

Learning Compilation Results Of Linux Kernel Configurations

Encadrants

Mathieu Acher (Contact)

Mail : mathieu.acher@irisa.fr, <http://www.mathieuacher.com/>

Jean-Marc Jézéquel (Contact)

Mail : jean-marc.jezequel@irisa.fr

Structure d'accueil

Ville : Rennes

Désignation de l'établissement : Laboratoire

Nom de l'établissement : Inria / IRISA

Équipe : DiverSE

Mots-clés :

- Artificial intelligence
- Machine Learning
- Software engineering
- Operating system

Description :

With more than 15000 configuration options, Linux is certainly the most complex software configurable system ever developed. The Linux kernel offers an almost infinite number of different kernels for fitting the needs of users. Among the resulting many billions possible configurations, some of them may not compile, produce large kernels (around 2 gigabytes!) or small kernels (less than 1 megabyte!), etc.

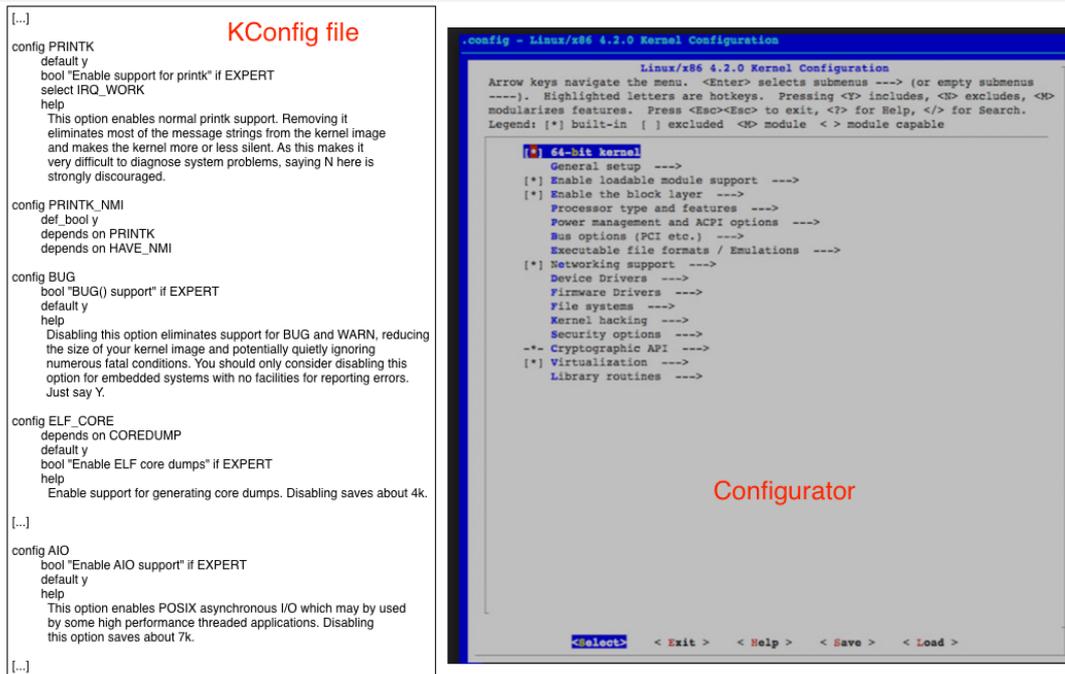


Figure 1 In the left-hand side, an KConfig file excerpt of the Linux kernel. At the right hand-side, the configurator of Linux. Challenge: how to select options that will lead to a small kernel (eg less than 1 megabyte) ?

The goal of this project is to predict, in advance, what are the compilation results of a given Linux kernel configuration. We have already gathered the largest dataset with compilation results of 120000 configurations.

The intern will explore machine learning-based techniques (decision trees, random forests, boosting trees, neural networks, etc.) to effectively learn compilation results. We are targetting different use cases :

- Understanding important options (eg that have a strong impact on the size of the kernel)
- Predicting with a high accuracy some outcomes of the compilation process (eg does this configuration compile ?)
- Narrowing the space of configurations to reach certain objectives (eg kernel less than 1Mo)

Large experiments as well as the development of innovative statistical machine learning techniques are needed to answer these questions. Another possible, related direction is to perform contributions for computing at very large scale Linux kernel configurations.

Skills: We plan to use either R or scikit-learn for performing machine learning experiments. It will be an opportunity to use and apply popular tools in machine learning.

This work is part of an ANR research project called VaryVary. There are 3 open positions (1 Engineer, 1 PhD student, 1 Postdoc)

<https://varyvary.github.io/>

It is a good starting point for discovering the project and applying to these positions.

Bibliographie :

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