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# Whats this all about?

- The world is full of **SILENTLY OVERFLOWING** integers
- In my last lightning talk, I've shown that **SILENT INTEGER OVERFLOWS** can happen in **numpy** and **scipy**
- In this, we'll have a look at some other commonly used libraries

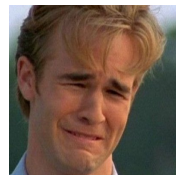
# Bring on the failing Code

- *(Note to self: open the Jupyter Notebook now)*

# OK, that looks **fairly bad**

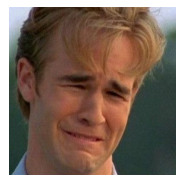
- Is this a *python-specific* thing?
- Not really, I've tried the same with a few other languages:

- Octave



*It fails*

- Matlab



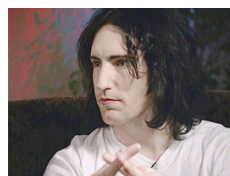
*It fails*

- Julia



*It works, no silent integer overflows in Julia!!!*

- R



*(I don't know, I didn't try)*

# So, why??

- There are two important questions to ask
  - *Why does it **FAIL** for **integers**?*
  - *Why does it **NOT** fail for **floats**?*
- Why it doesn't fail for floats
  - Floating point errors are checked at the **hardware level**, i.e. the FPU sets a flag whenever something goes wrong. Errors are propagated upwards. This is very efficient.
- Why it fails for integers
  - Integer overflows are **NOT** checked at the hardware level. Any client code needs to **implement its own checks**. In the case of **numpy**, the issue is known since at least 2009, but there was a deliberate decision to **NOT** check integer overflows in arrays for performance reasons (see [here](#) and [here](#)).