

# Hierarchical-block conditioning approximations for high-dimensional multivariate normal probabilities

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# Introduction

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- The computation of the multivariate normal (MVN) probability

$$\Phi_n(\mathbf{a}, \mathbf{b}; 0, \Sigma) = \int_a^b \frac{1}{\sqrt{(2\pi)^n |\Sigma|}} \exp\left(-\frac{1}{2} \mathbf{x}^T \Sigma^{-1} \mathbf{x}\right) d\mathbf{x}, \quad (1)$$

where  $\mathbf{a}$  and  $\mathbf{b}$  are integration limits, the mean vector  $\mu$  is assumed to be 0,  $\Sigma$  is a positive-definite covariance matrix, is required for a variety of applications.

- Various methods to compute MVN probability are suggested such as Richtmyer Quasi-Monte Carlo (QMC) (?)
- However, In high-dimensional settings (large  $n$ ), it is hard to compute (1) directly.
- We review new approaches proposed by Cao et al. (2019) to approximate high-dimensional multivariate normal probability (1) using the hierarchical matrix  $\mathcal{H}$  (Hackbusch, 2015) for the covariance matrix  $\Sigma$ .

The methods are based on

1. the bivariate conditioning method (Trinh and Genz, 2015) and
2. the hierarchical QMC method (Genton et al., 2018).

# Elements

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The theme provides sensible defaults to  
`\emph{emphasize}` text, `\alert{accent}` parts  
or show `\textbf{bold}` results.

becomes

The theme provides sensible defaults to *emphasize* text, **accent** parts or  
show **bold** results.

# Font feature test

- Regular
- *Italic*
- SMALLCAPS
- **Bold**
- **Bold Italic**
- **Bold SmallCaps**
- Monospace
- *Monospace Italic*
- Monospace Bold
- *Monospace Bold Italic*



## Items

- Milk
- Eggs
- Potatos

## Enumerations

1. First,
2. Second and
3. Last.

## Descriptions

**PowerPoint** Meeh.

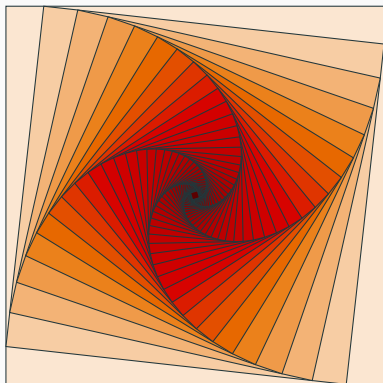
**Beamer** Yeeeha.

- This is important

- This is important
- Now this

- This is important
- Now this
- And now this

- This is really important
- Now this
- And now this



**Figure 1:** Rotated square from texample.net.

**Table 1:** Largest cities in the world (source: Wikipedia)

City	Population
Mexico City	20,116,842
Shanghai	19,210,000
Peking	15,796,450
Istanbul	14,160,467

# Blocks

Three different block environments are pre-defined and may be styled with an optional background color.

## **Default**

Block content.

## **Alert**

Block content.

## **Example**

Block content.

## **Default**

Block content.

## **Alert**

Block content.

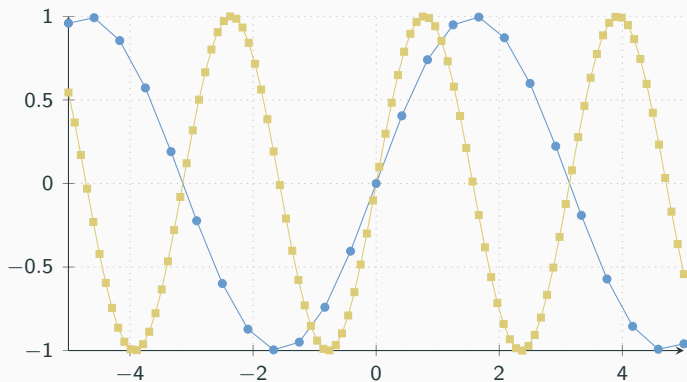
## **Example**

Block content.

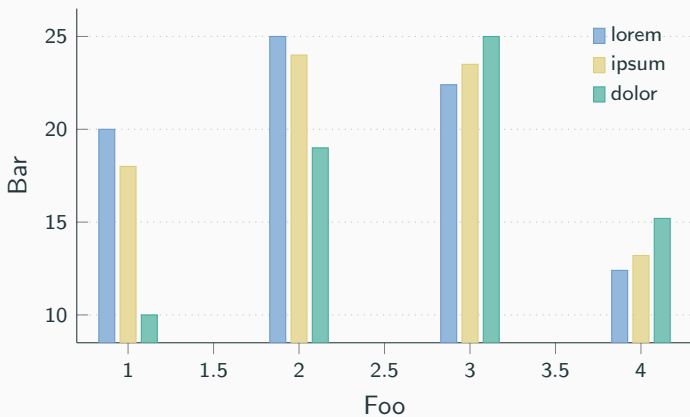


$$e = \lim_{n \rightarrow \infty} \left(1 + \frac{1}{n}\right)^n$$

# Line plots



# Bar charts



*Veni, Vidi, Vici*

**metropolis** defines a custom beamer template to add a text to the footer. It can be set via

```
\setbeamertemplate{frame footer}{My custom footer}
```

Some references to showcase `[allowframebreaks]` ?????

## Conclusion

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Get the source of this theme and the demo presentation from

`github.com/matze/mtheme`

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**Questions?**

# Backup slides

Sometimes, it is useful to add slides at the end of your presentation to refer to during audience questions.

The best way to do this is to include the `appendixnumberbeamer` package in your preamble and call `\appendix` before your backup slides.

**metropolis** will automatically turn off slide numbering and progress bars for slides in the appendix.

### References

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- Cao, J., Genton, M. G., Keyes, D. E., and Turkiyyah, G. M. (2019). Hierarchical-block conditioning approximations for high-dimensional multivariate normal probabilities. *Statistics and Computing*, 29(3):585–598.
- Genton, M. G., Keyes, D. E., and Turkiyyah, G. (2018). Hierarchical decompositions for the computation of high-dimensional multivariate normal probabilities. *Journal of Computational and Graphical Statistics*, 27(2):268–277.
- Hackbusch, W. (2015). *Hierarchical matrices: algorithms and analysis*, volume 49. Springer.

Trinh, G. and Genz, A. (2015). Bivariate conditioning approximations for multivariate normal probabilities. *Statistics and Computing*, 25(5):989–996.