

Development of nitrogen prediction model for 320-tonne converter

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1 INTRODUCTION

Nitrogen(N) is widely known as would be controlled in various levels for different steel grades production. Specially, **ultra-low carbon(ULC)** steel requires **extremely low N level** in steelmaking.

However, it is **difficult to prevent N pick-up** from Atm. and raw Mats., because the N is one of the **interstitial elements** is easily absorbed into molten Fe.

Moreover, in the converter process which is **N removal is occurred by CO gas** formed during de-carburization, the demand to **reduce carbon emission(Carbon neutral)** would force an **unfavourable environment** for N control by reducing hot metal ratio (HMR) in the converter.

Thus, a **nitrogen prediction model(NPM)** has been developed based on **thermodynamic and kinetic approaches** using FactSage™ 8.3. In addition, in present study, it was tuned through simulation of key reactions combined **with time series data** in the converter.

2 MODEL DESCRIPTION

► Important variables of the NPM

Source	Input	Removal
1 st Variable	Molten Iron Solid Scrap purity of gas Leakage in hood	CO _(g) evolution
2 nd Variable	Temp., Comp., Time, Length, Grade, Density, Gas flow (O ₂ , Ar, N ₂ ...) Hood press., Comp. of outgas, Flow rate, Stirring, DeC Rxn., Temp., Comp.,	

① Initial condition

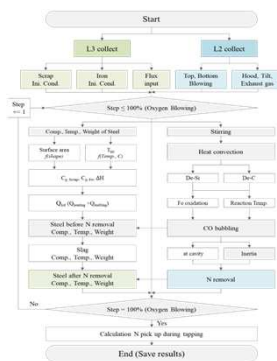
- N content in Molten iron → **Thermodynamic Calc.**

② During oxygen blowing

- N pick-up: Scrap melting/impurity of gases/leakage
- **N removal**: Decarburization
→ **Converter Simulation**: Real time Calc. of **N content**

③ Tapping (N pick-up from atm.: Kinetic Calc.)

► Outline of the NPM



3 R&D(I. Converter simulator)

► Main modules for converter simulation

① Operation data load & pre-processing

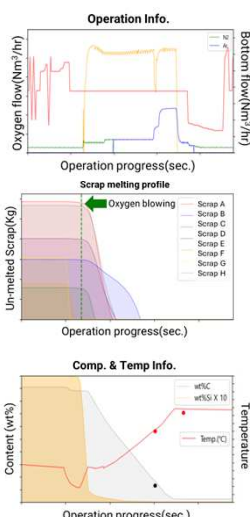
- Load Lv.3(processed) & Lv.2(time series)
- Improve data readability for Python
- Convert to **thermodynamic & kinetic data**

② Scrap melting module

- Specification of each scrap (Specific length(L), apparent density(ρ), ...)
- Melt Calc. by **heat & mass transfer model**

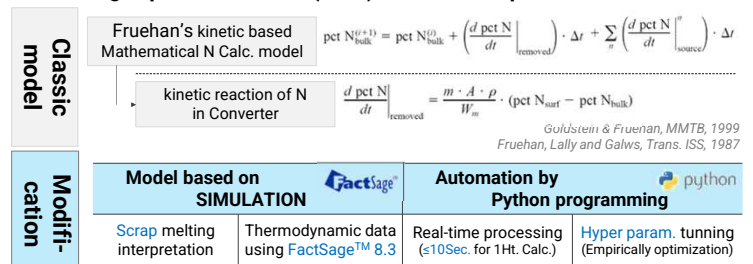
③ Oxidation & Heat balance module

- Calc. the oxidation reaction at each stage (Stage order: Si(~15%)→C(~75%)→Fe(~100%))
- Calc. the temperature of molten Fe (Factors: Oxidation, Scrap, Slag making,...)
- **Verification**: comparison with operation results

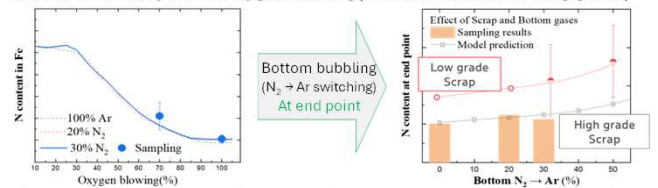


4 R&D(II. N prediction)

► The nitrogen prediction model(NPM) for converter process



► N content at end point of oxygen blowing(Effect of bottom blowing gases)

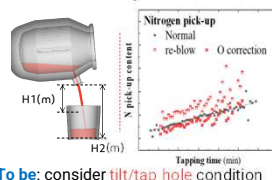


As is: Quantification of N content by **%N₂ B.B.** (N₂ B.B. after 50%: acceleration of N pick-up)

To be: Improvement of interpretation of **scrap melting** and **grade effects**

► Calc. N pick-up during tapping

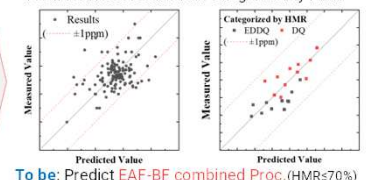
N pick-up $A \cdot \frac{P_{atm}}{100 \cdot W_{Fe}} [H_a] [(\%N_2)^2 - (\%N_{eq})^2] \cdot t_{tapping}$
Jung and Ende, MMTB, 2020



To be: consider **tilt/tap hole condition**

► Effect of hot metal ratio(HMR) on N content

As is: Quantification of N by **HMR(80~90%)**
- Pred. vs. Meas. results with categorized by HMR



To be: Predict **EAF-BF combined Proc.** (HMR≤70%)

5 CONCLUSION

A **nitrogen prediction model(NPM)** for converter process has been developed based on thermodynamic and kinetic approaches combining with operational data. Since this NPM was modified based on the **converter simulation**, it has various advantages in terms of versatility and expandability.

- (1) It is possible to **predict removal and distribution for each component** by adding calculation module using python script.
- (2) It can provide prediction results about N for **high-difficulty trial production such as EAF-BF combined Process(HMR ≤ 70%)**.

In verification through comparison with about 18,000 production results, the accuracy of the NPM's nitrogen prediction was found to be **over 90%(±1ppm)**.

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