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FISH PHYSIOLOGY, TOXICOLOGY, AND WATER QUALITY

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CAN WE PREDICT JOINT EFFECTS OF HYPOXIA AND METALS ON FISH SURVIVAL?

Michael C. Newman

ABSTRACT

Fish are suddenly exposed to hypoxic conditions during diverse events such as seiche- or turnover-related water movements, bottom water release from reservoirs, ice-over of eutrophic arctic lakes, and rapid shifts in respiration: photosynthesis associated with cultural eutrophication. In each case, chemical equilibria established under hypoxic conditions that result in metal dissolution and accumulation suddenly shift toward chemical equilibria of oxic conditions. Critical changes in speciation include those determining the free ion activity that, as expressed by the Free Ion Activity Model (FIAM), is often the most bioactive form of a dissolved metal. Metal phase can also change rapidly and, in some cases, result in a precipitate on respiratory surfaces. Exposure of fish gills to metal (and integument of larval or small fish) changes O₂ exchange dynamics. Changes in mucus quality and production and lamellae morphology decrease the amount of effective gill exchange surface and increase the diffusive layer thickness. These changes exacerbate those associated with the reduced O₂ partial pressure gradient. Consequent shifts in blood chemistry (e.g., pH and ion composition) and ventilation also affect metal transport and deposition within fish tissues. Some of these changes have immediate consequences, but others can continue for long periods after the hypoxic conditions pass. Long-term metal effects can influence fish tolerance during future hypoxic episodes.

A joint, similar action model can be applied if the parsimonious assumption is made that asphyxiation constitutes a common mode-of-action for both acute metal effects and hypoxia. Joint action models are applicable based on either conventional dose-effect or survival time approaches. Expansion of such models to a physiologically-based toxicokinetics-toxicodynamics framework (e.g., framed around the Fick equation) would be desirable, provided that model parameter requirements remain realistic. Long-term effects may be better addressed with models such as the binary logistic models used by epidemiologists.

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Metal toxicity to fish is influenced by biogeochemical, cellular, physiological, and anatomical factors. Many of these factors are also influenced by hypoxia.

Metals released from solid phases under anoxic conditions accumulate through time and, during physical mixing with oxic waters, participate in complex chemical reactions. Some (e.g., iron and manganese) can form colloidal phases, and potentially irritate fish gill surfaces. Speciation shifts for other metals that remain in the dissolved phase can elevate free metal ion activity.

Exposure of fish gills to dissolved or colloidal metals can alter gas exchange, ammonia excretion, and ion and osmotic regulation. Irritation from colloids can increase mucus production and, consequently, lower the diffusion rate of gases across the gills. According to Leland and Kuwabara (1985), production and coagulation of excess mucus during acute exposure to dissolved metals results in asphyxiation, i.e., the “coagulation film anoxia hypothesis” for metal lethality. Spaces between primary and secondary gill lamellae fill with coagulated mucus (Figure 1), increasing the effective diffusion distance. For example, colloidal aluminum precipitation on fish gills causes excessive mucus secretion that clogs interlamellar spaces (Poléo 1995).

Metals can also cause swelling of the gill epithelium, a general filling-in between the secondary lamellae, and an increase in numbers of chloride cells. The epithelium of the secondary lamellae, which is composed of two cell layers separated by an intercellular lymphoid space, swells within the lymphoid space to separate the two cell layers. Necrosis and inflammation can also occur (Daoust et al. 1984). Blasco et al. (2000) noted fusion of secondary lamellae during copper exposure in the Senegalese sole. Jagoe et al. (1996) observed thickening of the primary lamellar epithelium to such an extent that the secondary lamellae appeared absent on gills of mercury-exposed mosquitofish, greatly reducing the area available for O₂ exchange. The chloride cell volume density increased with mercury exposure at the expense of pavement cells and, because chloride cells are involved more in ion regulation than with gas exchange, the available gill area for effective respiratory exchange decreased. Chloride cell proliferation can also increase the thickness of the blood-to-water diffusion barrier (Perry 1998) and, as a consequence, impair O₂ diffusion (Greco et al. 1995). In turn, these changes produce shifts in ventilation (elevated ventilation amplitude and generally depressed ventilation frequency (Bindon et al. 1994). However, Witters et al. (1996) noted increased ventilation frequency for brown trout experiencing acute respiratory stress due to aluminum polymerization on their gills. Blood chemistry also shifts.
Figure 1. Teleost gill structure. A. Cross section of a primary lamella with three secondary lamellae extending at right angles (upward) from it. Secondary lamellae also extend downward but, in this panel, are not shown completely. Water drawn in via ventilation passes along the surface of each secondary lamella and blood within the lamella flows in the opposite direction (modified from Figure 3 of Randall (1982)). B. Two secondary lamellae with magnified areas B1 and B2. B1. Interlamellar space within which O₂ bearing water passes over the exchange surfaces of the lamellae. Each lamella has a mucus layer and specified thickness. The diffusion distance (T) influences the O₂ diffusion rate across the gill. B2. Cells covering the lamellae. Here cells at the junction of the primary and secondary lamellae are depicted. The pavement cells (white) are the principle cells involved in O₂ exchange. The larger and thicker chloride cells (dark) are involved primarily with ion regulation. Although not depicted, mucus-producing cells on the gill also respond to metal exposure. Pillar cells also reduce the gill area for O₂ exchange. Evans (1998) estimates that as much as 30% of the gill surface is directly above pillar cells and unavailable for respiratory exchange.
Hypoxia Effects

Abrupt exposure to hypoxic conditions results in predictable changes in respiration as the fish adjusts to the change in the O$_2$ partial pressure gradient across the gill. Initially, ventilation and general activity level change in an attempt to maintain adequate oxygen delivery to tissues (Wu 2002). An extreme example is the eelpout, which becomes immobile under hypoxic conditions (Fischer et al. 1992). As an example of change in ventilation, Randall (1982) reports that gill water flow of dogfish is inversely related to arterial O$_2$ concentration. Increased ventilation volume is generally achieved by large changes in ventilatory stroke volume and smaller changes in ventilation frequency (Gilmour 1998), suggesting respiratory impairment; water softness-induced increases in chloride cells also results in elevated ventilation stroke volume and lowered ventilation frequency (Bindon et al. 1994). Heart stroke volume can increase and heart rate decrease under hypoxic conditions (Randall 1982). These changes are energetically efficient, short-term means of coping with hypoxia.

Other compensatory responses to hypoxia can enhance O$_2$ diffusion rates (Gilmour 1998). Changes in ventilation and thinning of the epithelium due to increased blood pressure tend to reduce the O$_2$ diffusion barrier thickness. Increased water and blood flow can also modify the difference in O$_2$ partial pressures across the gill.

Joint Effects of Hypoxia and Metals

There is a commonality in the effects on fishes from hypoxia and metals: asphyxiation. This can be easily described with Fick’s model for O$_2$ diffusion rate (dO$_2$/dt) (Table 1):

$$\frac{dO_2}{dt} = A \cdot K_{O_2} \cdot \frac{\Delta P_{O_2}}{T}$$

where $\Delta P_{O_2} =$ difference in O$_2$ partial pressures across the gill diffusional barrier, $K_{O_2} =$ gill barrier diffusion coefficient, $A =$ the effective diffusion area of the gills, and $T =$ the diffusion barrier thickness. By definition, hypoxia changes the gill partial pressure differential ($\Delta P_{O_2}$). Modifications to ventilation and cardiac dynamics have the purpose of minimizing changes to this differential. The gill barrier diffusion coefficient ($K_{O_2}$) can be increased by mucus and metal precipitates. The morphological changes to primary and secondary lamellae can decrease the effective diffusion area (A). Combined, the morphological changes and mucus increase the diffusion barrier thickness.
Table 1. Effects summary for hypoxia and acute metal exposures contributing to asphyxiation. Terms in the left column are those of Fick’s equation for O$_2$ diffusion rate.

<table>
<thead>
<tr>
<th>Term</th>
<th>Hypoxia</th>
<th>Acute Metal Exposure</th>
</tr>
</thead>
<tbody>
<tr>
<td>A (effective diffusion area)</td>
<td>Increased slightly</td>
<td>Decreased</td>
</tr>
<tr>
<td>$K_{O2}$ (diffusion coefficient)</td>
<td>Decreased slightly</td>
<td>Increased</td>
</tr>
<tr>
<td>$\Delta P_{O2}$ (partial pressure difference)</td>
<td>Greatly increased</td>
<td>None</td>
</tr>
<tr>
<td>T (diffusion barrier thickness)</td>
<td>Decreased slightly</td>
<td>Increased</td>
</tr>
</tbody>
</table>

Both acute metal exposure and hypoxia reduce the O$_2$ diffusion rate across the gills. Some metal-induced changes, such as excessive mucus production, might be coincident with those changes associated with hypoxia. A sudden release of hypoxic, metal-rich water from a lake hypolimnion is one situation in which this might occur. In one of many such situations, Baden et al. (1995) described the simultaneous exposure of decapods to low oxygen and high manganese conditions during periodic autumnal hypoxia resulting from coastal eutrophication. Some metal-induced changes in gill morphology could occur prior to the hypoxic event of concern and predispose an individual to succumb more quickly. Any exposure to high metal concentration (or soft water (Greco et al. 1995, Perry 1998)) prior to a low oxygen event could do this.

MODELING JOINT METAL-HYPOXIA MORTALITY

Assuming that asphyxiation has a common mode-of-action suggests the application of similar joint action models to predict the combined effects of low O$_2$ and high metal conditions. For concentration-effect experiments in which the proportion of exposed individuals that die is scored at a set exposure duration, a general approach exists for incorporating the joint effects of two or more similarly acting stressors. Finney (1947) established this approach by first observing that similarly acting stressors often have concentration-response curves with identical slopes (b).

\[
\text{Probit}(P_1) = a_1 + b \times (\log C_1) \\
\text{Probit}(P_2) = a_2 + b \times (\log C_2)
\]

where, $P_1$ and $P_2$ = proportions dying after exposure to stressor 1 or 2, $a_1$ and $a_2$ = intercepts for stressors 1 and 2, and $C_1$ and $C_2$ = concentrations for stressors 1 and 2.
By combining and then re-arranging these equations, the joint effect of the two stressors (Probit \((P_1 + P_2)\)) can be predicted for the binary mixture.

\[
\log \phi_2 = \frac{a_2 - a_1}{b}
\]

\[
\text{Probit}(P_1 + P_2) = a_1 + b(\log C_1 + \phi_2(\log C_2))
\]

Unfortunately, this conventional model is not directly applicable because the slope for the metal concentration-effect model would be positive and that for the O2 concentration-effect model would be negative. That is, mortality increases as metal concentration increases or as O2 concentration decreases. Despite a deviation from convention, the two probit models can be combined and successfully rearranged. Specifically, the O2 concentration can be expressed as the absolute deviation from normoxic O2 concentration. The slope for the O2 concentration-effect model would then be positive.

A more direct model formulation might be possible if a common, physiologically-based metric of stressor intensity were available. Fick’s equation suggests O2 diffusion rate might be the most appropriate metric that reflects the common effect of both low O2 and high metal concentrations.

\[
\log \text{it}(P) \text{ or } \text{Probit}(P) = a + b_1(\log \frac{dO_2}{dt})
\]

The O2 diffusion rate would be estimated with Fick’s equation. This model is likely to be appropriate only below a threshold O2 diffusion rate; thus a threshold value might also need to be estimated. The threshold would reflect the minimum O2 diffusion rate below which the tissue O2 demands are not met. A physiologically-based, toxicokinetics-toxicodynamics model could facilitate such a formulation but, in many cases, would likely require more parameter estimates than one has the resources and time to generate.

Instead, a simpler logistic or probit model could be applied with O2 and metal concentrations as covariates \((C_1 \text{ and } C_2)\). The coefficients \((b_1 \text{ and } b_2)\) would have opposite signs in this case. The O2 concentrations for which the model would be appropriate would be those below some minimum threshold, \(i.e.,\) below normoxic conditions. Such a model would not strictly require a common mode-of-action.

\[
\log \text{it}(P) \text{ or } \text{Probit}(P) = a + b_1(\log C_1) + b_2(\log C_2)
\]

Choosing between the logistic (logit) and normal (probit) models would require a goodness-of-fit statistic such as the \(\chi^2\) statistic. Although log transformations of both concentration variables are shown here, a \(\chi^2\) statistic could also be used to select the best transformations of concentrations. Whether expressed as a logit or probit model, the model could include an interaction term if warranted.
\[
\text{logit}(P) \text{ or Probit}(P) = a + b_1(\log C_1) + b_2(\log C_2) + b_{12}(\log C_1 \cdot \log C_2)
\]

These same models could be used if previous metal exposure had occurred and the deaths associated with an hypoxic event were to be predicted. The fish might be classified with a categorical variable relative to whether it had or had not been previously exposed to high metal concentrations. This common approach taken by epidemiologists (e.g., Ahlbom 1993) to estimate relative risks for etiological factors would allow one to estimate the increase in risk of mortality under hypoxic conditions as a function of past metal exposure. The logistic model including log O\textsubscript{2} concentration and the categorical variable (previously exposed to metal or not) would be the following:

\[
\text{Logit}(P) = \ln\left(\frac{P}{1-P}\right) = a + b_1(\log C_1) + b_2(E)
\]

where \(P\) = proportion of individuals exposed to hypoxic conditions that die, \(a = \) intercept, \(b_1 = \) coefficient for the effect of log O\textsubscript{2} concentration, \(C_1 = \) oxygen concentration, \(b_2 = \) coefficient for metal exposure status effect, and \(E = \) a categorical score denoting whether an individual had (1) or had not (0) been previously exposed to high concentrations of metal. The risk of an individual previously exposed to metal relative to that of an individual with no previous metal exposure would be the following (modified from Ahlborn (1993)),

\[
e^{b_2} = \frac{\frac{P_e}{1-P_e}}{\frac{P_n}{1-P_n}} = RR
\]

where \(P_e\) and \(P_n\) = the proportion of individuals dying during hypoxic exposure with previous metal exposure (e) and non-previous metal exposure (n) respectively. The relative risk (RR) can be approximated with the estimated \(b_2\) of the logistic model fit to these data.

Another set of models would be useful if survival time were used as the effect metric instead of the proportion dying by a set time. In such a design, individuals are exposed to lethal conditions and the time required to die for each individual is recorded. A rich array of survival time methods is available (see Newman 1995 and Crane \textit{et al.} 2002). Most such methods accommodate censoring, \textit{i.e.}, a time-to-death data set in which some individuals were still alive at the end of the exposure. These methods have been applied to separate effects of oxygen (\textit{e.g.}, Dixon and Newman 1991) and metals (\textit{e.g.}, Newman 1995). Roy and Campbell (1995) used these methods for the joint effects of aluminum and zinc on Atlantic salmon juveniles. There is no apparent reason they could not be applied to the joint effects of hypoxia and metal exposure. As a particularly relevant example, Veldhuizen-Tsoerkan \textit{et al.} (1991) concluded from survival time analyses that mussels previously exposed to metals were less tolerant of anoxia. For this metric, simple parametric models can be generated.
\[
\ln TTD = a + b_1(\log C_1) + b_2(\log C_2) + \varepsilon
\]

where \( TTD = \text{time-to-death} \) and \( \varepsilon = \) an error term that has a specified distribution. Such a simple model could include \( O_2 \) and metal concentrations, or some transformation of these concentrations, as covariates to predict time-to-death during acute, joint exposures. If the situation of concern was survival during exposure to hypoxic conditions for fish with different metal exposure histories \( (i.e., \) fish with modified gill structure), a similar model could be applied in which \( O_2 \) concentration was included as a continuous variable and past metal exposure status was included as a categorical variable.

**CONCLUSION**

The parsimonious assumption that asphyxiation is a common mode-of-action for both hypoxia and metal toxicity suggests that a wide range of models could be used to predict the combined effects of hypoxia and high metal concentrations (Figure 2). Such models were described for two relevant scenarios. The first is the simultaneous exposure of fish to high metal and low \( O_2 \) concentrations. This might occur during the sudden mixing of oxic and anoxic waters. The second scenario involves an initial exposure to metals resulting in gill morphology changes that make individuals more susceptible to the lethal effects of hypoxia. This might occur in fish surviving a hypoxic/high metal event that are exposed again to low \( O_2 \) conditions.

![Diagram of potentially useful models](image)

**Figure 2.** A summary of the potentially useful models for predicting the combined effect of hypoxia and elevated metal concentrations. Models are separated into those dealing with simultaneous exposure to low \( O_2 \) and high metal concentrations, and those dealing with low \( O_2 \) tolerance after fish gills have been modified during a previous exposure to high metal concentrations. Models that can be derived using conventional concentration-effect or time-to-death data are also shown.
Models for both scenarios were discussed for data sets from conventional concentration-effect and survival time experiments. A wide array of plausible models exists, suggesting that predictive models can be developed for the combined effects of hypoxia and high metal concentrations. However, deciding which model will be the most effective requires the application of candidate models to other appropriate data sets.

REFERENCES


