Probabilistic Programming for Voucher Information Extraction
Preliminary Practical Experiences
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Introduction to Skanned.com

Skanned.com provides a Voucher Scanning service for extracting information from vouchers like product lines, total amounts, payment date, sender and recipient.

Vouchers vary heavily in size, layout, purpose and content; the scan quality is occasionally suboptimal. Probabilistic programming provides an opportunity to:

- Combine domain knowledge and machine learning to effectively extract features in a systematic fashion.
- Quantify confidence in results, which is important for manual validation.

Finding Features w/Keywords

Features are usually located around identifying keywords. Keywords can be positive or negative depending on the feature to be found.

<table>
<thead>
<tr>
<th>Total Amount Excl. VAT</th>
<th>23613.00 DKK</th>
</tr>
</thead>
<tbody>
<tr>
<td>Total VAT</td>
<td>5903.25 DKK</td>
</tr>
<tr>
<td>Total Amount</td>
<td>29516.25 DKK</td>
</tr>
</tbody>
</table>

Probabilistic model below tries to infer a latent score $r$ from the vector of observed angles $\theta^+$ and distances $d^+$ from positive keywords to potential target features.

$$\begin{align*}
r & \sim \mathcal{B}(0.5,0.5) \\
w_1^+ &= (0.7,0.3) \\
\mu_1^+ &= (0, \frac{\pi}{2}) \\
w_2^+ &= (0.5,0.2,0.3) \\
\mu_2^+ &= (-\frac{\pi}{2}, -\frac{\pi}{4}, \frac{3\pi}{4}) \\
\theta^+ | r & \sim \mathcal{N}(r \cdot \bar{r}, \frac{1}{2} \sum_{j=1}^{2} \sum_{i=1}^{w_j^+ \cdot v} w_{ij}^+ \cdot v \cdot (\mu_{ij}^+, \frac{4}{\pi}) \\
d^+ | r & \sim \mathcal{N}(r \cdot \bar{d}, \frac{1}{2} \sum_{j=1}^{2} \sum_{i=1}^{w_j^+ \cdot v} w_{ij}^+ \cdot v \cdot (\mu_{ij}^+, \frac{4}{\pi})
\end{align*}$$

Practical Experiences

Sampling

- **Ease of use**
- **Precision**
- **Scalability**

Variational Inference

- **Scalability**
- **Set-up**
- **Precision**

Voucher Grouping

To provide more accurate models, to partition the voucher into groups of similar layout and style. We rely on probabilistic Latent Dirichlet Allocation (LDA) to perform the grouping, using visual (colors, lines) and textual cues (keywords).

For 1000 sample vouchers we achieved 21 topics, and our next goal is to rely on these topics to construct more precise feature extraction models.

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