

Brussels Data Science Meetup @VUB - April 23, 2015

Data4Good proof-of-concept: a Micro-Finance Loan Funding Predictor for kiva.org

Frederik Durant



Frederik Durant

- MA, MSc, MBA
- 20 years of software engineering, research, management and consulting positions in Speech and Natural Language Processing
- Currently R&D Manager of Natural Language Understanding at Nuance Communications
- About to finish a self-sponsored 4-month sabbatical
- Just returned from 12-week Data Science **Bootcamp** in New York City
- Final project: loan funding predictor for KIVA









VH a global micro-finance platform

- Non-profit organization based in San Francisco, CA
- Mission: connect people through lending to alleviate poverty
- Worldwide network of micro-finance institutions ("field partners")
- Lets individuals lend as little as \$25 to help create opportunity

Since Kiva was founded in 2005:

1,283,922 Kiva lenders

\$698,814,300 in loans

98.72% repayment rate

We work with:

296 Field Partners

450 volunteers around the world

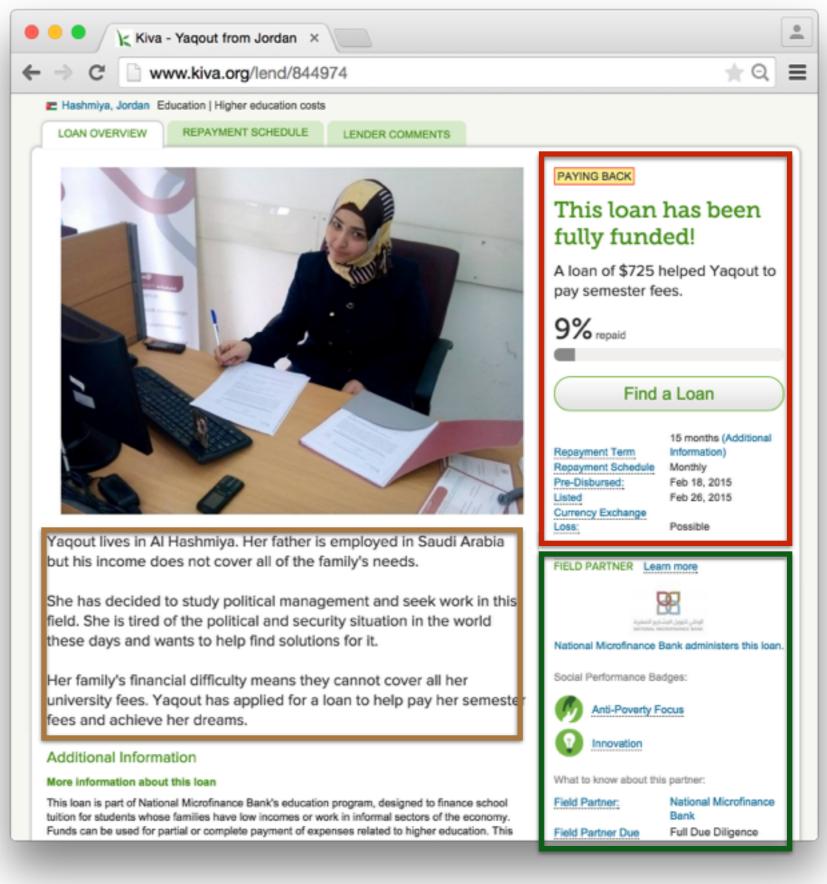
86 different countries

Source: http://www.kiva.org/about (April 23, 2015)



Meet Yaqout from Jordan

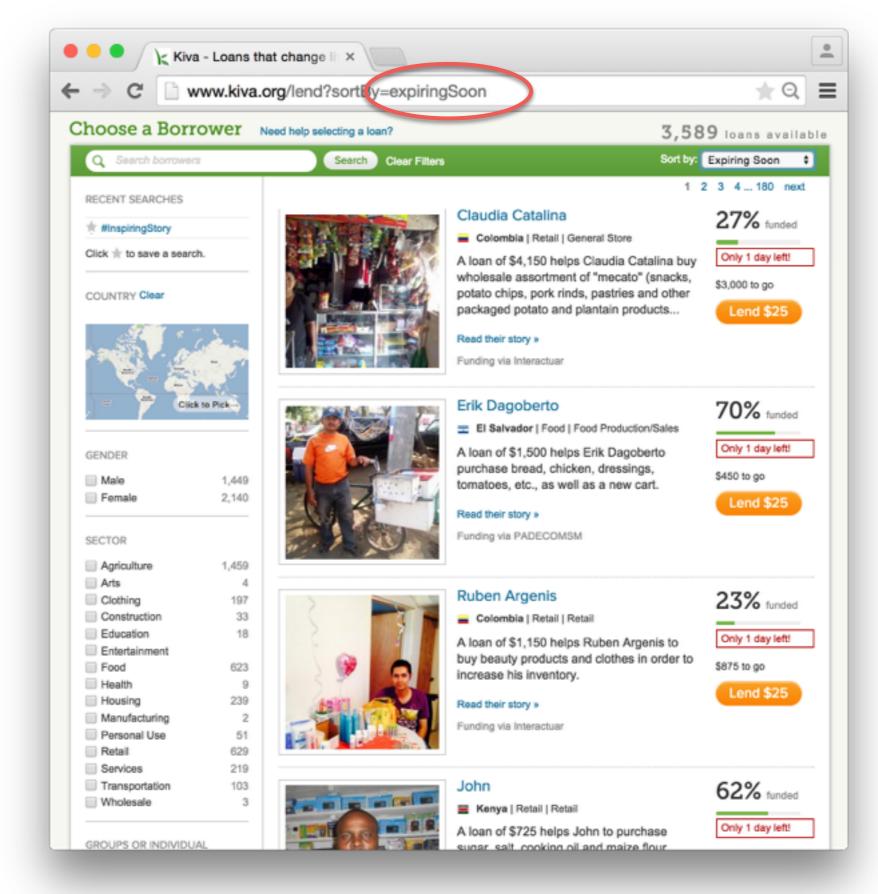
- When selecting my first borrower & loan, I was probably influenced by:
 - [Image features]
 - Loan features
 - Loan description features (unstructured text)
 - Field partner features
 - [...]





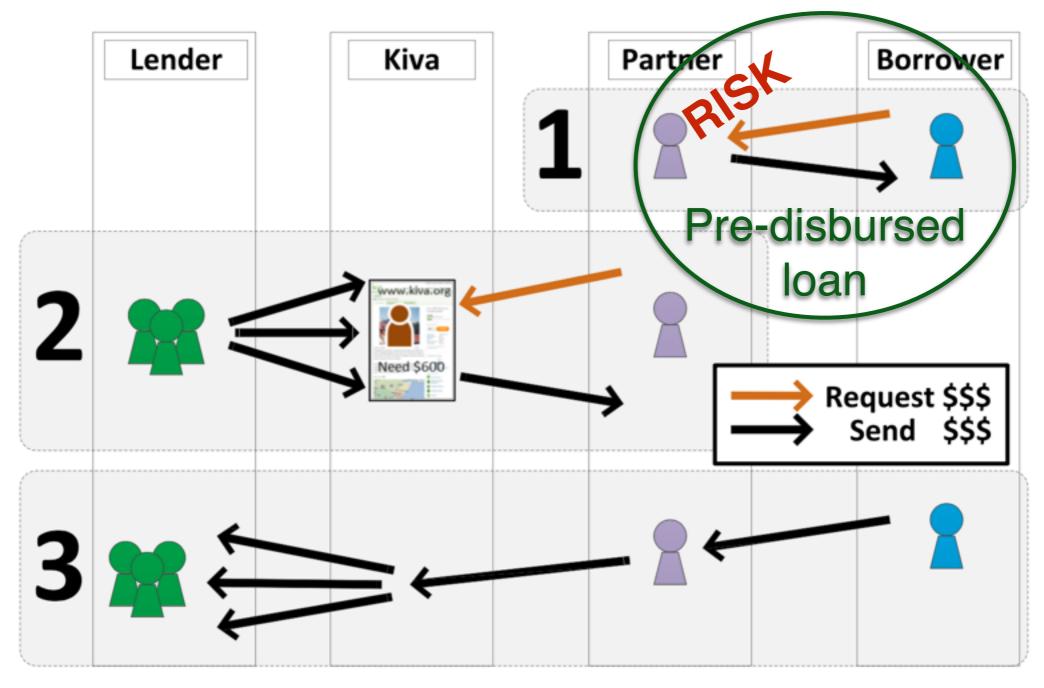
Some loans might not get fully funded, though

- Is this a bad thing?
- If so, for whom?
- If so, can we do something about this?





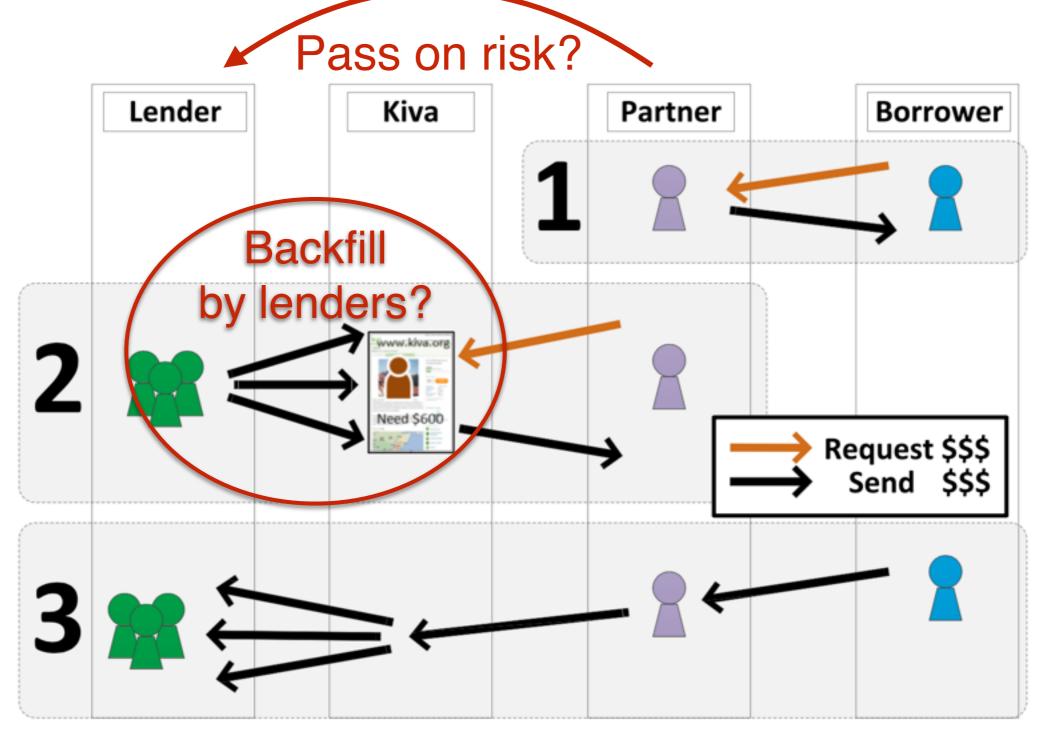
The Kiva lending process





Source: http://www.cc.gatech.edu/~joyfull/resources/2014_wsdm_kiva.pdf

Q: Will lenders backfill pre-disbursed loans?

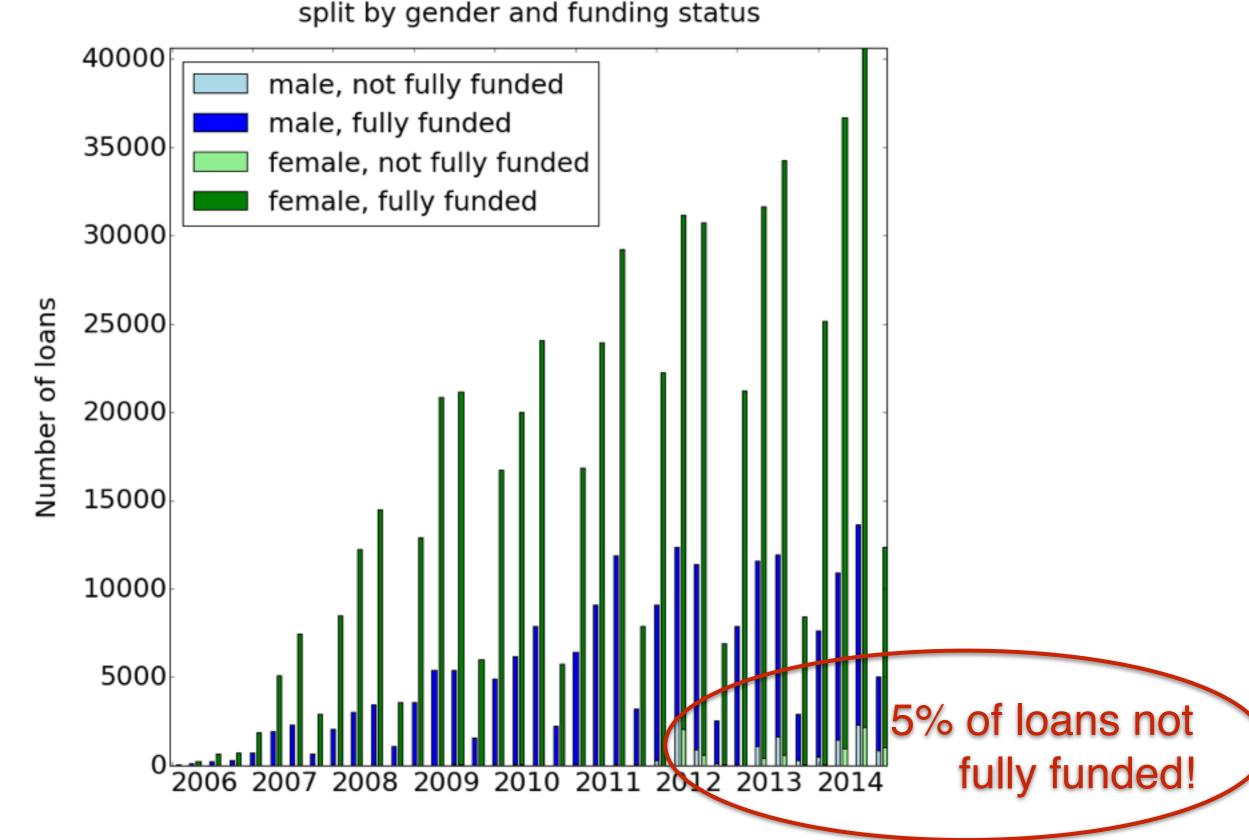




Source: http://www.cc.gatech.edu/~joyfull/resources/2014_wsdm_kiva.pdf

Non-backfilled loans appeared in 2012

Number of loans per quarter split by gender and funding status



Kiva Micro-Finance is Big

2012 through 2014, combined Lending regions and countries Borrowing countries and regions Canada \$23M **Asia \$116M USA \$145M** THE PROPERTY OF THE PROPERTY O Africa \$80M **Anonymous \$65M** South America \$66M Europe \$56M **North America \$45M** DODDOD BLOW Oceania \$16M **Europe \$5M** Asia \$6M Oceania \$3M

Solution: an end2end loan funding predictor

Train and evaluate logistic regression model using a.o. topic modelling features

GET data snapshot

http://s3.kiva.org/snapshots/kiva ds ison.zip



http://www.kiva.org/lend/844974

GET page from Kiva

4

Deploy model

GET 102n from Kiv2

GET 102n from Kiv2

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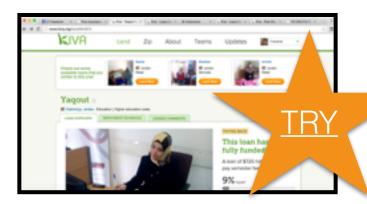
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Deployed predictor embedded in Flask web app

Get prediction from predictor

http://104.236.210.43/kivapredictor/api/v1.0/loanprediction?loanid=844974



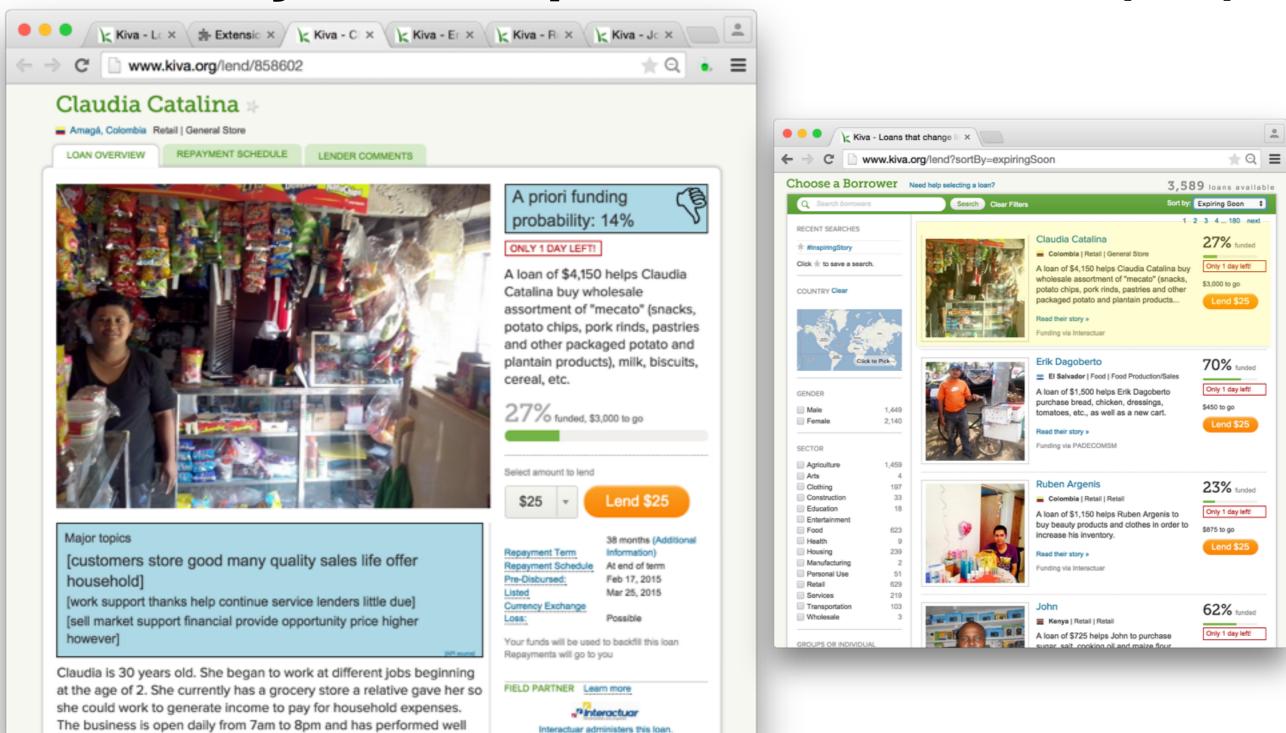
Google Chrome with self-developed plugin

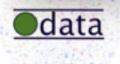
Odata

http://frederikdurant.com/projects/

http://www.kiva.org/browse

Reality check: prediction results (1/4)

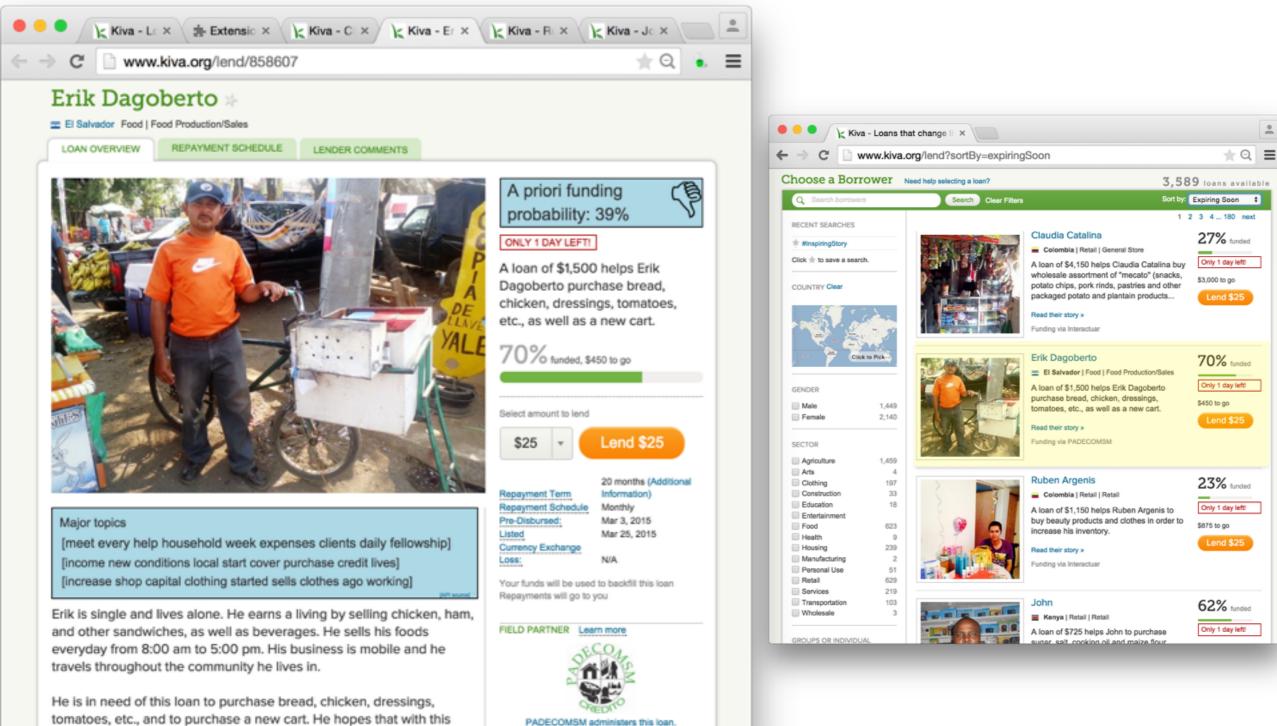


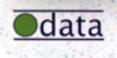


http://frederikdurant.com/projects/

docaito the competition in the coctor, generating enough income to

Reality check: prediction results (2/4)

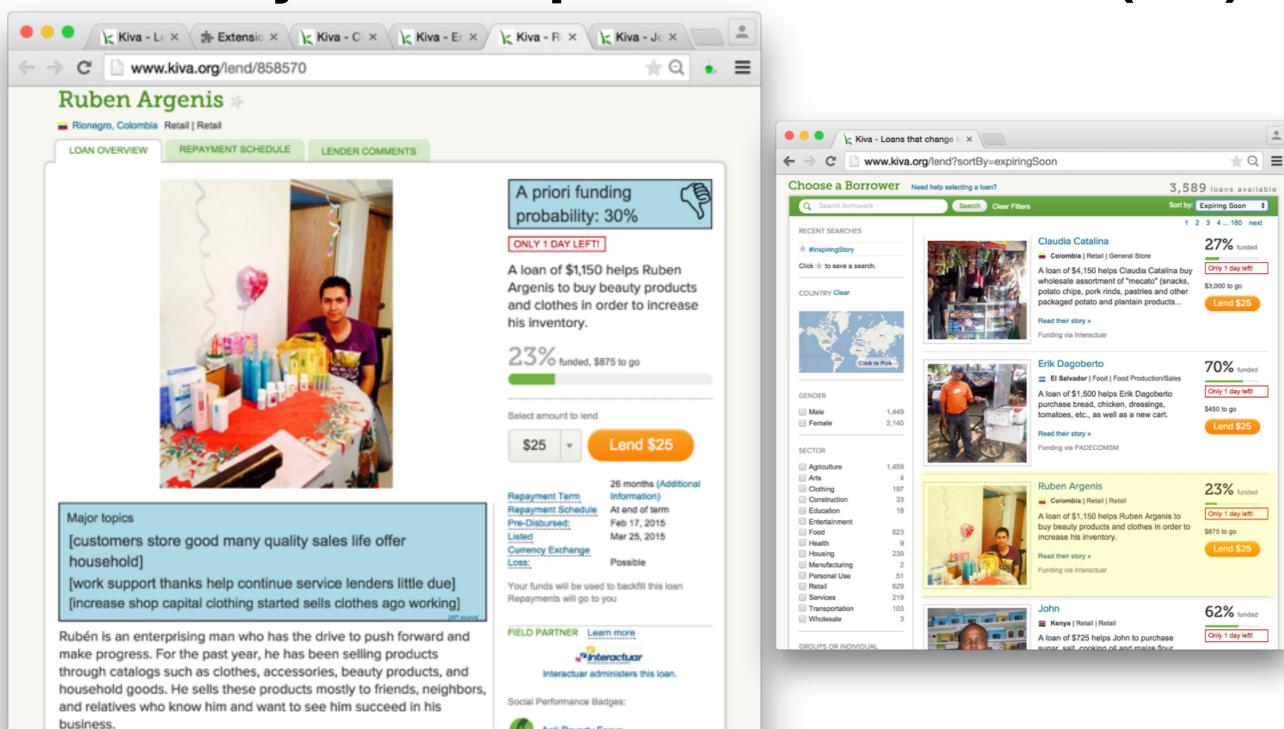




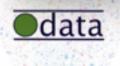
http://frederikdurant.com/projects/

investment, his productivity will increase and he will be able to cover

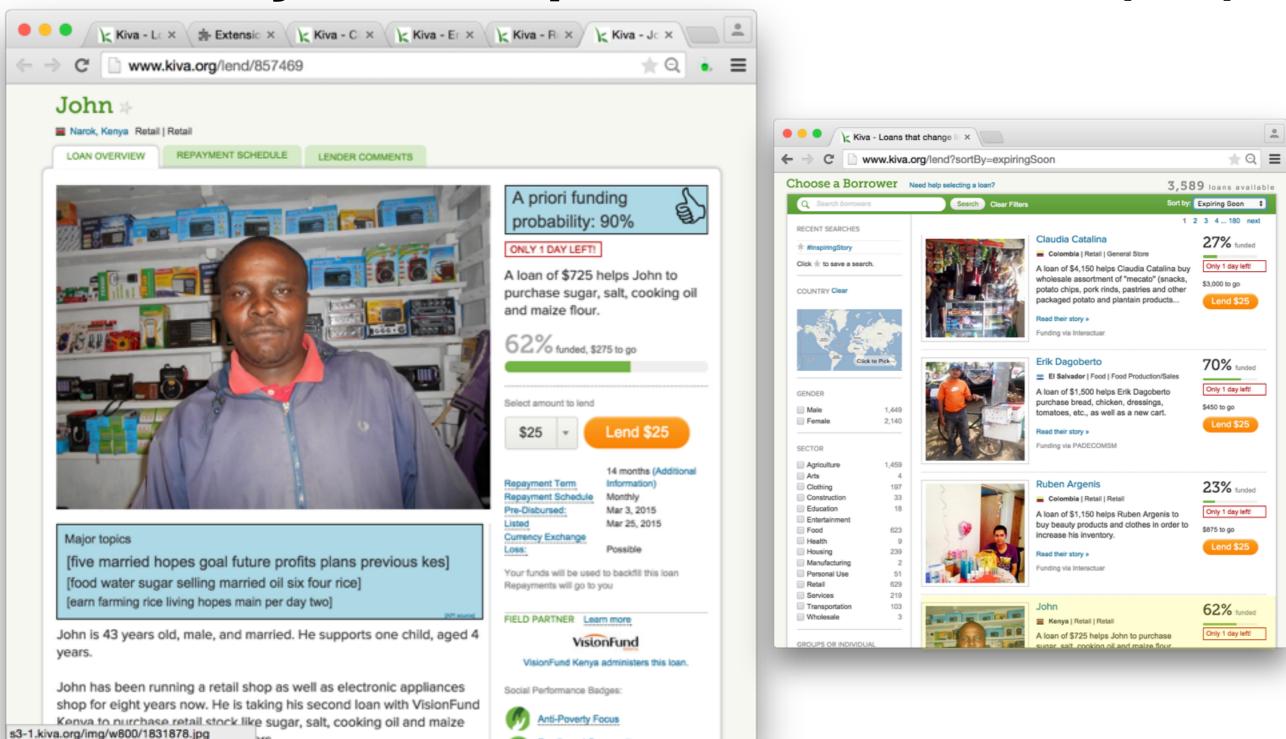
Reality check: prediction results (3/4)



Anti-Poverty Focus



Reality check: prediction results (4/4)



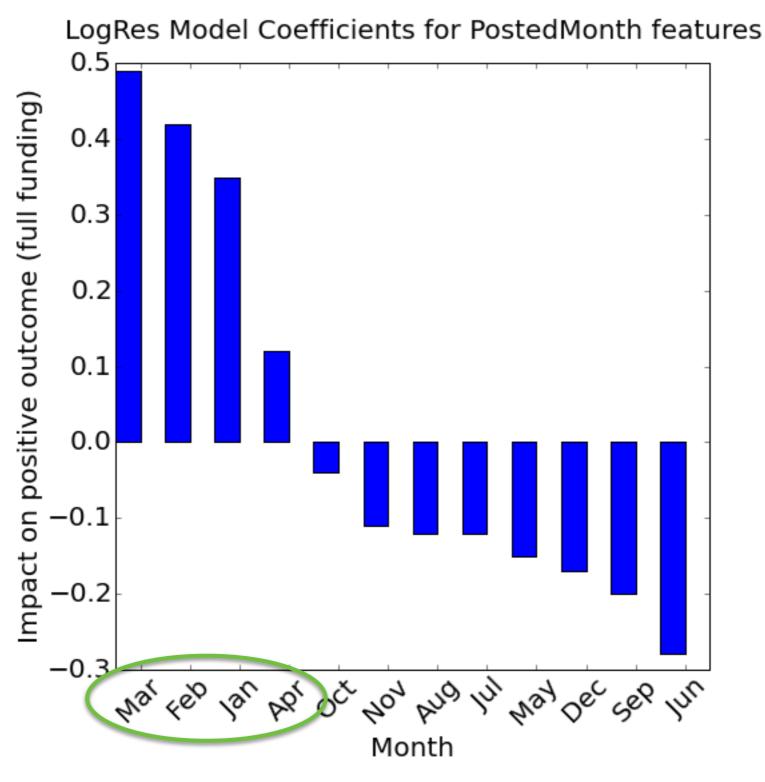


Incremental feature contributions

Feature	ROC Area Under Curve	Percentage point delta
Baseline	50.00%	-
Log10LoanAmount	76.50%	+ 26.50
20 Loan Desc. Topics	80.72%	+ 4.22
PostedMonth[JanDec]	83.55%	+ 2.83
MajorityGender	85.99%	+ 2.44
Log10NumberOfBorrowers	87.13%	+ 1.14
[Partner] LoansPosted + TotalAmountRaised	87.42%	+ 0.29
GeoLatitude + GeoLongitude	87.63%	+ 0.21
RepaymentTerm	87.77%	+ 0.14
[Partner] DelinquencyRate + Rating	87.77%	+ 0.00
BonusCreditEligibility	87.82%	+ 0.05

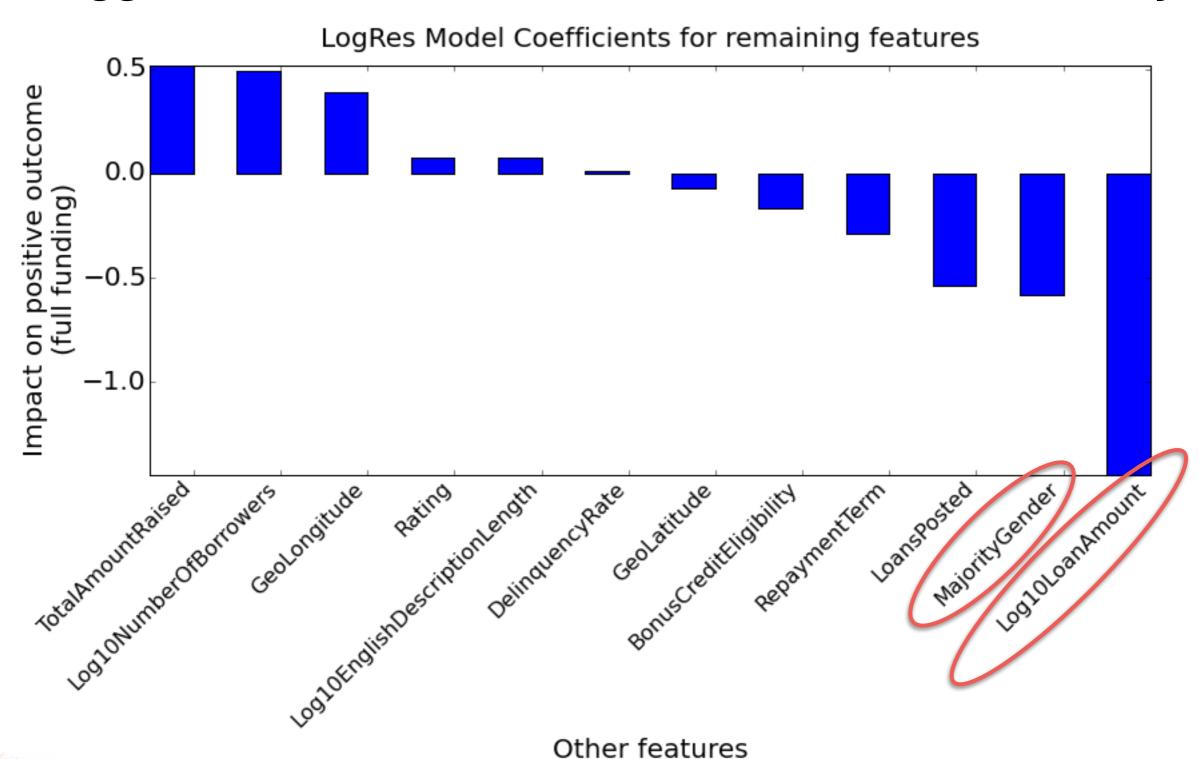


Loan posted early in year => funded more easily





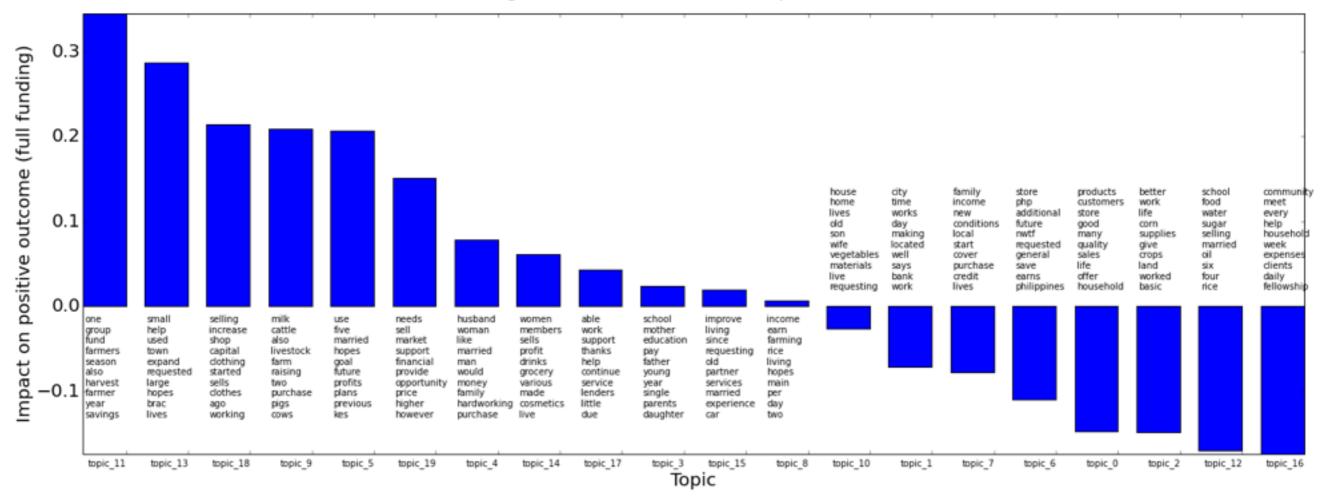
Bigger loans to male borrowers => funded less easily





Feature polarity: topics in loan descriptions







Status & follow-up

- Received very positive feedback from Kiva
- To be discussed: potential internal deployment
 - Not just a technical, also an ethical/behavioral matter!
 - The predictor can be used by different actors for different purposes
- Cool idea: improve model by adding automated trust/ sympathy detection in <u>images</u>
 - See Jenq, Pan & Theseira: "What Do Donors
 <u>Discriminate On? Evidence from kiva.org</u>"
 (based on image assessment by humans)







Thank you!



