



Initial Situation:

- fin_15_2 prepared code to **optimize pension decision** in the Liechtenstein pension system
- Problem:
 - Computationally quite demanding
 - Not easily accessible for everyone

Overall Project Goals:

1. Make model available to everyone
2. Better understanding of model and implications

Step 1:

1. Improve Code
 - a. Avoid infeasible in- & outputs* ✓
 - b. Speed up optimization ✓
 - c. Make code publicly available ✓
 - d. Document code/law/tax scheme ✓

R-package on [github](#): `install_github("sstoeckl/pensionfinanceLi")`
Detailed documentation: `vignette("model")`
*Moral Hazard: Borrow pension from the bank without intention of payback

Step 2:

2. Determine grid of feasible input parameters (3'110'400) ✓
3. Run optimization for every parameter combination
 - a. Massive parallelization necessary ✓
 - b. To avoid data loss in case of crash & provide easy access data save to high-performance database (Amazon RDS) ✓
 - c. Rent clusters & run code (Amazon AWS) ✓

To better understand the model and its drivers | Computationally demanding (approx. 10 minutes per optimization) | Currently at 800'000 | Expected to finish in Dec 2021 (necessary cost reduction)

Step 3:

4. Develop heuristic model to predict (near-optimal) pension decisions in real-time through Machine Learning
 - a. Linear & nonlinear models (k-nearest neighbor, random forest) using *scikit-learn* ✓
 - b. Random Forest shows considerable forecasting power ✓
5. Use models to gain better understanding of relation between in- and output variables (preliminary) ✓

Strong non-linearities driving results ($c, \alpha, w_3 \sim ra + c_age$)

Step 4:

6. Make results available to the public through online app
 - a. Use *plumber/swagger* to create api that weekly updates Machine Learning models ✓
 - b. Create *Shiny App* that connects to api and allows for real-time near-optimal pension decisions <https://apps.resqfin.com/pfli> ✓

	cons	alpha	w_stocks	w_bonds	w_realt
Linear Model	0.58	0.93	0.97	0.02	0.46
K-Nearest Neighbors	0.75	0.93	0.85	0.00	0.65
Random Forest	0.95	0.92	0.85	0.04	0.56