

Resume Extraction

Group Name: Overloading

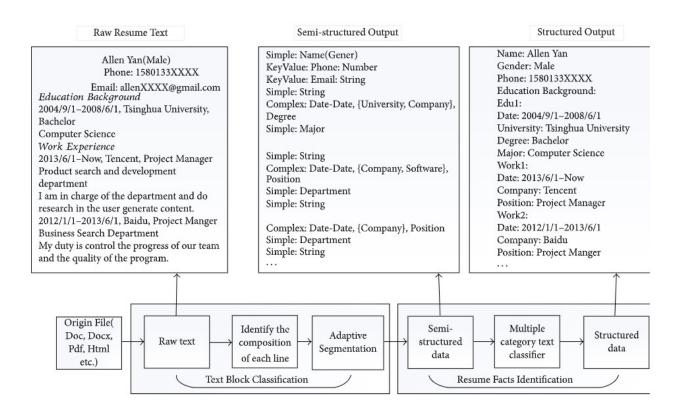
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Github Repo link: https://github.com/ardaozmen/Resume_Extraction

Problem Description



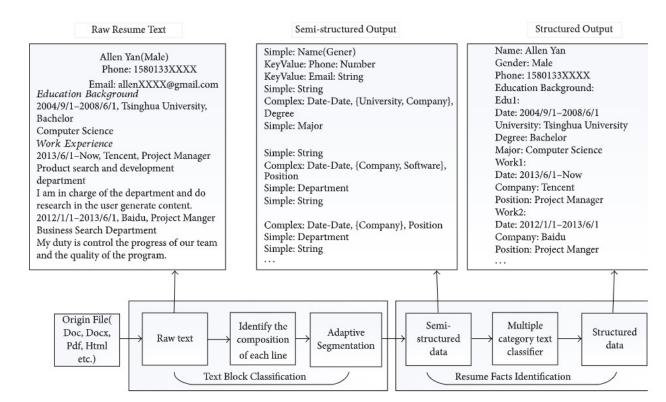
Resumes contain surfeit information that is not relevant for the HR/authority, and they have to manually process the resumes to shortlist the promising candidates for them. And, thus making the shortlisting task a herculean task for HR.



Problem Description



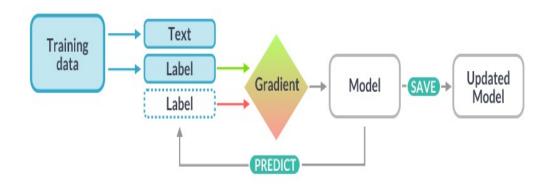
By making use of the NER(Named Entity Recognition) model of NLP this problem can be solved by finding and classifying the entities that are present in each resume into predefined classes such as person name, college name, academics information, relevant experiences, skill set, etc.



Approach



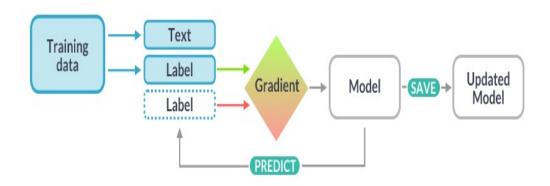
We used Spacy library, it has made advanced Natural Language Processing (NLP) much simpler in Python. Spacy provides an exceptionally efficient statistical system for named entity recognition in Python, which can assign labels to groups of tokens which are contiguous.



Approach



It provides a default model which can recognize a wide range of named or numerical entities, which include company-name, location, organization, product-name, etc to name a few. Apart from these default entities, Spacy enables the addition of arbitrary classes to the entity-recognition model, by training the model to update it with newer trained examples.



Exploratory Data Analysis



We got an average result with the available data. Increasing the number of data allows us to get more serious results.

| | content | annotation | Companies worked at | Skills | Graduation Year | College Name | Degree | Designation | Email Address | Location | Name | Years of Experienc |
|----|--|--|---|---|--------------------|--|---|---|--|---|--|---|
| 63 | Sivaganesh Selvakumar\nDevOps Consultant with | [{'label': ['Skills'], 'points': [{'start': 23 | | [{'start': 2349, 'end': 2641, 'text': 'TECHNIC | NaN | [{'start': 2257, 'end': 2290, 'text': 'Saranat | [{'start': 1990, 'end': 2043, 'text': 'Bachelo | [{'start': 1895, 'end': 1912, 'text': 'Technol | NaN | [{'start': 1934, 'end': 1940, 'text': 'Chennai | [{'start': 0, 'end': 20, 'text': 'Sivaganesh S | NaN |
| 59 | Sharan Adla\n- Email me on Indeed: indeed.com/ | [{'label': ['College Name'], 'points': [{'star | NaN | NaN | NaN | [{'start': 2421, 'end': 2449, 'text': 'Vignan | [{'start': 2416, 'end': 2418, 'text': 'SSC'}, | [{'start': 1092, 'end': 1115, 'text': 'Special | [{'start': 1801, 'end': 1841, 'text': 'indeed | NaN | [{'start': 0, 'end': 10, 'text': 'Sharan Adla'}] | [{'start': 86 'end': 90, 'text': '4yrs.'}] |
| 64 | Snehal Jadhav\nMumbai, Maharashtra - Email me | [{'label': ['Companies worked at'], 'points': | [{'start': 1794, 'end': 1799, 'text': 'Cisco ' | [{'start': 415, 'end': 1909, 'text': 'Architec | NaN | [{'start': 387, 'end': 404, 'text': 'Shivaji U | [{'start': 347, 'end': 384, 'text': 'B.E in El | [{'start': 118, 'end': 137, 'text': 'L1 networ | [{'start': 1077, 'end': 1119, 'text': 'indeed | [{'start': 148, 'end': 153, 'text': 'Mumbai'}, | [{'start': 0, 'end': 13, 'text': 'Snehal Jadha | NaN |

Recommendations



Training should be provided with larger data. Correct data must be obtained. In addition, models should continue to be produced with different approaches.

Thank You

