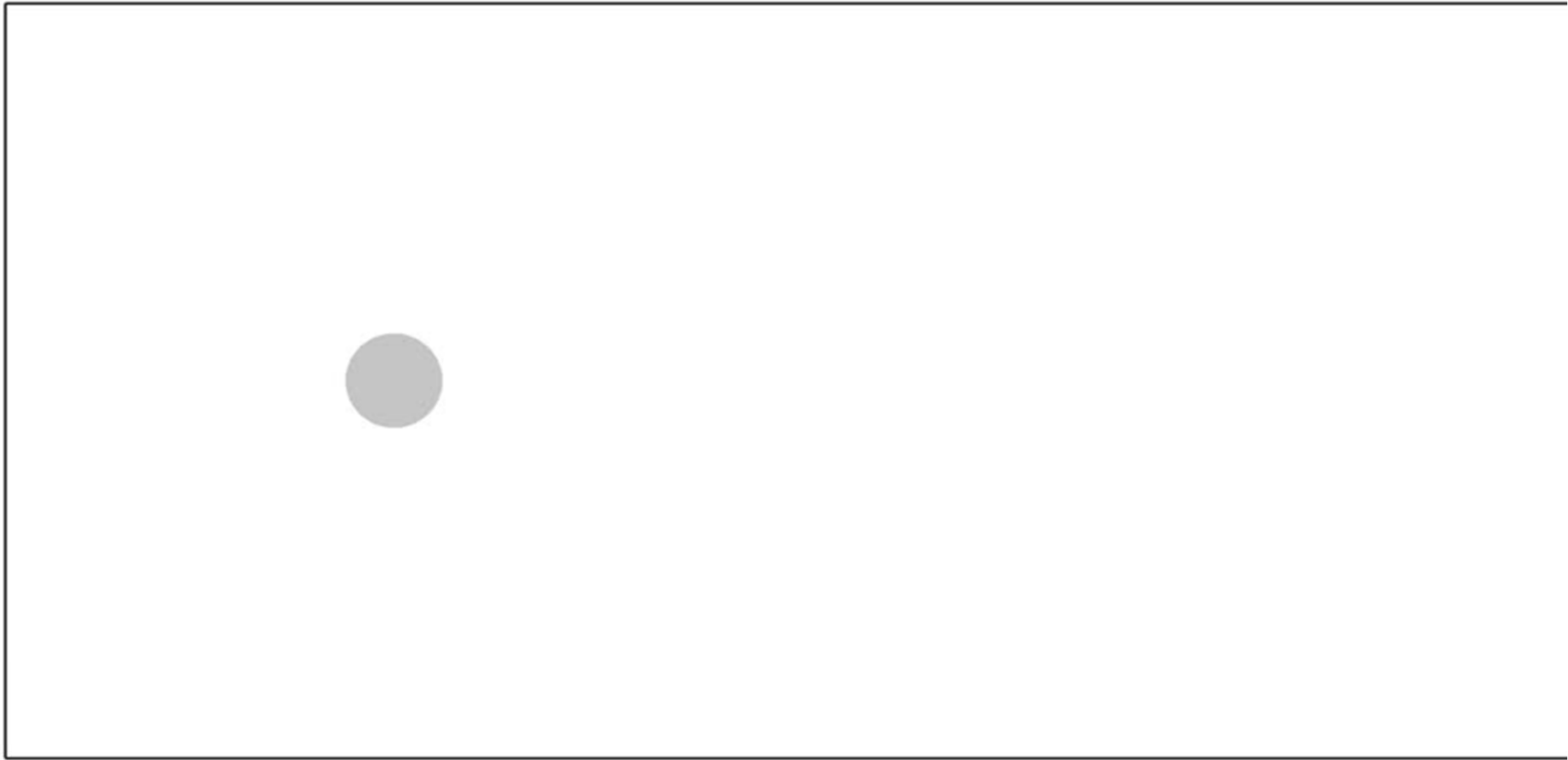


Time



streak line

location of all particles set out at a fixed point at different times



Particle visualization

2D time-dependent flow around a cylinder

time line

location of all particles set out on a certain line at a fixed time

The Flow / Flow Map of a Vector Field (1)



Flow of a *steady (time-independent)* vector field

- Map source position x “forward” ($t > 0$) or “backward” ($t < 0$) by time t

$$\boxed{\phi(x, t)}$$

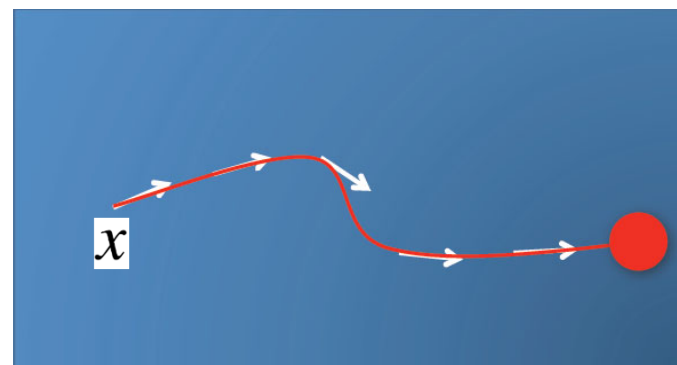
$$\phi: \mathbb{R}^n \times \mathbb{R} \rightarrow \mathbb{R}^n, \\ (x, t) \mapsto \phi(x, t).$$

$$\boxed{\phi_t(x)}$$

$$\phi_t: \mathbb{R}^n \rightarrow \mathbb{R}^n, \\ x \mapsto \phi_t(x).$$

with $\phi_0(x) = x$

$$\phi_s(\phi_t(x)) = \phi_{s+t}(x)$$



The Flow / Flow Map of a Vector Field (1)



Flow of a *steady (time-independent)* vector field

- Map source position x “forward” ($t > 0$) or “backward” ($t < 0$) by time t

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$$\phi: M \times \mathbb{R} \rightarrow M, \\ (x, t) \mapsto \phi(x, t).$$

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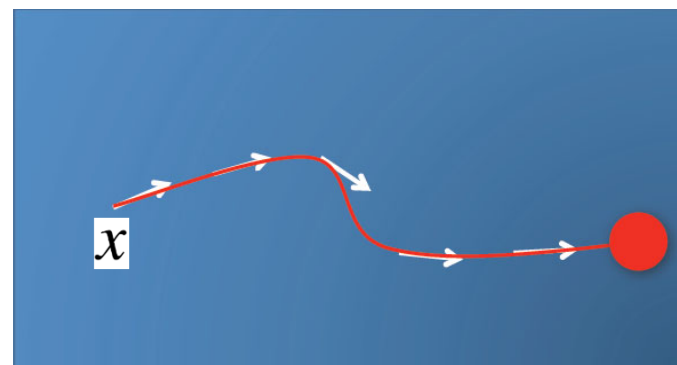
$$\phi_t: M \rightarrow M, \\ x \mapsto \phi_t(x).$$

with $\phi_0(x) = x$

$$\phi_s(\phi_t(x)) = \phi_{s+t}(x)$$

$$\phi(x, t) = x + \int_0^t \mathbf{v}(\phi(x, \tau)) \, d\tau$$

(on a general manifold M , integration is performed in coordinate charts)



The Flow / Flow Map of a Vector Field (1)



Flow of a *steady (time-independent)* vector field

- Map source position x “forward” ($t > 0$) or “backward” ($t < 0$) by time t

$$\boxed{\phi(x, t)}$$

$$\boxed{\phi_t(x)}$$

with $\phi_0(x) = x$

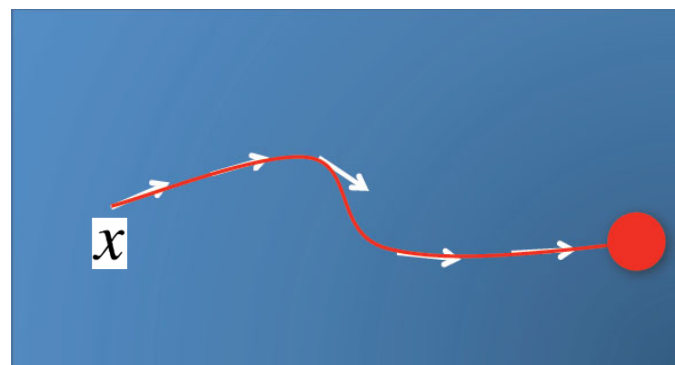
$$\phi_s(\phi_t(x)) = \phi_{s+t}(x)$$

$$\begin{aligned} \phi: M \times \mathbb{R} &\rightarrow M, & \phi_t: M &\rightarrow M, \\ (x, t) &\mapsto \phi(x, t). & x &\mapsto \phi_t(x). \end{aligned}$$

- Unsteady flow? Just fix arbitrary time T

$$\phi(x, t) = x + \int_0^t \mathbf{v}(\phi(x, \tau), T) d\tau$$

(on a general manifold M , integration is performed in coordinate charts)



The Flow / Flow Map of a Vector Field (1)



Flow of a *steady (time-independent)* vector field

- Map source position x “forward” ($t > 0$) or “backward” ($t < 0$) by time t

$$\boxed{\phi(x, t)}$$

$$\phi: M \times \mathbb{R} \rightarrow M, \\ (x, t) \mapsto \phi(x, t).$$

$$\boxed{\phi_t(x)}$$

$$\phi_t: M \rightarrow M, \\ x \mapsto \phi_t(x).$$

with

$$\phi_0(x) = x$$

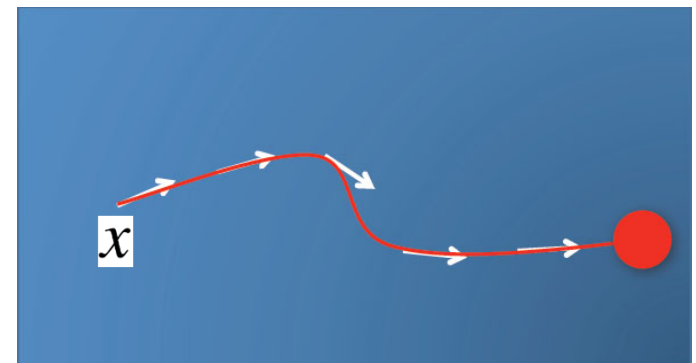
$$\phi_s(\phi_t(x)) = \phi_{s+t}(x)$$

Can write explicitly as function of independent variable t , with *position x fixed*

$$t \mapsto \phi(x, t)$$

$$t \mapsto \phi_t(x)$$

= stream line going through point x



The Flow / Flow Map of a Vector Field (2)



Flow of an *unsteady (time-dependent)* vector field

- Map source position x from time s to destination position at time t ($t < s$ is allowed: map forward or backward in time)

$$\boxed{\psi_{t,s}(x)}$$

with

$$\psi_{t,s}(x) = x + \int_s^t \mathbf{v}(\psi_{\tau,s}(x), \tau) d\tau$$

$$\psi_{s,s}(x) = x$$

$$\psi_{t,r}(\psi_{r,s}(x)) = \psi_{t,s}(x)$$

The Flow / Flow Map of a Vector Field (3)



Flow of an *unsteady (time-dependent)* vector field

- Map source position x from time s to destination position at time t ($t < s$ is allowed: map forward or backward in time)

$$\boxed{\psi_{t,s}(x)} \quad \psi_{t,s}(x) = x + \int_s^t \mathbf{v}(\psi_{\tau,s}(x), \tau) d\tau$$

Can write explicitly as function of t , *with s and x fixed*

$$t \mapsto \psi_{t,s}(x) \quad \rightarrow \text{path line}$$

Can write explicitly as function of s , *with t and x fixed*

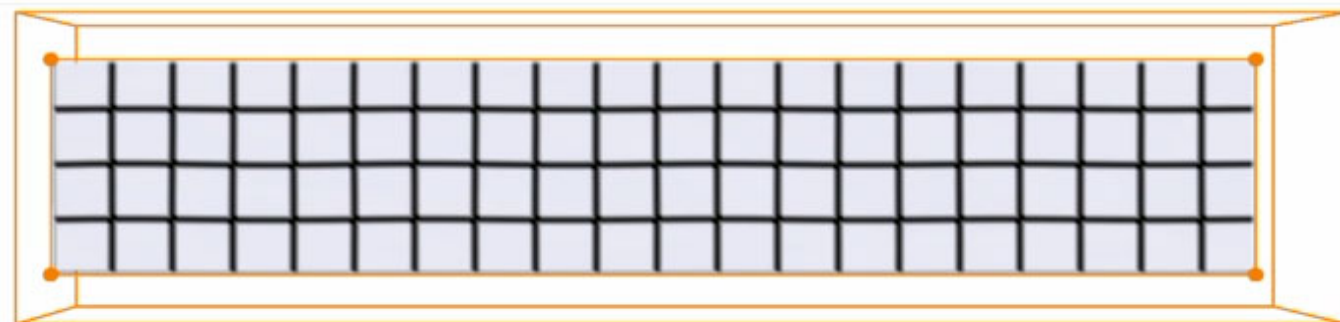
$$s \mapsto \psi_{t,s}(x) \quad \rightarrow \text{streak line}$$

$\psi_{t,s}(x)$ is also often written as **flow map** $\phi_t^\tau(x)$ (with $t:=s$ and either $\tau:=t$ or $\tau:=t-s$)

The Flow / Flow Map of a Vector Field (4)



Can map a whole set of points (or the entire domain) through the flow map (this map is a *diffeomorphism*): $t \mapsto \psi_{t,s}(U)$



U

$= \psi_{s,s}(U)$



$\psi_{t,s}(U)$

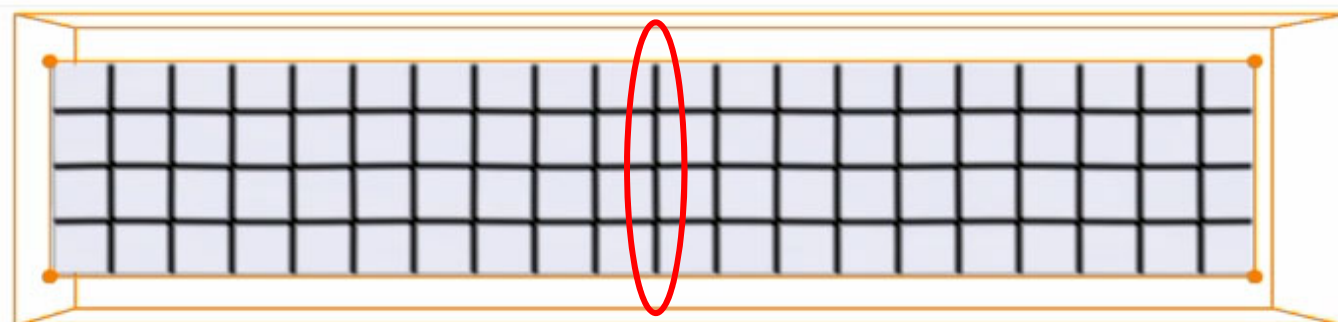
(this is a
time surface!)

The Flow / Flow Map of a Vector Field (5)

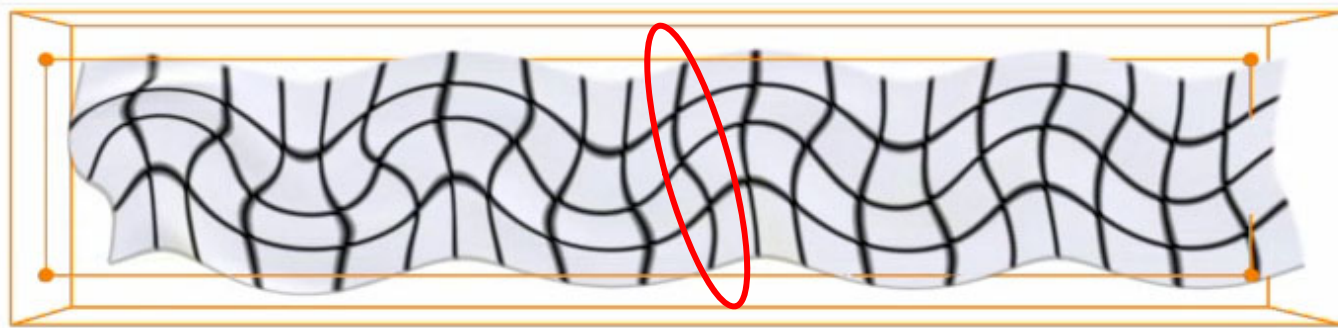


Time line: Map a whole curve from one fixed time (s) to another time (t)

$$t \mapsto \psi_{t,s}(c(\lambda))$$



$$c(\lambda) \\ = \psi_{s,s}(c(\lambda))$$



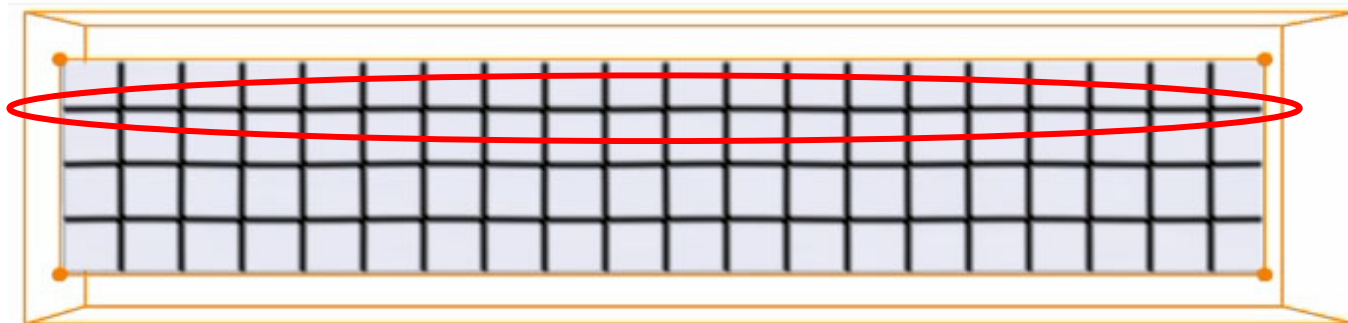
$$\psi_{t,s}(c(\lambda))$$

The Flow / Flow Map of a Vector Field (5)



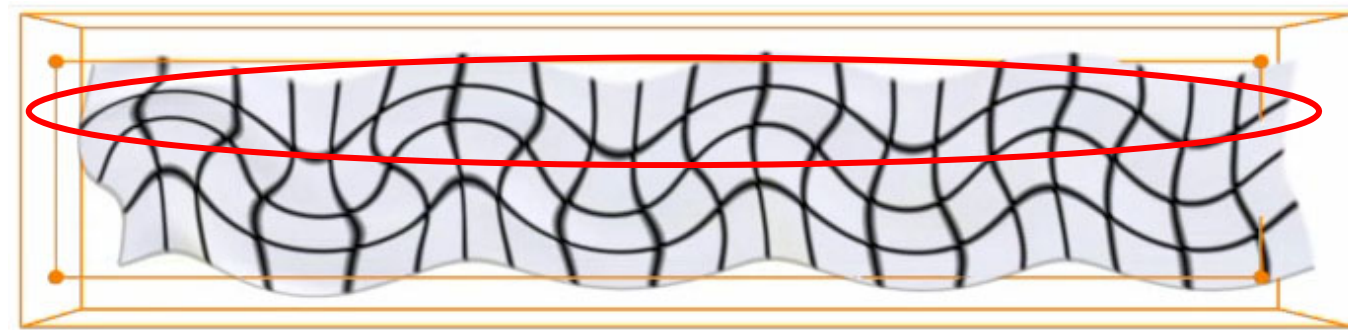
Time line: Map a whole curve from one fixed time (s) to another time (t)

$$t \mapsto \psi_{t,s}(c(\lambda))$$



$$c(\lambda)$$

$$= \psi_{s,s}(c(\lambda))$$



$$\psi_{t,s}(c(\lambda))$$

Streamline

- Curve parallel to the vector field in each point for a fixed time

Pathline

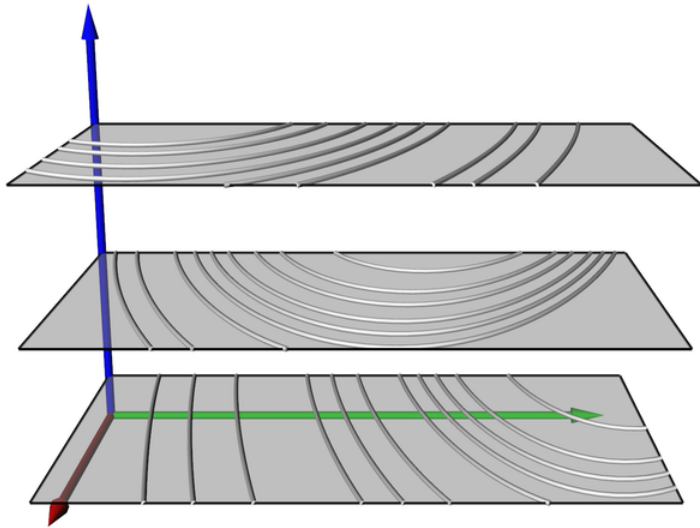
- Describes motion of a massless particle over time

Streakline

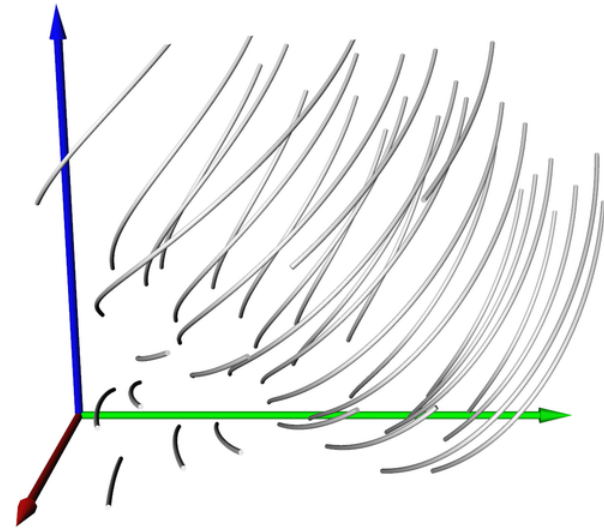
- Location of all particles released at a *fixed position* over time

Timeline

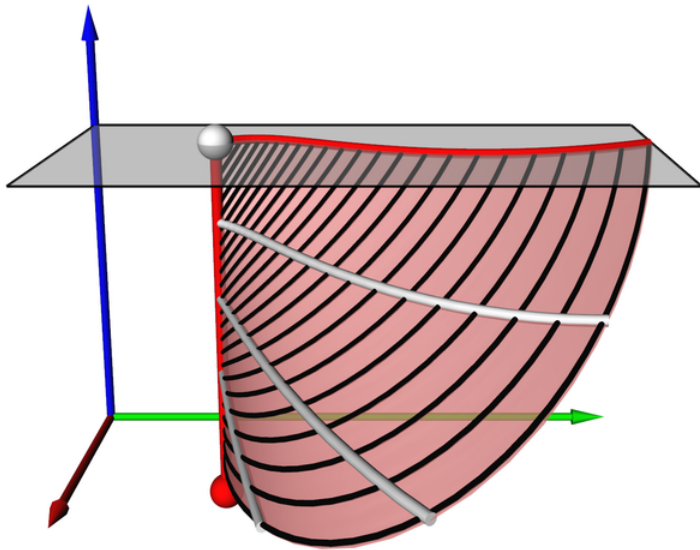
- Location of all particles released along a line at a *fixed time*



stream lines

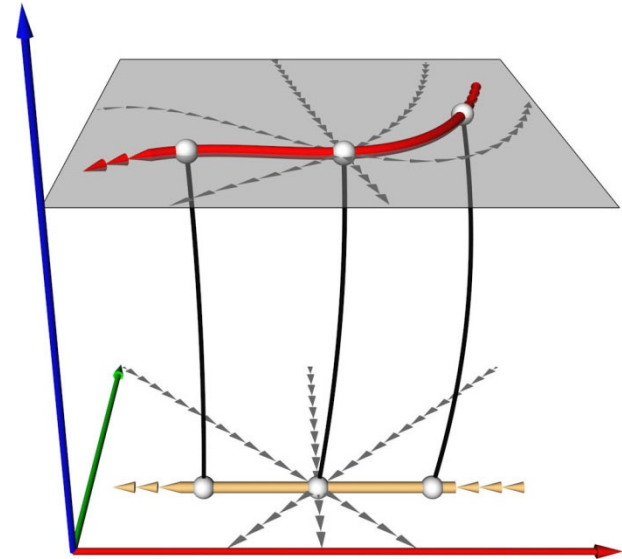


path lines



streak lines

time lines



Streamlines, pathlines, streaklines, timelines

Comparison of techniques:

(1) Pathlines:

- are physically meaningful
- allow comparison with experiment (observe marked particles)
- are well suited for dynamic visualization (of particles)

(2) Streamlines:

- are only geometrically, not physically meaningful
- are easiest to compute (no temporal interpolation, single IVP)
- are better suited for static visualization (prints)
- don't intersect (under reasonable assumptions)

Streamlines, pathlines, streaklines, timelines

(3) Streaklines:

- are physically meaningful
- allow comparison with experiment (dye injection)
- are well suited for static and dynamic visualization
- good choice for fast moving vortices
- can be approximated by set of disconnected particles

(4) Timelines:

- are physically meaningful
- are well suited for static and dynamic visualization
- can be approximated by set of disconnected particles

Thank you.

Thanks for material

- Helwig Hauser
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- Ronny Peikert
- Philipp Muigg
- Christof Rezk-Salama