## ECE 204 Numerical methods

## Newton's'method in $n$ dimensions

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## Introduction

- In this topic, we will
- Derive Newton's method in $n$ dimensions
- Observe the formula is analogous to the one we've seen
- Look at an example
- Observe the rate of convergence is still $\mathrm{O}\left(h^{2}\right)$


## Newton's method

- Recall Newton's method:
- Find the root of the tangent line at $\left(x_{k}, f\left(x_{k}\right)\right)$



## Generalization of Newton's method

- Recall the Taylor series for a real-valued function of a vector variable

$$
\begin{aligned}
f(\mathbf{u}) & \approx f\left(\mathbf{u}_{0}\right)+\vec{\nabla} f\left(\mathbf{u}_{0}\right) \cdot\left(\mathbf{u}-\mathbf{u}_{0}\right) \\
& =f\left(\mathbf{u}_{0}\right)+\left(\frac{\partial}{\partial u_{1}} f\left(\mathbf{u}_{0}\right) \cdots \frac{\partial}{\partial u_{n}} f\left(\mathbf{u}_{0}\right)\right)\left(\mathbf{u}-\mathbf{u}_{0}\right)
\end{aligned}
$$

- Suppose we have $n$ such functions, so for each we have:

$$
f_{k}(\mathbf{u}) \approx f_{k}\left(\mathbf{u}_{0}\right)+\left(\begin{array}{lll}
\frac{\partial}{\partial u_{1}} f_{k}\left(\mathbf{u}_{0}\right) & \cdots & \left.\frac{\partial}{\partial u_{n}} f_{k}\left(\mathbf{u}_{0}\right)\right)\left(\mathbf{u}-\mathbf{u}_{0}\right)
\end{array}\right.
$$

## Generalization of Newton's method

- Thus, we have:



## Generalization of Newton's method

$$
\mathbf{f}\left(\mathbf{u}_{0}\right)
$$

- Thus, we have:
$\mathbf{f}(\mathbf{u}) \longrightarrow\left(\begin{array}{c}f_{1}(\mathbf{u}) \\ \vdots \\ f_{n}(\mathbf{u})\end{array}\right) \approx\left(\begin{array}{c}f_{1}\left(\mathbf{u}_{0}\right) \\ \vdots \\ f_{n}\left(\mathbf{u}_{0}\right)\end{array}\right)+\left(\begin{array}{ccc}\frac{\partial}{\partial u_{1}} f_{k}\left(\mathbf{u}_{0}\right) & \cdots & \frac{\partial}{\partial u_{n}} f_{1}\left(\mathbf{u}_{0}\right) \\ \vdots & \ddots & \vdots \\ \frac{\partial}{\partial u_{1}} f_{n}\left(\mathbf{u}_{0}\right) & \cdots & \frac{\partial}{\partial u_{n}} f_{n}\left(\mathbf{u}_{0}\right)\end{array}\right)\left(\mathbf{u}-\mathbf{u}_{0}\right)$

This is the Jacobian evaluated at $\mathbf{u}_{0}$

$$
J(\mathbf{f})(\mathbf{u})=\left(\begin{array}{ccc}
\frac{\partial}{\partial u_{1}} f_{k}(\mathbf{u}) & \cdots & \frac{\partial}{\partial u_{n}} f_{1}(\mathbf{u}) \\
\vdots & \ddots & \vdots \\
\frac{\partial}{\partial u_{1}} f_{n}(\mathbf{u}) & \cdots & \frac{\partial}{\partial u_{n}} f_{n}(\mathbf{u})
\end{array}\right)
$$

## Generalization of Newton's method

- Thus, we have $n$ tangent ( $n-1$ )-dimensional hyperplanes at $\mathbf{u}_{0}$ :

$$
\mathbf{f}(\mathbf{u}) \approx \mathbf{f}\left(\mathbf{u}_{0}\right)+J(\mathbf{f})\left(\mathbf{u}_{0}\right)\left(\mathbf{u}-\mathbf{u}_{0}\right)
$$

- A root of these hyperplanes may be found by equating this to the zero vector:

$$
\begin{aligned}
\mathbf{0} & =\mathbf{f}\left(\mathbf{u}_{0}\right)+J(\mathbf{f})\left(\mathbf{u}_{0}\right)\left(\mathbf{u}-\mathbf{u}_{0}\right) \\
J(\mathbf{f})\left(\mathbf{u}_{0}\right)\left(\mathbf{u}-\mathbf{u}_{0}\right) & =-\mathbf{f}\left(\mathbf{u}_{0}\right)
\end{aligned}
$$

- This is a system of $n$ linear equations in $n$ unknowns
- Let $\Delta \mathbf{u}_{0}=\mathbf{u}-\mathbf{u}_{0}$, so we are solving

$$
J(\mathbf{f})\left(\mathbf{u}_{0}\right) \Delta \mathbf{u}_{0}=-\mathbf{f}\left(\mathbf{u}_{0}\right)
$$

- Having found $\Delta \mathbf{u}_{0}$, we now assign $\mathbf{u}_{1} \leftarrow \mathbf{u}_{0}+\Delta \mathbf{u}_{0}$
- We can now repeat this until $\left\|\mathbf{u}_{k+1}-\mathbf{u}_{k}\right\|_{2}<\varepsilon_{\text {step }}$ and $\left\|\mathbf{f}\left(\mathbf{u}_{k+1}\right)\right\|_{2}<\varepsilon_{a b} \underbrace{0}_{0}$


## Generalization of Newton's method

- You may be wondering, how are these related?

$$
\begin{gathered}
x_{k+1}=x_{k}-\frac{f\left(x_{k}\right)}{f^{(1)}\left(x_{k}\right)} \quad \begin{array}{c}
\text { Solve } J(\mathbf{f})\left(\mathbf{u}_{k}\right) \Delta \mathbf{u}_{k}=-\mathbf{f}\left(\mathbf{u}_{k}\right) \text { for } \Delta \mathbf{u}_{k}, \\
\text { and assign } \mathbf{u}_{k+1} \leftarrow \mathbf{u}_{k}+\Delta \mathbf{u}_{k}
\end{array} \\
0=f\left(x_{k}\right)+f^{(1)}\left(x_{k}\right)\left(x_{k+1}-x_{k}\right) \\
f^{(1)}\left(x_{k}\right)\left(x_{k+1}-x_{k}\right)=-f\left(x_{k}\right) \\
f^{(1)}\left(x_{k}\right) \Delta x_{k}=-f\left(x_{k}\right) \quad \Delta x_{k}=-\frac{f\left(x_{k}\right)}{f^{(1)}\left(x_{k}\right)} \\
x_{k+1}=x_{k}+\Delta x_{k} \\
\text { That is, } x_{k+1}=x_{k}-\frac{f\left(x_{k}\right)}{f^{(1)}\left(x_{k}\right)}
\end{gathered}
$$

## Example

- Suppose we have the following:

$$
\mathbf{f}(\mathbf{u})=\binom{u_{1}^{2}+2 u_{2}-1}{u_{2}^{2}+3 u_{1}-2}
$$



## Example

- First, we calculate the Jacobian

$$
\begin{array}{r}
\mathbf{f}(\mathbf{u})=\binom{u_{1}^{2}+2 u_{2}-1}{u_{2}^{2}+3 u_{1}-2} \\
J(\mathbf{f})(\mathbf{u})=\left(\begin{array}{cc}
2 u_{1} & 2 \\
3 & 2 u_{2}
\end{array}\right)
\end{array}
$$



## Example

- Note that $\mathbf{f}\binom{0.75}{0.5}=\binom{0.5625}{0.5}$
- Thus, $\mathbf{u}_{0}=\binom{0.75}{0.5}$ is sort-of close to a root $\begin{gathered}f_{1}\left(\mathbf{u}_{0}\right)=0.5625\end{gathered}$

$11 \int_{0}^{0}$


## Example

- We can find tangent planes at each of these points
$J(\mathbf{f})\left(\mathbf{u}_{0}\right) \Delta \mathbf{u}_{0}=-\mathbf{f}\left(\mathbf{u}_{0}\right) \quad\left(\begin{array}{ll}1.5 & 2 \\ 3 & 1\end{array}\right) \Delta \mathbf{u}_{0}=\binom{-0.5625}{-0.5} \quad \Delta \mathbf{u}_{0}=\binom{-0.0972222}{-0.2083333}$

$\mathbf{u}_{0}=\binom{0.75}{0.5}$

$$
12 \int_{0}^{0}
$$

## Example

- We solved for $\Delta \mathbf{u}_{0}$ and so $\mathbf{u}_{1} \leftarrow \mathbf{u}_{0}+\Delta \mathbf{u}_{0}$

$$
\text { - Thus, } \mathbf{u}_{1}=\binom{0.6527778}{0.2916667} \text { and } \mathbf{f}\left(\mathbf{u}_{1}\right)=\binom{0.009452}{0.04340}
$$



## Example

- Here is a sequence of iterations:

$$
\begin{array}{ll}
\mathbf{u}_{0}=\binom{0.75}{0.5} & \mathbf{f}\left(\mathbf{u}_{0}\right)=\binom{0.5625}{0.5} \\
\mathbf{u}_{1}=\binom{0.6527777777777778}{0.2916666666666667} & \mathbf{f}\left(\mathbf{u}_{1}\right)=\binom{0.009452}{0.04340} \\
\mathbf{u}_{2}=\binom{0.6372594147395296}{0.2970706289586095} & \mathbf{f}\left(\mathbf{u}_{2}\right)=\binom{0.0002408}{0.00002920} \\
\mathbf{u}_{3}=\binom{0.6372755656421493}{0.2969399268481651} & \mathbf{f}\left(\mathbf{u}_{3}\right)=\binom{0.0000000002609}{0.00000001708}
\end{array}
$$

## Example

- The actual root is closer to

$$
\mathbf{u}=\binom{0.6372755591552685}{0.2969399308516699} \quad \mathbf{u}_{3}=\binom{0.6372755656421493}{0.2969399268481651}
$$



## Summary

- Following this topic, you now
- Understand the generalization of Newton's method
- In two dimensions, we have two expressions in two variables
- Given an initial approximation, find two tangent planes, and find the simultaneous root of those tangent planes
- Know this generalizes to $n$ dimensions
- Find the tangent hyper-planes and find the root of the tangent hyper-planes
- Are aware that the convergence is still $\mathrm{O}\left(h^{2}\right)$


## References

[1] https://en.wikipedia.org/wiki/Newton\'s_method

## Acknowledgments

Jeffrey Cornelis for noting I left out the most significant digit in the leading entry of the approximation

$$
\mathbf{u}_{1}=\binom{0.6527778}{0.2916667}
$$

## Colophon

These slides were prepared using the Cambria typeface. Mathematical equations use Times New Roman, and source code is presented using Consolas. Mathematical equations are prepared in MathType by Design Science, Inc.
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The photographs of flowers and a monarch butter appearing on the title slide and accenting the top of each other slide were taken at the Royal Botanical Gardens in October of 2017 by Douglas Wilhelm Harder. Please see https://www.rbg.ca/
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