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- A. *Microarray Gene Expression Data*
- B. *Principal Component Analysis (PCA)*
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- D. *sfsdfsdfs*

Introduction

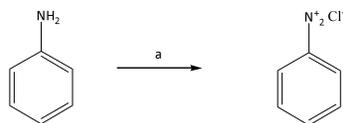
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Materials and methods

The authors should give in this section of the report a detailed account of the procedure that was followed in completing the experiment(s) discussed in the report.

Results and discussion

In this section, the authors should provide a detailed explanation of the presented work and include theoretical discussions [2-4]. The results may be presented as tables, figures, and schemes. The authors must ensure that all tables **Tab. 1, Fig. 1 / Figs 2-5**, and schemes (**Sch.1**) are cited in the text in numerical order.



Sch. 1: Synthesis of diazonium salt. Reagents and conditions: (a) NaNO₂, HCl, 0 °C, H₂O, 95 %.

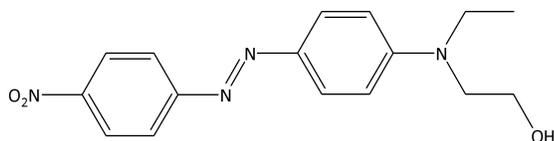


Fig. 1: Chemical structure of disperse red 1

Tab 1: Chemical composition of air

Gaz name	Symbol	% by volume
Nitrogen		78.084
Oxygen		20.9476
Argon		0.934
Carbon dioxide		0.0314
Neon		0.001818
Methane		0.0002
Helium		0.000524
Krypton		0.000114

Conclusions

The conclusions section should come in this section at the end of the presented work, before the acknowledgements.

Acknowledgements

The acknowledgements come at the end of an article after the conclusions and before the notes and references.

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