

- Recall the solution to the IVP $y' = \lambda y$, $y(0) = y_0$ is $y(t) = e^{\lambda t}y_0$, with $y: \mathbb{R} \rightarrow \mathbb{R}$
- The solution to the system IVP $\mathbf{y}' = \mathbf{A}\mathbf{y}$, $\mathbf{y}(0) = \mathbf{y}_0$ is similar, $\mathbf{y}(t) = e^{\mathbf{A}t}\mathbf{y}_0$, with:
 - $\mathbf{y}_0 \in \mathbb{R}^n$, $\mathbf{y}: \mathbb{R} \rightarrow \mathbb{R}^n$
 - $\mathbf{A} \in \mathbb{R}^{n \times n}$
 - $e^{\mathbf{A}t} \in \mathbb{R}^{n \times n}$ is known as the matrix exponential defined by

$$e^{\mathbf{A}t} = \mathbf{I} + \mathbf{A}\frac{t}{1!} + \mathbf{A}^2\frac{t^2}{2!} + \dots + \mathbf{A}^k\frac{t^k}{k!} + \dots$$

- If \mathbf{A} is diagonalizable, i.e., the eigenvector matrix $\mathbf{X} \in \mathbb{R}^{n \times n}$

$$\mathbf{X} = (\mathbf{x}_1 \ \mathbf{x}_2 \ \dots \ \mathbf{x}_n), \mathbf{x}_k \in \mathbb{R}^n$$

arising in the eigenvalue problem $\mathbf{A}\mathbf{X} = \mathbf{X}\mathbf{\Lambda}$ (or $\mathbf{A}\mathbf{x}_k = \lambda_k\mathbf{x}_k$, $k = 1, \dots, n$) is invertible, then

$$e^{\mathbf{A}t} = \mathbf{X}e^{\mathbf{\Lambda}t}\mathbf{X}^{-1}, e^{\mathbf{\Lambda}t} = \text{diag}(e^{\lambda_1 t}, \dots, e^{\lambda_n t}), \mathbf{\Lambda} = \text{diag}(\lambda_1, \dots, \lambda_n).$$

- Diagonalizable matrices include:
 - matrices with distinct eigenvalues, $j \neq k \Rightarrow \lambda_j \neq \lambda_k$
 - symmetric matrices, $\mathbf{A}^T = \mathbf{A}$, the eigenvalues of which are purely real
 - skew-symmetric matrices, $\mathbf{A}^T = -\mathbf{A}$, the eigenvalues of which are purely imaginary
- Assume henceforth that \mathbf{A} is diagonalizable
- Rewrite solution as

$$\mathbf{y}(t) = e^{\mathbf{A}t}\mathbf{y}_0 = \mathbf{X}e^{\mathbf{\Lambda}t}\mathbf{X}^{-1}\mathbf{y}_0 = \mathbf{X}e^{\mathbf{\Lambda}t}\mathbf{c} = \begin{pmatrix} e^{\lambda_1 t}\mathbf{x}_1 & \dots & e^{\lambda_n t}\mathbf{x}_n \end{pmatrix} \mathbf{c} = c_1 e^{\lambda_1 t}\mathbf{x}_1 + \dots + c_n e^{\lambda_n t}\mathbf{x}_n$$

and observe that $\{\mathbf{y}_1, \dots, \mathbf{y}_n\} = \{e^{\lambda_1 t}\mathbf{x}_1, \dots, e^{\lambda_n t}\mathbf{x}_n\}$ are independent and form a basis for solutions of $\mathbf{y}' = \mathbf{A}\mathbf{y}$

- From $\mathbf{y}(t) = e^{\mathbf{A}t}\mathbf{y}_0 = \mathbf{X}e^{\mathbf{\Lambda}t}\mathbf{c} = c_1e^{\lambda_1 t}\mathbf{x}_1 + \dots + c_n e^{\lambda_n t}\mathbf{x}_n$ and $\lambda_k = \alpha_k + i\beta_k$ observe:
 - If for any k , $\text{Re}(\lambda_k) > 0$ and $c_k \neq 0$ then $\lim_{t \rightarrow \infty} \|\mathbf{y}(t)\| = \infty$
 - If for all k , $\text{Re}(\lambda_k) < 0$, then $\lim_{t \rightarrow \infty} \|\mathbf{y}(t)\| = 0$
- In general, many qualitative features of the solution are discernible from the eigenvalues/eigenvectors

Definition. The parametric curve $\{y_1(t), \dots, y_n(t)\}$ is a *trajectory* in the *phase plane* y_1, \dots, y_n . The set of all trajectories is the *phase portrait* of the ODE system.

Example 1 (Improper node). $\mathbf{y}' = \mathbf{A}\mathbf{y}$ with

$$\mathbf{A} = \begin{pmatrix} -3 & 1 \\ 1 & -3 \end{pmatrix}$$

```
In[64] := A={{-3,1},{1,-3}}; lambda=Eigenvalues[A]; Lambda=DiagonalMatrix[lambda];
X=Transpose[Eigenvectors[A]]; Y[t_] = X . Exp[Lambda t]; Map[MatrixForm[#]&,{Lambda,X,Exp[Lambda
t],Y[t]}]
```

$$\left\{ \begin{pmatrix} -4 & 0 \\ 0 & -2 \end{pmatrix}, \begin{pmatrix} -1 & 1 \\ 1 & 1 \end{pmatrix}, \begin{pmatrix} e^{-4t} & 1 \\ 1 & e^{-2t} \end{pmatrix}, \begin{pmatrix} 1 - e^{-4t} & -1 + e^{-2t} \\ 1 + e^{-4t} & 1 + e^{-2t} \end{pmatrix} \right\}$$

```
In[65] := A . X == X . Lambda
```

True

```
In[66] := y[t_,c_] := Y[t] . c; {y[t,{1,0}],y[t,{0,1}]}
```

$$\begin{pmatrix} 1 - e^{-4t} & 1 + e^{-4t} \\ -1 + e^{-2t} & 1 + e^{-2t} \end{pmatrix}$$

$$\mathbf{A} \in \mathbb{R}^{2 \times 2}, \mathbf{y}' = \mathbf{A}\mathbf{y}, \mathbf{y}(t) = e^{\mathbf{A}t}\mathbf{y}_0 = \mathbf{X}e^{\mathbf{\Lambda}t}\mathbf{X}^{-1}\mathbf{y}_0 = \mathbf{X}e^{\mathbf{\Lambda}t}\mathbf{c} = \begin{pmatrix} e^{\lambda_1 t} \mathbf{x}_1 & e^{\lambda_2 t} \mathbf{x}_2 \end{pmatrix} \mathbf{c} = c_1 e^{\lambda_1 t} \mathbf{x}_1 + c_2 e^{\lambda_2 t} \mathbf{x}_2$$

```
In[21]:= PhasePortrait = StreamPlot[A.{y1,y2},{y1,-1,1},{y2,-1,1},StreamColorFunction->Hue,
      FrameLabel->{Subscript["y",1],Subscript["y",2]},PlotLegends->Automatic];
      Export["/home/student/courses/MATH528/L10Fig01.pdf",PhasePortrait]
```

/home/student/courses/MATH528/L10Fig01.pdf

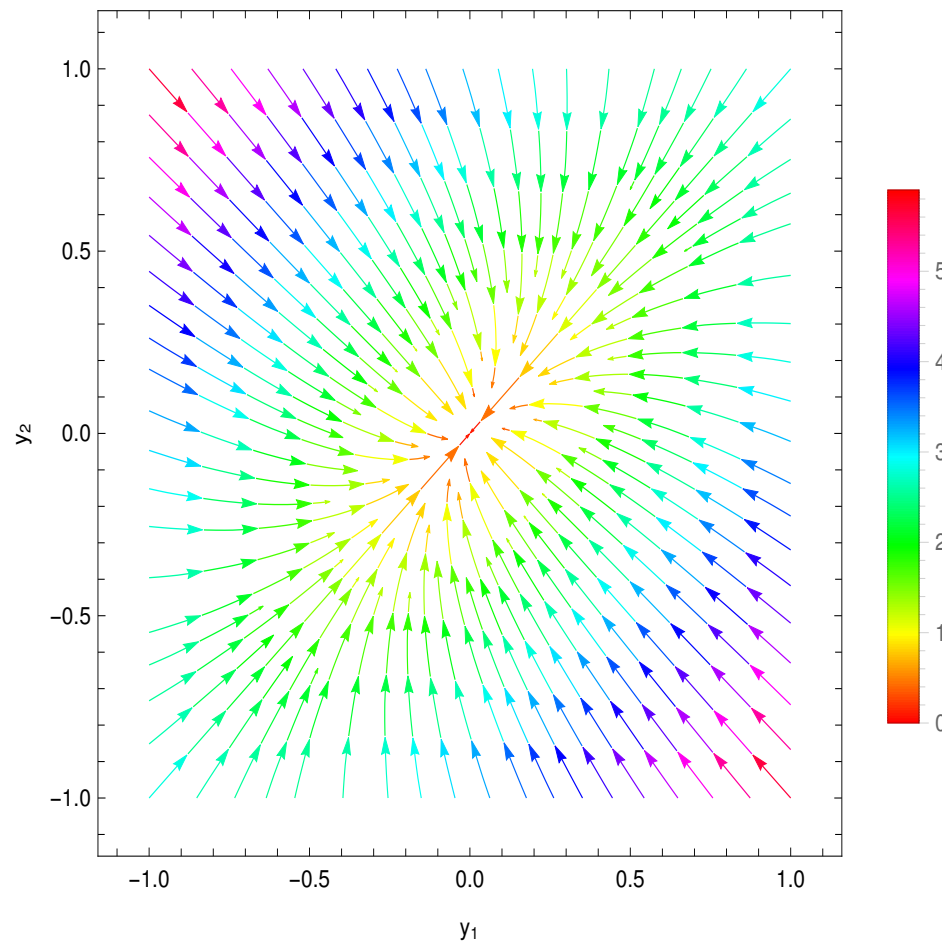


Figure 1. Phase portrait of an improper node

Example 2 (Proper node). $y' = Ay, A = I$.

```
In[22]:= PhasePortrait = StreamPlot[IdentityMatrix[2].{y1,y2},{y1,-1,1},{y2,-1,1},  
    StreamColorFunction->Hue,  
    FrameLabel->{Subscript["y",1],Subscript["y",2]},PlotLegends->Automatic];  
Export["/home/student/courses/MATH528/L10Fig02.pdf",PhasePortrait]
```

/home/student/courses/MATH528/L10Fig02.pdf

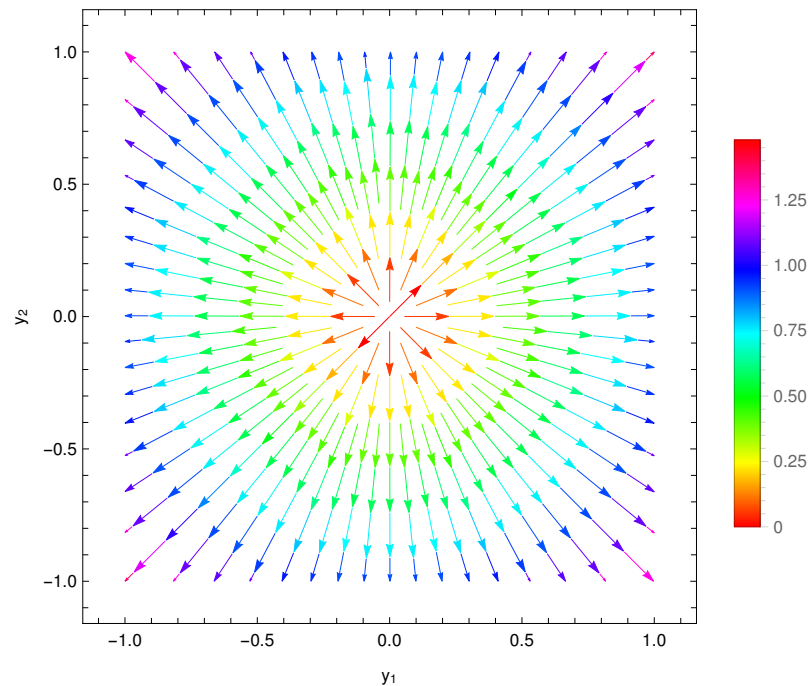


Figure 2. Phase portrait of a proper node

Example 3 (Saddle node). $y' = Ay, A = \text{diag}(1, -1)$.

```
In[23]:= PhasePortrait = StreamPlot[DiagonalMatrix[{1, -1}].{y1, y2}, {y1, -1, 1}, {y2, -1, 1},
    StreamColorFunction->Hue,
    FrameLabel->{Subscript["y", 1], Subscript["y", 2]}, PlotLegends->Automatic];
Export["/home/student/courses/MATH528/L10Fig03.pdf", PhasePortrait]
```

/home/student/courses/MATH528/L10Fig03.pdf

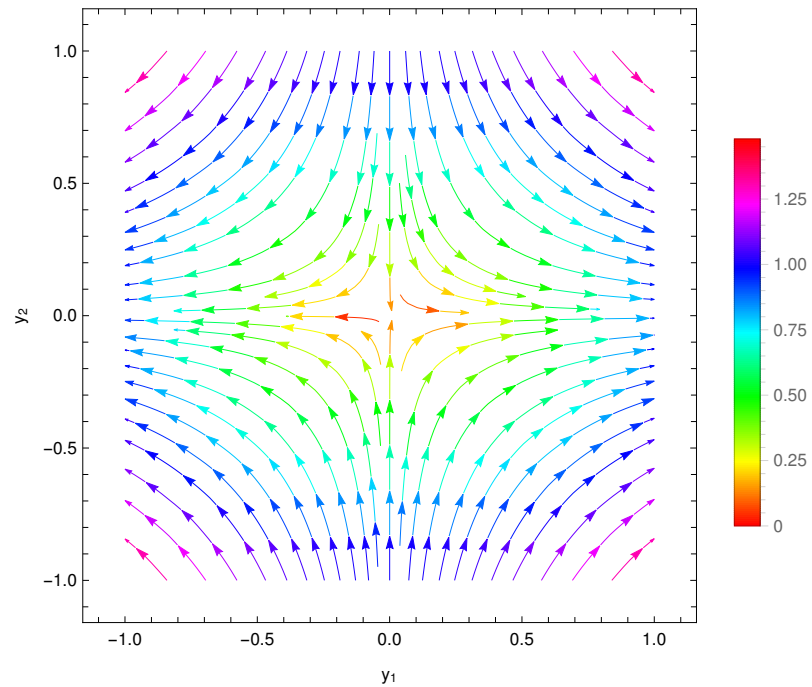


Figure 3. Phase portrait of a saddle node

Example 4 (Center node). $y' = Ay, A = \begin{pmatrix} 0 & 1 \\ -4 & 0 \end{pmatrix}$.

```
In[25]:= PhasePortrait = StreamPlot[{{0,1},{-4,0}}.{y1,y2},{y1,-1,1},{y2,-1,1},
    StreamColorFunction->Hue,
    FrameLabel->{Subscript["y",1],Subscript["y",2]},PlotLegends->Automatic];
Export["/home/student/courses/MATH528/L10Fig04.pdf",PhasePortrait]
```

/home/student/courses/MATH528/L10Fig04.pdf

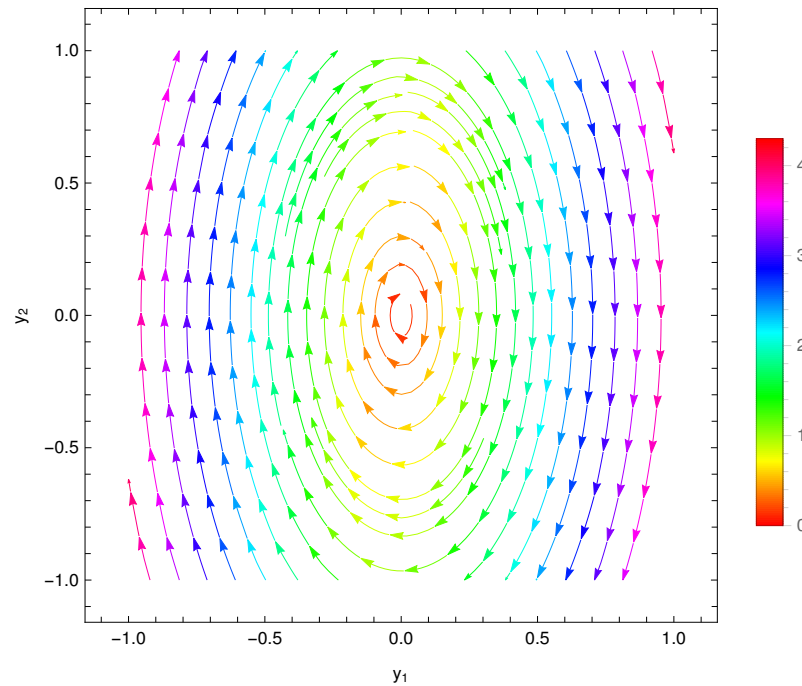


Figure 4. Phase portrait of a center node

Example 5 (Spiral node). $y' = Ay, A = \begin{pmatrix} -1 & 1 \\ -1 & 1 \end{pmatrix}$.

```
In[26]:= PhasePortrait = StreamPlot[{{-1,1},{-1,-1}}.{y1,y2},{y1,-1,1},{y2,-1,1},
    StreamColorFunction->Hue,
    FrameLabel->{Subscript["y",1],Subscript["y",2]},PlotLegends->Automatic];
Export["/home/student/courses/MATH528/L10Fig05.pdf",PhasePortrait]
```

/home/student/courses/MATH528/L10Fig05.pdf

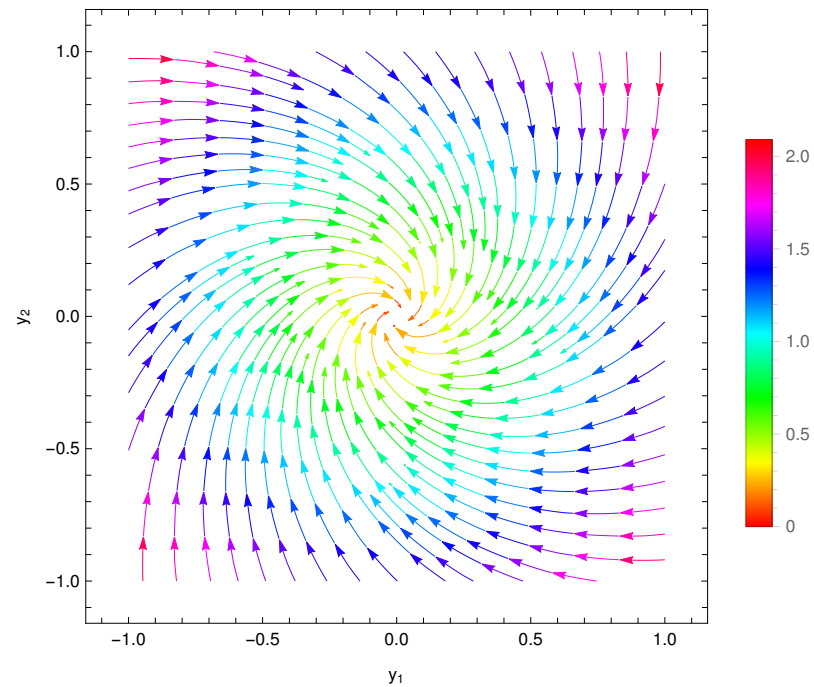


Figure 5. Phase portrait of a spiral node

Example 6 (Degenerate node). $y' = Ay, A = \begin{pmatrix} 4 & 1 \\ -1 & 2 \end{pmatrix}$.

```
In[27]:= PhasePortrait = StreamPlot[{{4,1},{-1,2}}.{y1,y2},{y1,-1,1},{y2,-1,1},
    StreamColorFunction->Hue,
    FrameLabel->{Subscript["y",1],Subscript["y",2]},PlotLegends->Automatic];
Export["/home/student/courses/MATH528/L10Fig06.pdf",PhasePortrait]
```

/home/student/courses/MATH528/L10Fig06.pdf

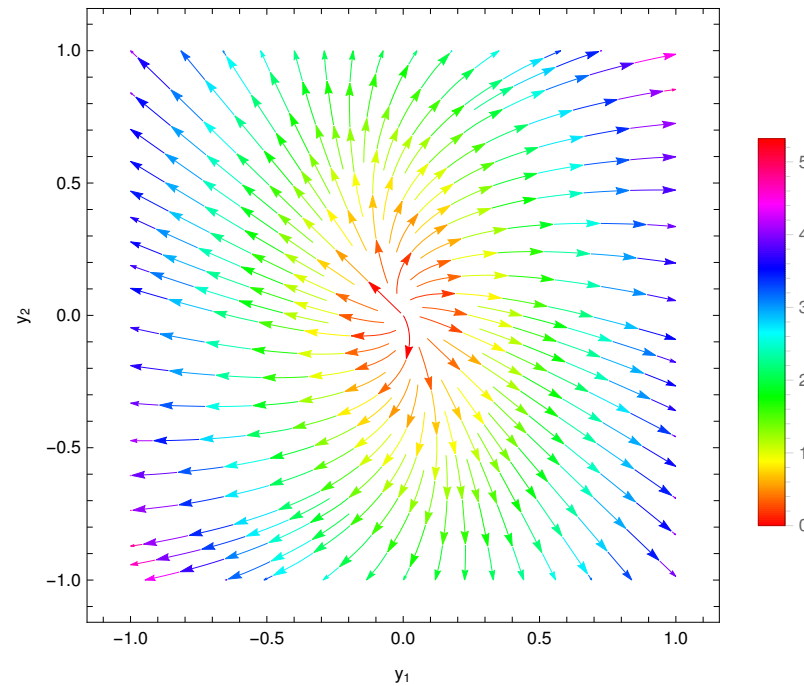


Figure 6. Phase portrait of a center node

Definition. Roots of $f(\mathbf{y}) = \mathbf{0}$ are called *critical points* of the ODE system $\mathbf{y}' = f(\mathbf{y})$.

Definition. A critical point \mathbf{y}^* is *stable* if all trajectories near \mathbf{y}^* remain close to \mathbf{y}^* as time increases.

Definition. The critical point \mathbf{y}^* of ODE system $\mathbf{y}' = f(\mathbf{y})$ is *stable* if $\forall \varepsilon > 0, \exists \delta_\varepsilon$ such that if $\exists t_0$ such that $\|\mathbf{y}(t_0) - \mathbf{y}^*\| < \delta_\varepsilon$, with $\mathbf{y}(t)$ a solution of the ODE system, then $\forall t \geq t_0, \|\mathbf{y}(t) - \mathbf{y}^*\| < \varepsilon$. Otherwise, the critical point is *unstable*. If, in addition, $\lim_{t \rightarrow \infty} \mathbf{y}(t) = \mathbf{y}^*$ for all solutions, the critical point is *stable and attractive*.

Algorithm ODE system qualitative analysis

1. For ODE $\mathbf{y}' = f(\mathbf{y})$, find roots \mathbf{y}^* , $f(\mathbf{y}^*) = \mathbf{0}$.
2. Taylor series expand $f(\mathbf{y}) = f(\mathbf{y}^*) + \mathbf{A}(\mathbf{y} - \mathbf{y}^*) + \mathcal{O}(\|\mathbf{y} - \mathbf{y}^*\|^2)$
3. Determine type of critical point from eigenvalues of the Jacobian

$$\mathbf{A} = \mathbf{J}(f(\mathbf{y}^*)) = \frac{\partial f}{\partial \mathbf{y}}(\mathbf{y}^*).$$