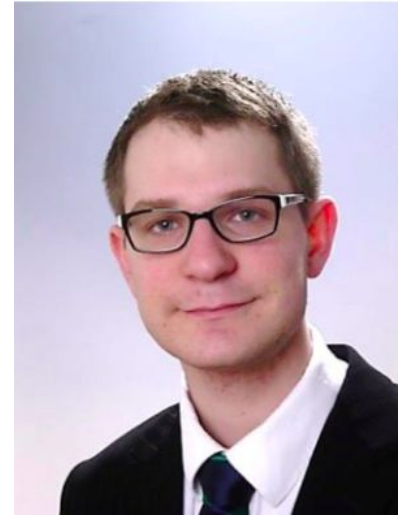


Optimizing job ad titles at Jobcloud using Google BERT and neural networks

David Furrer & Adrian Imfeld



- MSc Chemistry ETHZ
- Novartis
- MA Economics HSG
- Data Science



- Cognitive Neuroscience / Artificial Intelligence, MSc
- Co-Founder / Lead Web Developer at Bright Answer
- Moving into Data Science / Machine Learning



Business problem



Idea: Good job titles yield more views for job ads




Identify characteristics of high-quality job titles




Autocomplete tool: suggest good job title variants

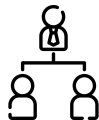
Elektro-Fachmann als **BETRIEBS-ELEKTRIKER** **100%** **(w/m)**

 aggression

 percentage

 male/female

Location, **qualification**, **CEO**, **language**, **length**



What makes a good job title?



Top 7 features for predicting view count

Contract %

Package B

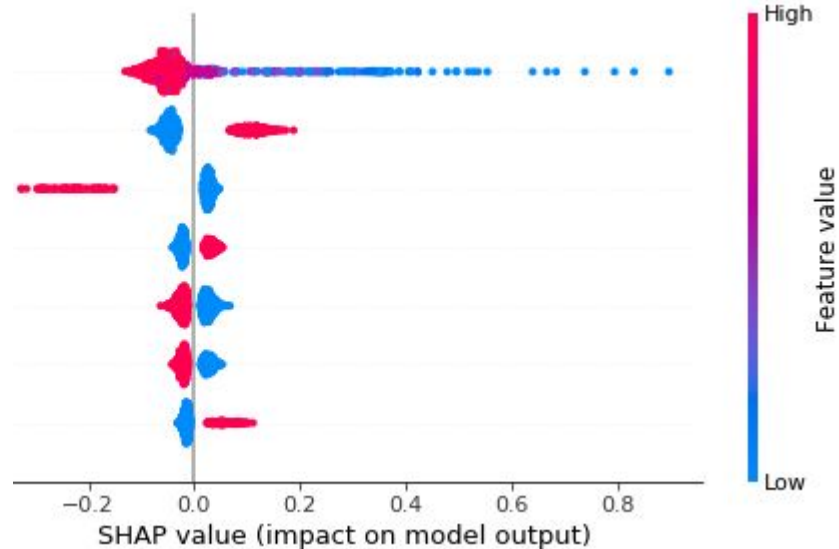
Package D

Title percent


City: Other

Package C

City: Zurich



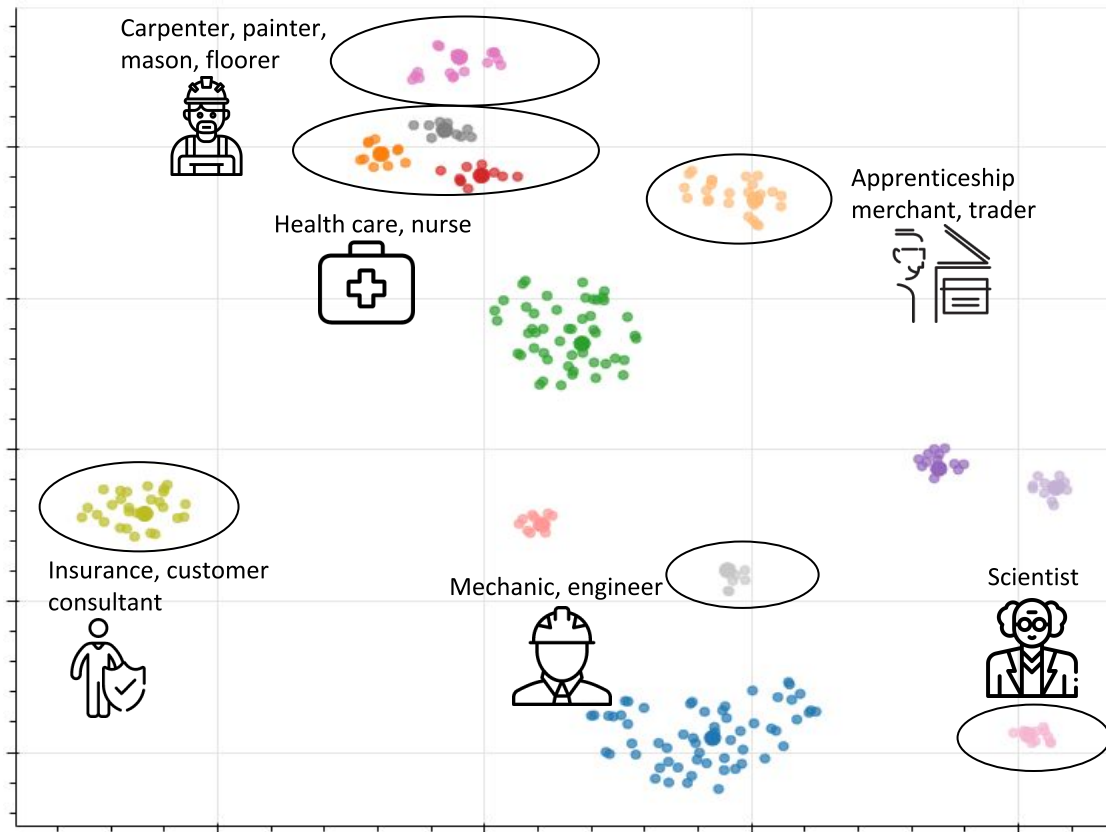
- State of the art pretrained language model for NLP
- Trained on wikipedia and BookCorpus (11k books)

Model	# Parameters
BERT-Base	110M
BERT-Large	340M
GPT-2	1.5B
Honey Bee Brain 	~1B synapses



Reference: BERT: Pre-training of Deep Bidirectional Transformers for Language Understanding (Oct 2018)

Clustering job titles with BERT embeddings



- Job titles are encoded as **BERT embeddings** (vectors)
- BERT embeddings **represent meaning** of the job titles
- BERT embeddings can be used to **cluster similar job titles**

Interactive App Demo



- Improve model (e.g. unemployment rate, company name)
- Implementation at Jobcloud

Contact



linkedin.com/in/david-furrer



furrer.david1@gmail.com



github.com/davidfurrer



medium.com/@fulowa



linkedin.com/in/aimfeld/



aimfeld80@gmail.com



github.com/aimfeld

Approach



Data exploration



Data cleaning



Feature engineering

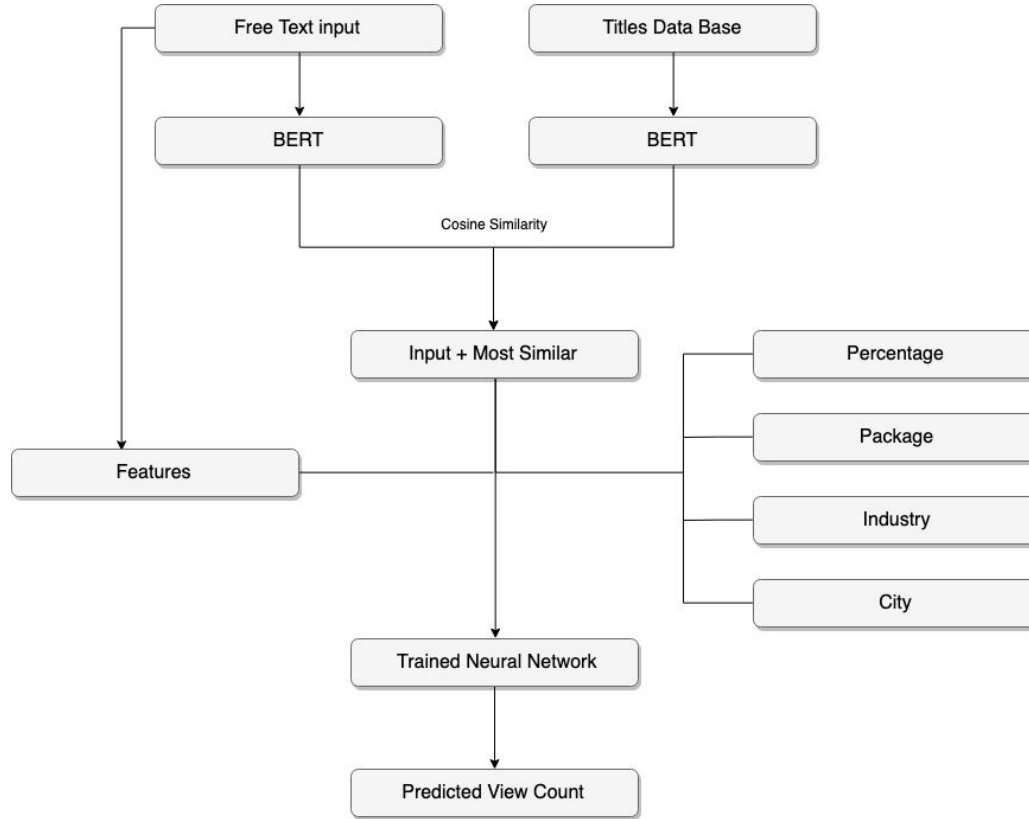


Model to predict views



Interactive app

Interactive App Demo

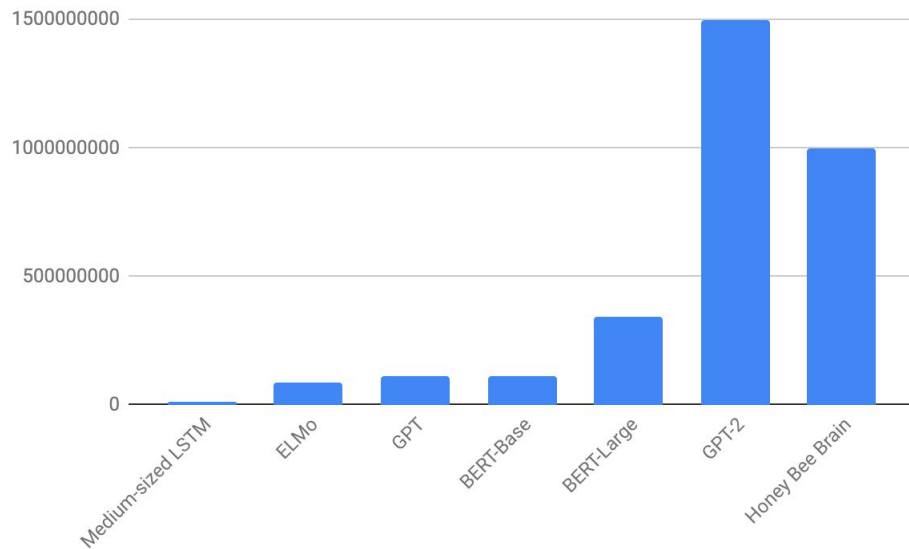


BERT = Bidirectional Encoder Representations from Transformers.



Model	# Parameters
Medium-sized LSTM	10M
ELMo	90M
GPT	110M
BERT-Base	110M
BERT-Large	340M
GPT-2	1.5B
Honey Bee Brain	~1B synapses

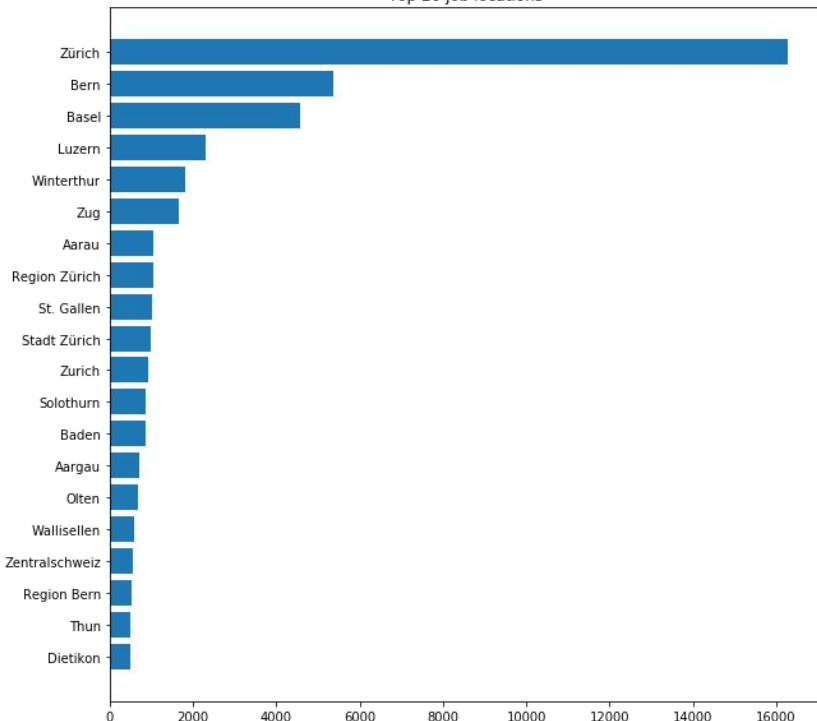
Backup



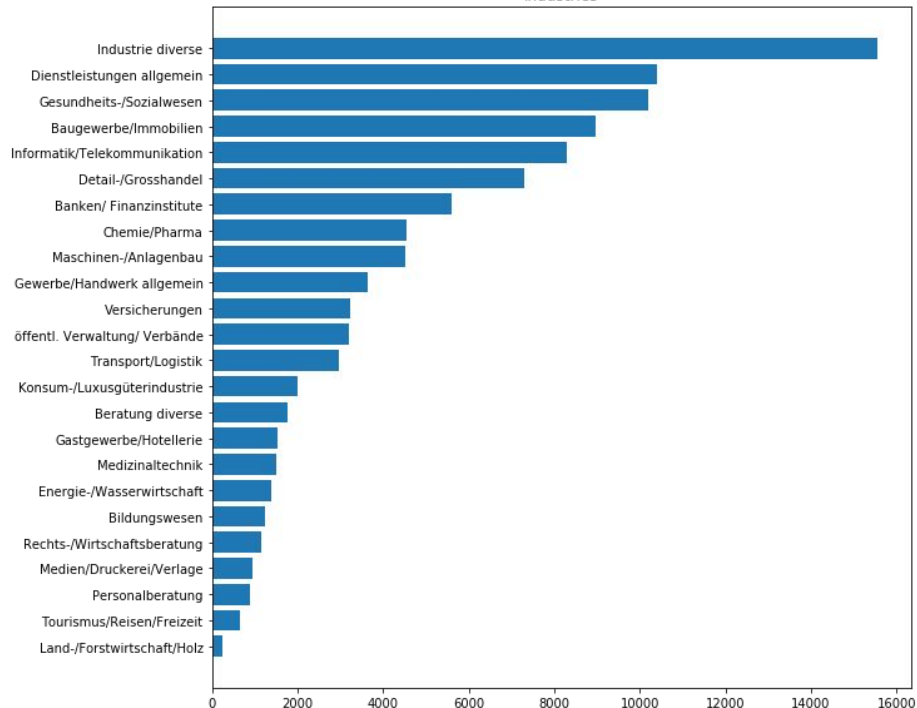
Data Description



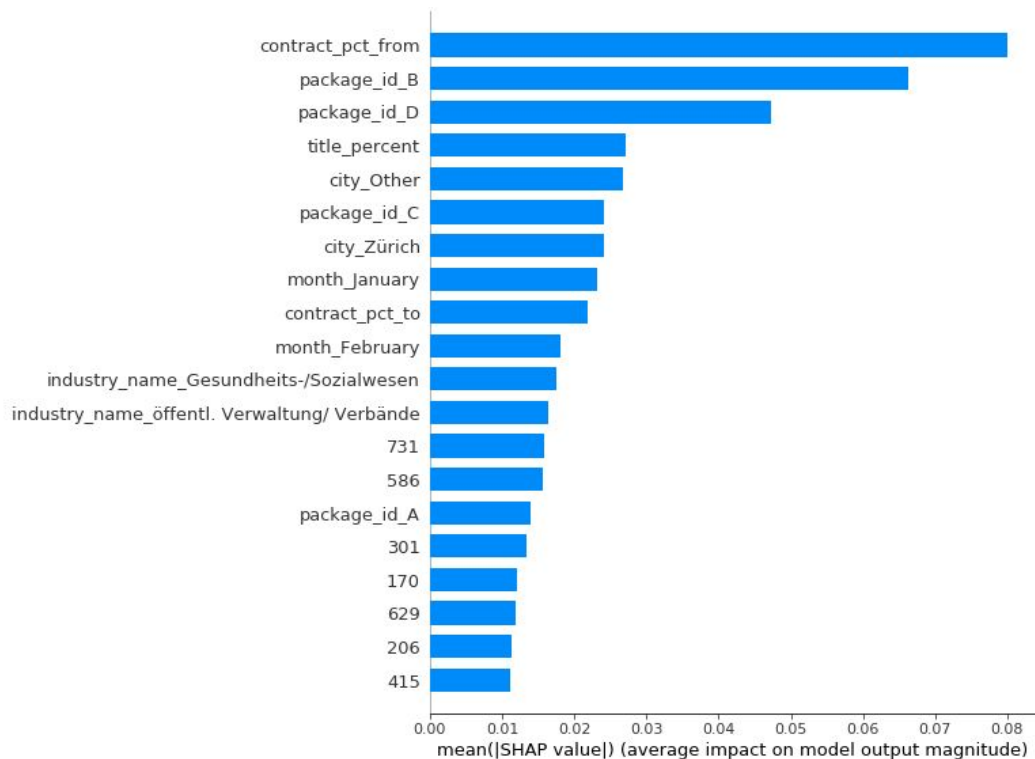
Top 20 job locations



Industries



SHAP - Feature importance



Embeddings

- Can be used for words, sentences or paragraphs
- BERT embeddings have 768 dimensions



choose a city

Zürich

choose an industry

Industrie diverse

Enter Your Job Ad Title here

Data Scientist

choose a package

A

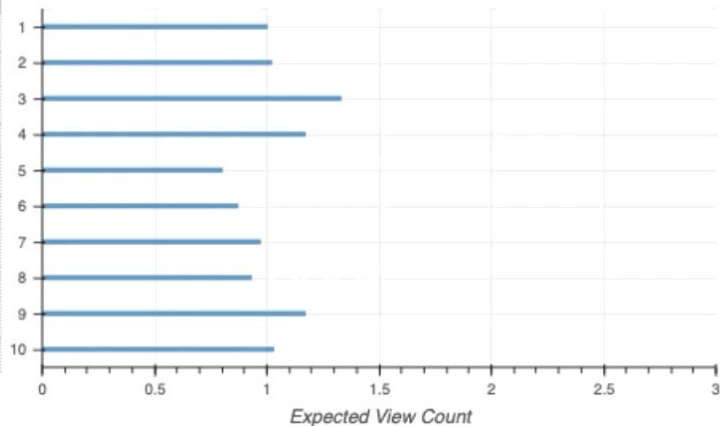
choose from %

100

choose to %

100

#	title	pred	title_no
0	Data Scientist	1	1
1	Data Scientist (80-100%)	1.020	2
2	Data Scientists (m/w) 80-100%	1.330	3
3	Data Scientist/-in (80% bis 100%)	1.170	4
4	Data Scientist Auditor/in	0.800	5
5	Data Scientist Expert (m/w)	0.870	6
6	Senior Data Scientist	0.970	7
7	Data Scientist Imaging	0.930	8
8	Data Analyst	1.170	9
9	Data Science Expert	1.030	10



Insights about using BERT



- Improved results by removing noise
- Tricky to use
- Needs a lot of computing power, RAM

	title	title_clean
0	Softwarearchitekt / Projektmanager (m/w) - All-in-One Datenmanagement in Design, Produktion und Qual	softwarearchitekt projektmanager all one datenmanagement design produktion und qual
1	Prozessingenieur Lasertechnologie - Industrialisierung innovativer Fertigungstechnologien	prozessingenieur lasertechnologie industrialisierung innovativer fertigungstechnologien
2	Softwarearchitekt / Projektmanager (m/w) - All-in-One Datenmanagement in Design, Produktion und Qual	softwarearchitekt projektmanager all one datenmanagement design produktion und qual
3	Projektleiter (m/w) - Werkzeug- oder Maschinenbau	projektleiter werkzeug oder maschinenbau
4	Fachverantwortlichen Metrologie - Produkteentwicklung und -validierung	fachverantwortlichen metrologie produkteentwicklung und validierung

Data set

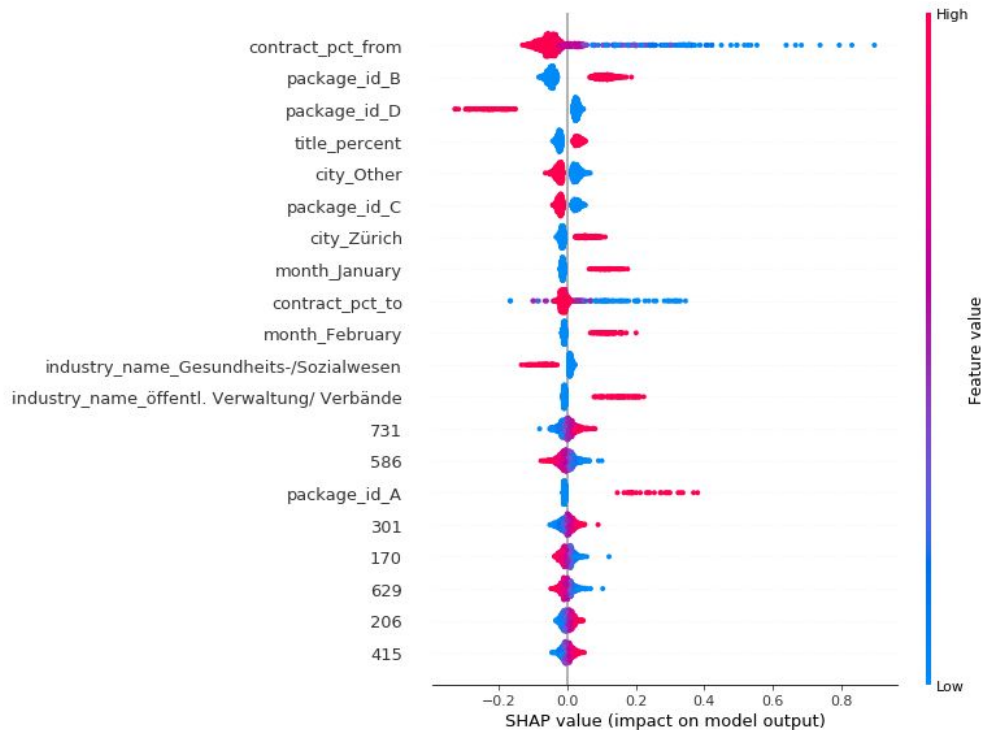


- Dataset
 - Date
 - Contract percent
 - Industry
 - Package
 - Location
 - **Job title**
 - **View count**

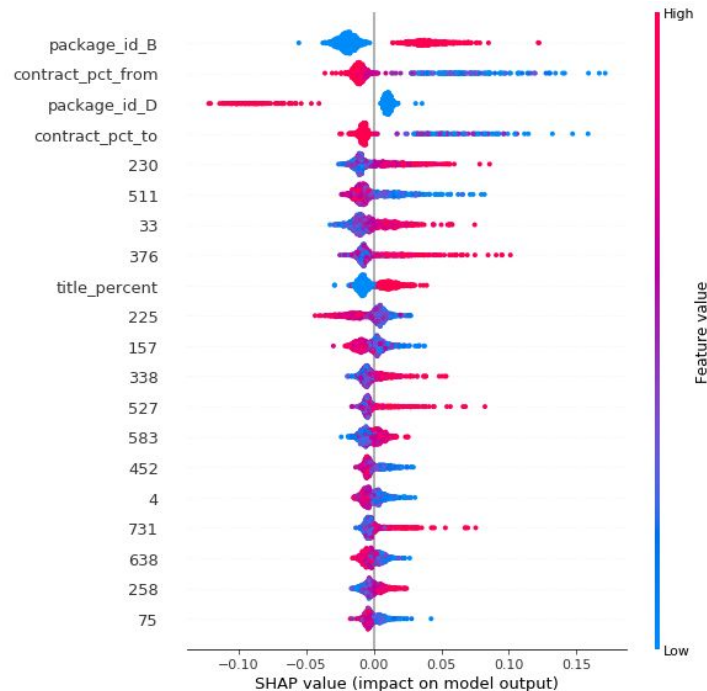
Feature importance



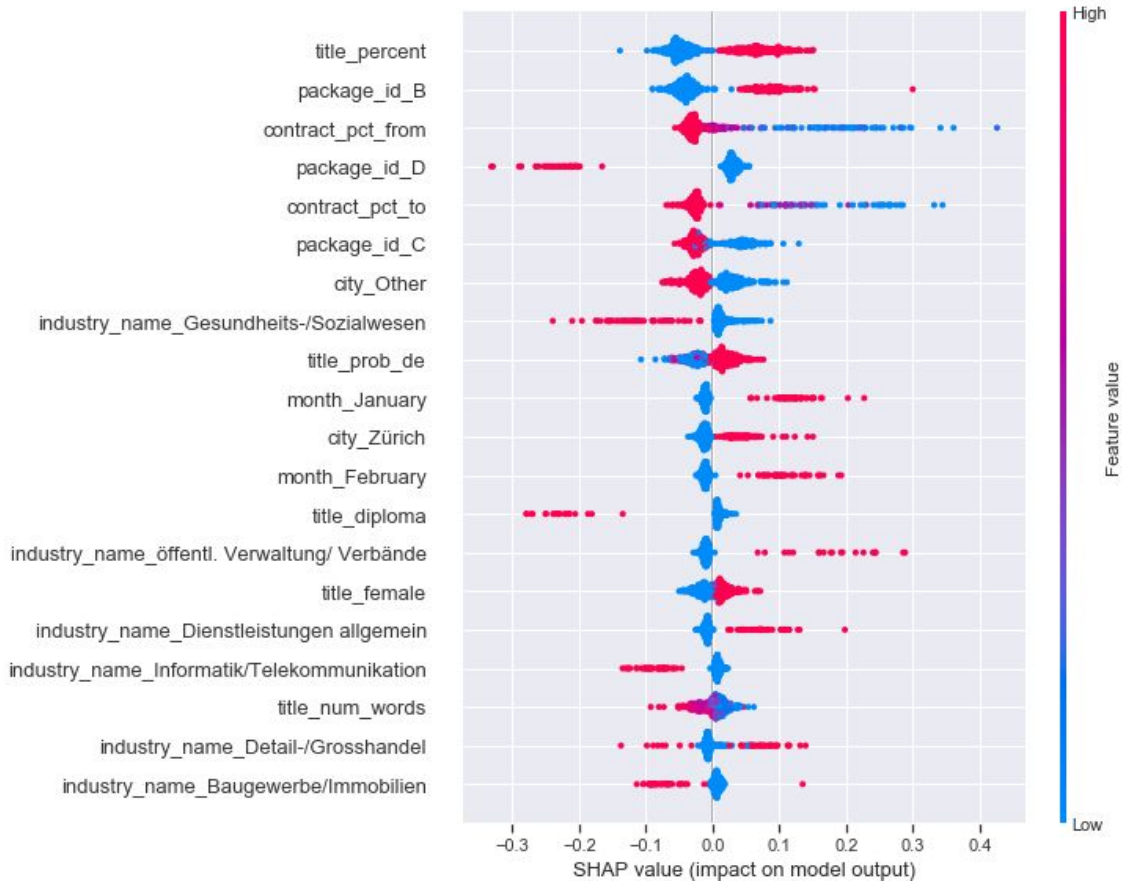
Neural network ($r^2 = .37$)



Random forest ($r^2 = .34$)



Feature importance



Random forest without BERT embeddings ($r^2 = .25$)